

Lean Six Sigma Green Belt Training
Featuring Examples from Minitab 18



1.0 Define Phase



1.1 Overview of Six Sigma



Black Belt Training: Define Phase

1.1 Six Sigma Overview

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- 1.1.2 Six Sigma History
- 1.1.3 Six Sigma Approach Y = f(x)
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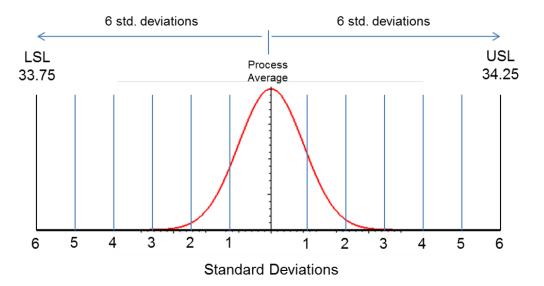


1.1.1 What is Six Sigma



What is Six Sigma?

- What is "sigma"?
 - In statistics, **sigma** (σ) refers to "standard deviation," which is a measure of variation.
 - You will come to learn that variation is the enemy of any quality process. We need to understand, manage, and minimize process variation.
- What is "Six Sigma"?
 - Six Sigma is an aspiration or goal of process performance.
 - A Six Sigma "goal" is for a process average to operate approximately 6σ away from customer's high and low specification limits.

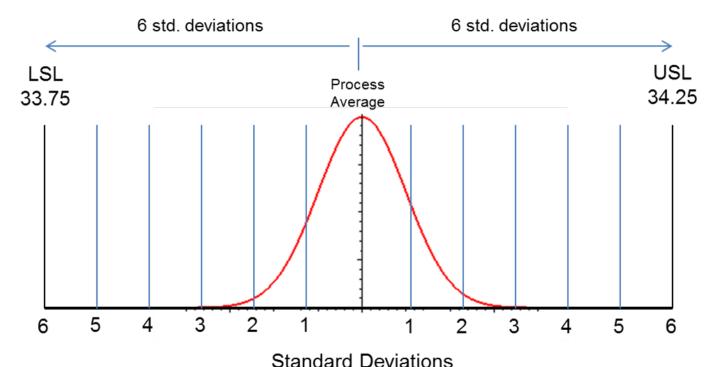




What is Six Sigma?

• A process whose average is about 6σ away from the customer's high and low specification limits has abundant room to "float" before approaching the customer's specification limits.

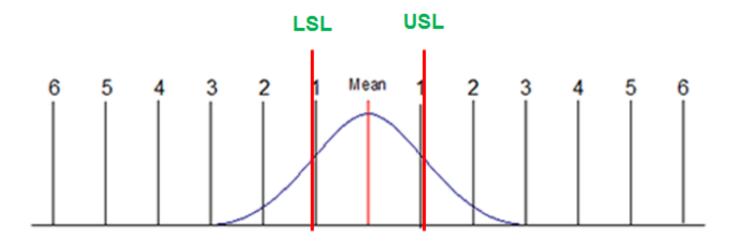
• A Six Sigma process only yields 3.4 defects for every million opportunities! In other words, 99.9997% of the products are defect-





free!

- Sigma level measures how many "sigma" there are between your process average and the nearest customer specification.
- Let us assume that your customers upper and lower specifications limits (USL & LSL) were narrower than the width of your process spread.
- The USL & LSL below stay about 1 standard deviation away from the process average. Therefore, this process operates at **1 sigma**.





A process operating at 1 sigma has a defect rate of approximately 70%.



- This means that the process will generate defect-free products only 30% of the time.
- What about processes with more than 1 sigma level?
- A higher sigma level means a lower defect rate.
- Let us take a look at the defect rates of processes at different sigma levels.



- This table shows each sigma level's corresponding defect rate and DPMO (defects per million opportunities).
- The higher the sigma level, the lower the defective rate and DPMO.

Sigma Level	Defect Rate	DPMO
1	69.76%	697612
2	30.87%	308770
3	6.68%	66810
4	0.62%	6209
5	0.023%	232
6	0.00034%	3.4

These Defect Rates Assume a 1.5 sigma shift

How does this translate into things you might easily relate to?



- Let us take a look at processes operating at 3 sigma.
- 3 sigma processes have a defect rate of approximately 7%. What would happen if processes operated at 3 sigma?
 - Virtually no modern computer would function*.
 - 10,800,000 health care claims would be mishandled each year.
 - 18,900 US savings bonds would be lost every month.
 - 54,000 checks would be lost each night by a single large bank.
 - 4,050 invoices would be sent out incorrectly each month by a modest-sized telecommunications company.
 - 540,000 erroneous call details would be recorded each day from a regional telecommunications company.
 - 270 million erroneous credit card transactions would be recorded each year in the United States.



- What if processes operated with 1% defect rate?
 - 20,000 lost articles of mail per hour*.
 - Unsafe drinking water almost 15 minutes per day.
 - 5,000 incorrect surgical operations per week.
 - Short or long landings at most major airports each day.
 - 200,000 wrong drug prescriptions each year.
 - No electricity for almost 7 hours per month.
- Even at 1% defect rate, some processes would be unacceptable to you and many others.
- So what is Six Sigma?
 - Sigma level is the measure!
 - Six is the goal!



(* Implementing Six Sigma – Forest W. Breyfogle III)

What is Six Sigma: The Methodology

- Six Sigma itself is the goal, not the method.
- In order to achieve Six Sigma, you need to improve your process performance by:
 - Minimizing the process variation so that your process has enough room to fluctuate within customer's spec limits
 - Shifting your process average so that it is centered between your customer's spec limits.
- Accomplishing these two process improvements (along with stabilization and control), you can achieve Six Sigma.
- DMAIC is the systematic methodology prescribed to achieve Six Sigma.



What is Six Sigma: The Methodology

- DMAIC is a systematic and rigorous methodology that can be applied to <u>any</u> process in order to achieve Six Sigma.
- It consists of 5 phases of a project:
 - **D**efine
 - Measure
 - Analyze
 - Improve
 - Control.
- You will be heavily exposed to many concepts, tools, and examples of the DMAIC methodology through this training.
- You will be capable of applying the DMAIC methodology to improve the performance of any process at the completion of the curriculum.



1.1.2 Six Sigma History



Toyota

Pricipals of Lean Manufacturing Started Taking Shape

Lean Six Sigma

History & Timeline Bombardier Nokia Sony Motorola & Bill Smith Develops Six Sigma at Motorola. It's Crane **Texas Instruments** Adopted & Supported by CEO Bob Galvin Polaroid Dupont Amex Ford Lockheed Martin 1985 Present 1900 1975 1980 1990 1995 2000 2005 2010 Merrill Lynch Bank of America Etc. **GE Refines & Operationalizes Six Sigma** Boeing **General Electric** Ford & Allied Signal



- The "Six Sigma" terminology was originally adopted by Bill Smith at Motorola in the late 1980s as a quality management methodology.
- As the "Father of Six Sigma," Bill forged the path for Six Sigma through Motorola's CEO Bob Galvin who strongly supported Bill's passion and efforts.
- Starting from the late 1980s, Motorola extensively applied Six Sigma as a process management discipline throughout the company, leveraging Motorola University.
- In 1988, Motorola was recognized with the prestigious Malcolm Baldrige National Quality Award for its achievements in quality improvement.



- Six Sigma has been widely adopted by companies as an efficient way of improving the business performance since General Electric implemented the methodology under the leadership of Jack Welch in the 1990s.
- As GE connected Six Sigma results to its executive compensation and published the financial benefits of Six Sigma implementation in their annual report, Six Sigma became a highly sought-after discipline of quality.

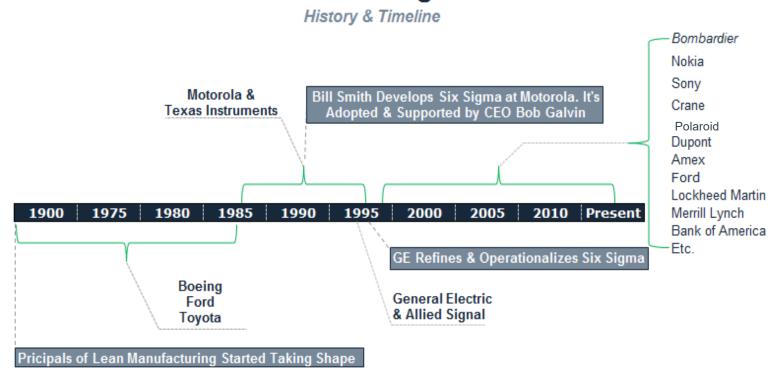


- Most Six Sigma programs cover the aspects, tools, and topics of Lean or Lean Manufacturing.
- The two work hand in hand, benefitting each other.
 - Six Sigma focuses on minimizing process variability, shifting the process average, and delivering within customer's specification limits.
 - Lean focuses on eliminating waste and increasing efficiency.
- Lean and its popularity began to form and gain significant traction in the mid 1960s with the Toyota initiative "TPS" or Toyota Production System.
- The concepts and methodology of Lean, however, were fundamentally applied much earlier by both Ford and Boeing in the early 1900s.



• Despite the criticism and immaturity of Six Sigma in many aspects, its history continues to be written with every company and organization striving to improve its business performance.

Lean Six Sigma





1.1.3 Six Sigma Approach



- The Six Sigma approach to problem solving uses a transfer function.
- A transfer function is a mathematical expression of the relationship between the inputs and outputs of a system.
- Y = f(x) is the relational transfer function that is used by all Six Sigma practitioners.
- It is absolutely critical that you understand and embrace this concept.

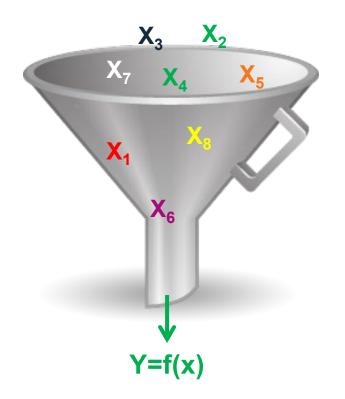


- "Y" refers to the measure or output of a process.
 - Y is usually your primary metric
 - Y is the measure of process performance that you are trying to improve.
- f(x) means "function of x."
 - x's are factors or inputs that affect the Y
- Combined, the Y = f(x) statement reads "Y is a function of x."
- In simple terms: "My process performance is dependent on certain x's."
- The objective in a Six Sigma project is to identify the critical x's that
 have the most influence on the output (Y) and adjust them so that the Y
 improves.



- Let us look at a simple example of a pizza delivery company that desires to meet customer expectations of on-time delivery.
 - Measure = on-time pizza deliveries
 - Y = percent of on-time deliveries
 - f(x) would be the x's or factors that heavily influence timely deliveries
 - x1: might be traffic
 - x2: might be the number of deliveries per driver dispatch
 - x3: might be the accuracy of directions provided to the driver
 - x4: might be the reliability of the delivery vehicle
 - etc.
- The statement Y = f(x) in this example will refer to the proven x's determined through the steps of a Six Sigma project.





- With this approach, all potential x's are evaluated throughout the DMAIC methodology.
- The x's should be narrowed down until the vital few x's that significantly influence "on-time pizza deliveries" are identified!



- This approach to problem solving will take you through the process of determining all potential x's that **might** influence on-time deliveries and then determining through measurements and analysis which x's **do** influence on-time deliveries.
- Those significant x's become the ones used in the Y = f(x) equation.
- The Y = f(x) equation is a very powerful concept and requires the ability to measure your output and quantify your inputs.
- Measuring process inputs and outputs is crucial to effectively determining the significant influences to any process.



1.1.4 Six Sigma Methodology



Six Sigma Methodology

- Six Sigma follows a methodology that is conceptually rooted in the principles of a five-phase project.
- Each phase has a specific purpose and specific tools and techniques that aid in achieving the phase objectives.
- The 5 phases of DMAIC:
 - 1. Define
 - 2. Measure
 - 3. Analyze
 - 4. Improve
 - 5. Control



Six Sigma Methodology



 Define and quantify the problem and objective



 Baseline process, validate measures, and identify all possible x's and dependent Y's



 Analyze and validate causes identifying critical factors



Develop solutions



 Implement and sustain solutions



Six Sigma Methodology: Define Phase

- The goal of the **Define** phase is to establish a solid foundation and business case for a Six Sigma project.
- Define is arguably the most important aspect of any Six Sigma project.
- All successful projects start with a current state challenge or problem that can be articulated in a quantifiable manner.
 - It is not enough to just know the problem, you must quantify it and also determine the goal.
- Once problems and goals are identified and quantified, the rest of the define phase will be about valuation, team, scope, project planning, timeline, stakeholders, Voice Of the Customer (VOC), and Voice Of the Business (VOB).



Six Sigma Methodology: Define Phase

Define Phase Tools and Deliverables

- Project Charter Establish the:
 - Business Case
 - Problem Statement
 - Project Objective
 - Project Scope
 - Project Timeline
 - Project Team.
- Stakeholder Assessment
- High-Level Pareto Chart Analysis
- High-Level Process Map
- VOC/VOB and CTQs Identified and Defined
- Financial Assessment



Six Sigma Methodology: Measure Phase

- The goal of the **Measure** phase is to gather baseline information about the process (process performance, inputs, measurements, customer expectations etc.).
- Throughout the Measure phase you will seek to achieve a few important objectives:
 - Gather All Possible x's
 - Assess Measurement System and Data Collection Requirements
 - Validate Assumptions
 - Validate Improvement Goals
 - Determine COPQ (Cost of Poor Quality)
 - Refine Process Understanding
 - Determine Process Stability
 - Determine Process Capability.



Six Sigma Methodology: Measure Phase

Measure Phase Tools and Deliverables

- Process Maps, SIPOC, Value Stream Maps
- Failure Modes and Effects Analysis (FMEA)
- Cause-and-Effect Diagram
- XY Matrix
- Six Sigma Statistics
 - Basic Statistics
 - Descriptive Statistics
- Measurement Systems Analysis
 - Variable and/or Attribute Gage R&R
 - Gage Linearity and Accuracy or Stability
- Basic Control Charts
- Process Capability (Cpk, Ppk) and Sigma Levels
- Data Collection Plan



Six Sigma Methodology: Analyze Phase

- The Analyze phase is all about establishing verified drivers.
- In the DMAIC methodology, the Analyze phase uses statistics and higher-order analytics to discover relationships between process performance and process inputs (in other words, what are the root causes or drivers of the improvement effort).
- Ultimately, the Analyze phase establishes a reliable hypothesis for improvement solutions.
 - Establish the Transfer Function Y = f(x)
 - Validate the List of Critical x's and Impacts
 - Create a Beta Improvement Plan (e.g., pilot plan).



Six Sigma Methodology: Analyze Phase

Analyze Phase Tools and Deliverables

- The Analyze phase is about proving and validating critical x's using the appropriate and necessary analysis techniques. Examples include:
 - Hypothesis Testing
 - Parametric and Non-Parametric
 - Regression
 - Simple Linear Regression
 - Multiple Linear Regression
- The Analyze phase is also about establishing a set of solution hypotheses to be tested and further validated in the Improve phase.



Six Sigma Methodology: Improve Phase

- The goal of the **Improve** phase is. . .you guessed it! "make the improvement." Improve is about designing, testing, and implementing your solution.
- To this point you have defined the problem and objective of the project, brainstormed possible x's, analyzed and verified critical x's. Now it's time to make it real!
 - Statistically Proven Results from Active Study/Pilot
 - Improvement/Implementation Plan
 - Updated Stakeholder Assessment
 - Revised Business Case with Return on Investment (ROI)
 - Risk Assessment/Updated FMEA
 - New Process Capability and Sigma.



Six Sigma Methodology: Improve Phase

- Improve Phase Tools and Deliverables
 - Any Appropriate Tool from Previous Phases
 - Design of Experiment (DOE)
 - Full Factorial
 - Fractional Factorial
 - Pilot or Planned Study Using:
 - Hypothesis Testing
 - Valid Measurement Systems
 - Implementation Plan



Six Sigma Methodology: Control Phase

- The last of the 5 core phases of the DMAIC methodology is the Control phase.
- The goal of the Control phase is to establish automated and managed mechanisms to maintain and sustain your improvement.
- A successful control plan also establishes a reaction and mitigation plan as well as an accountability structure.



Six Sigma Methodology: Control Phase

Control Phase Tools and Deliverables

- Statistical Process Control (SPC/Control Charts)
 - IMR, XbarS, XbarR, P, NP, U, C etc.
- Control Plan Documents
 - Control Plan
 - Training Plan
 - Communication Plan
 - Audit Checklist
- Lean Control Methods
 - Poka-Yoke
 - Five-S
 - Kanban



Six Sigma Methodology

Six Sigma DMAIC Roadmap



- Goal: Problem Statement, Objective, Business Case, Project Scope, Team
- Main Tools: Project Charter, Pareto, Process Maps

Measure

- Goal: Brainstorm/Prioritize Possible x's, Validate measurement, Capability
- Tools: Basic Statistics, C & E, XY Matrix, Capability Analysis, MSA, Process Maps, Control Charts

Analyze

- Goal: Identify critical x's
- **Tools**: Hypotheses Tests (Normal/Non Normal), Regression and Correlation

Improve

- Goal: Design, Test, and Implement Improvement
- Tools: DOE, Implementation/Change/Communication Plan

Control

- Goal: Lock-in the Improvement
- Tools: Control Plan, Poka-Yoke, SPC, SOPs, Training Plans etc.



1.1.5 Roles and Responsibilities



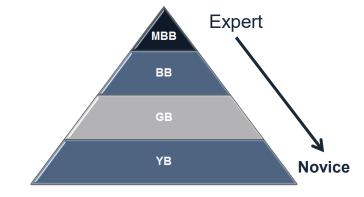
Roles and Responsibilities

- The various roles in a Six Sigma program are commonly referred to as "Belts."
- In addition to Belts, there are also other key roles with specific responsibilities.
- Let us explore the different roles and their corresponding responsibilities in a Six Sigma program.



Roles and Responsibilities

- Each of the four Six Sigma belts represents a different level of expertise in the field of Six Sigma.
 - Six Sigma Master Black Belt (MBB)
 - Six Sigma Black Belt (BB)
 - Six Sigma Green Belt (GB)
 - Six Sigma Yellow Belt (YB)



- In addition to Belts, there are other critical and complementary roles:
 - Champions
 - Sponsors
 - Stakeholders
 - Subject Matter Experts (SMEs).



Roles and Responsibilities: MBB

- The **Master Black Belt** (MBB) is the most experienced, educated, and capable Six Sigma expert.
- A typical MBB has managed dozens of Black Belt level projects.
- The MBB can simultaneously lead multiple Six Sigma Belt projects while mentoring and certifying Black Belt and Green Belt candidates.
- The MBB typically works with high-level operations directors, senior executives, and business managers to help with assessing and planning business strategies and tactics.



Roles and Responsibilities: MBB

 MBB commonly advises management team on the cost of poor quality of an operation and consults on methods to improve business performance.

Typical Responsibilities of a MBB

- Identifies and defines the portfolio of projects required to support a business strategy
- Establishes scope, goals, timelines, and milestones
- Assigns and marshals resources
- Trains and mentors Green Belts and Black Belts
- Facilitates tollgates or checkpoints for Belt candidates
- Reports-out/updates stakeholders and executives
- Establishes organization's Six Sigma strategy/roadmap
- Leads the implementation of Six Sigma.



Roles and Responsibilities: BB

- The **Black Belt** (BB) is the most active and valuable experienced Six Sigma professional among all the Six Sigma Belts.
- A typical BB has
 - led multiple projects
 - trained and mentored various Green Belts candidates
 - understood how to define a problem and drive effective solution.
- The BB is well rounded in terms of project management, statistical analysis, financial analysis, meeting facilitation, prioritization, and a range of other value-added capabilities, which makes a BB highly valuable asset in the business world.



Roles and Responsibilities: BB

 BBs commonly serve as the dedicated resource continuing their line management role while simultaneously achieving a BB certification.

Typical Responsibilities of a BB

- Project Management
 - Defines projects, scope, teams etc.
 - Marshals resources
 - Establishes goals, timelines, and milestones
 - Provides reports and/or updates to stakeholders and executives.



Roles and Responsibilities: BB

- Typical Responsibilities of a BB (continued)
 - Task Management
 - Establishes the team's Lean Sigma roadmap
 - Plans and implements the use of Lean Sigma tools
 - Facilitates project meetings
 - Does project management of the team's work
 - Manages progress toward objectives.
 - Team Management
 - Chooses or recommend team members
 - Defines ground rules for the project team
 - Coaches, mentors, and directs project team
 - Coaches other Six Sigma Belts
 - Manages the team's organizational interfaces.



Roles and Responsibilities: GB

- The **Green Belt** (GB) is considered as a less intense version of Six Sigma professional than the Black Belt (BB).
- A GB is exposed to all the comprehensive aspects of Six Sigma with less focus on the statistical theories and some other advanced analytical methodologies such as Design of Experiment (DOE).
- When it comes to project management, a GB has almost the same responsibilities as a BB.
- In general, the GB works on less complicated and challenging business problems than a BB.



Roles and Responsibilities: GB

Typical Responsibilities of a Green Belt

- Project Management
 - Defines the project, scope, team etc.
 - Marshals resources
 - Sets goals, timelines, and milestones
 - Reports-out/updates stakeholders and executives.
- Task Management
 - Establishes the team's Lean Sigma Roadmap
 - Plans and implements the use of Lean Sigma tools
 - Facilitates project meetings
 - Does Project Management of the team's work
 - Manages progress toward objectives.
- Team Management
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 - Defines ground rules for the project team
 - Coaches, mentors, and directs project team
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 - Manages the team's organizational interfaces.



Roles and Responsibilities: YB

- The Yellow Belt (YB) understands the basic objectives and methods of a Six Sigma project.
- YB has an elementary understanding about what other Six Sigma Belts (GB, BB, MBB) are doing to help them succeed.
- In a Six Sigma project, YB usually serves as a subject matter expert regarding some aspects of the process or project.
- Supervisors, managers, directors, and sometimes executives are usually trained at the YB level.



Roles and Responsibilities: YB

- Typical Responsibilities of a Yellow Belt
 - Helps define process scope and parameters
 - Contributes to team selection process
 - Assists in information and data collection
 - Participates in experiential analysis sessions (FMEA, Process Mapping, Cause and Effect etc.)
 - Assists in assessing and developing solutions
 - Delivers solution implementations.



Roles and Responsibilities: Champions & Sponsors

- Champions and sponsors are those individuals (directors, executives, managers etc.) chartering, funding, or driving the Six Sigma projects that BBs and GBs are conducting.
- Champions and sponsors need to have a basic understanding of the concepts, tools, and techniques involved in the DMAIC methodology so that they can provide proper support and direction.



Roles and Responsibilities: Champions & Sponsors

- Champions and sponsors play critical roles in the successful deployment of Six Sigma.
- Strong endorsement of Six Sigma from the leadership team is critical for success.

Typical Responsibilities of a Champion/Sponsor

- Maintains a strategic oversight
- Establishes strategy and direction for a portfolio of projects
- Clearly defines success
- Provides resolution for issues such as resources or politics
- Establishes routine tollgates or project reviews
- Clears the path for solution implementation
- Assists in project team formation.



Roles and Responsibilities: Stakeholders

- Stakeholders are usually the recipients or beneficiaries of the success of a Six Sigma project.
- Stakeholders are individuals owning the process, function, or production/service line that a Six Sigma Belt focuses on improving the performance of.
- BBs and GBs need to keep strong working relationships with stakeholders because without their support, it would be extremely difficult to make the Six Sigma project a success.



Roles and Responsibilities: SMEs

- Subject Matter Experts (SMEs) are commonly known as the experts of the process or subject matter.
- Six Sigma Belts should proactively look to key SMEs to round out their working project team.
- SMEs play critical roles to the success of a project.
 - Based on SMEs' extensive knowledge about the process, they have the experience to identify which solutions can work and which cannot work.
 - SMEs who simply do not speak up can hurt the chances of the process' success.
 - SMEs are also the same people who prefer to keep the status quo. Six Sigma Belts may find many of them unwilling to help implement the changes.



Roles and Responsibilities

- Throughout this module we have reviewed the various common roles and corresponding responsibilities in any Six Sigma program:
 - Six Sigma Master Black Belt
 - Six Sigma Black Belt
 - Six Sigma Green Belt
 - Six Sigma Yellow Belt
 - Champion and Sponsors
 - Stakeholders
 - Subject Matter Experts (SMEs)
- These Six Sigma belts and other roles are designed to deliver value to the business effectively and successfully.



1.2. Six Sigma Fundamentals



Black Belt Training: Define Phase

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1.2.1 Defining a Process



Defining a Process

- The basic method of defining and understanding a process is the process map.
- Process maps help determine where and how a process begins as well as all the steps and decisions in between.
- By learning the various types and methods of process maps, you can become adept at setting project scopes, identifying value-added and non-value-added steps, identifying problems in a process, etc.
- This module covers:
 - High-level process maps
 - Detailed process maps
 - Functional maps.
- In the Measure section we will touch on several other types and methods of process mapping.



What is a Process Map?

- A process map is a graphical representation of a process flow.
- It illustrates how the business process is accomplished step by step.
- It describes how the materials or information sequentially flow from one business entity to the next.
- It illustrates who is responsible for what between the process boundaries.
- It depicts the inputs and outputs of each individual process step.
- Always encourage your project team to map the current state of the process instead of the ideal state. Be honest with each other!



Process Map Basic Symbols

 The following four symbols are the most commonly used symbols in a process map.



Terminator (Oval):

Shows the start and end points in the process.



Process (Rectangle):

Indicates a single process step.



Decision (Diamond):

Indicates a question with two choices (e.g. Yes/No)



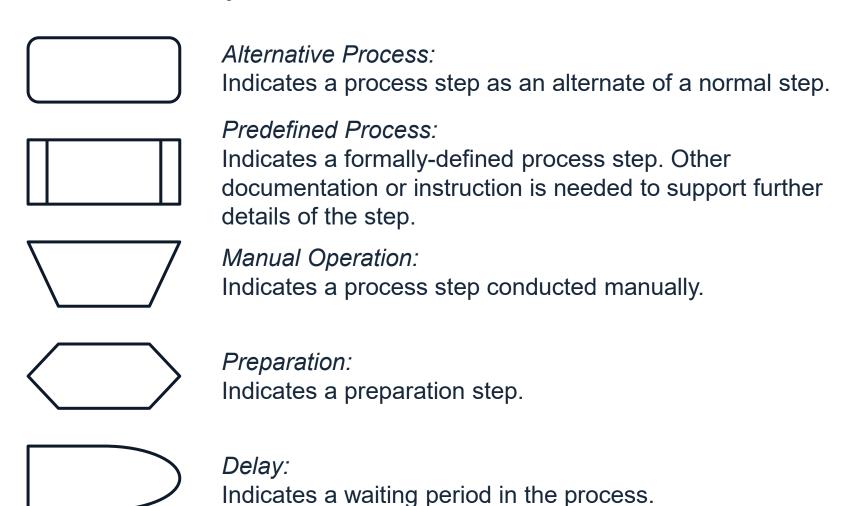
Flow Line (Arrow):

Shows the direction of the process flow.



Additional Process Symbols

Additional Process Symbols:



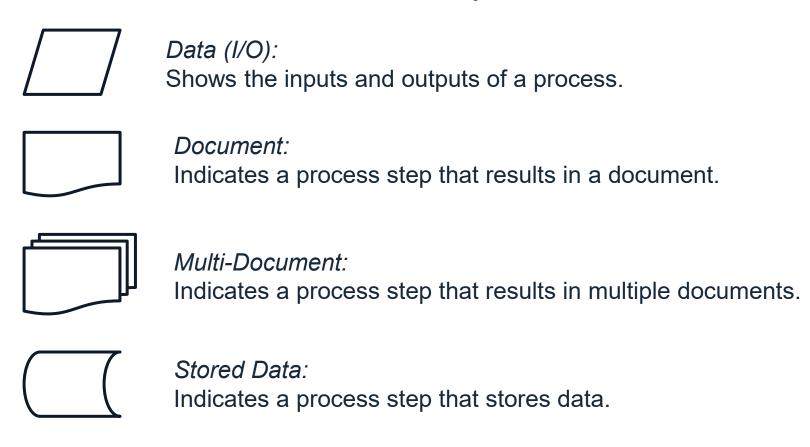


Additional Process Symbols

Additional File and Information Related Symbols:

Magnetic Disk:

Indicates a database.





Additional Process Symbols

Additional Control of Flow Symbols:



Off-Page Connector:

Indicates the process flow continues onto another page.



Merge:

Indicates multiple processes merge into one.



Extract:

Indicates a process splits into multiple parallel processes.



Or:

Indicates a single data processing flow diverges to multiple branches with different criteria requirements.



Summing Junction:

Indicates multiple data processing flows converge into one.



- Step 1: Define the boundaries of the process you want to map.
 - A process map can depict the flow of an entire process or a segment of it.
 - You need to identify and define the beginning and ending points of the process before starting to plot.
 - Use operational definitions where possible.



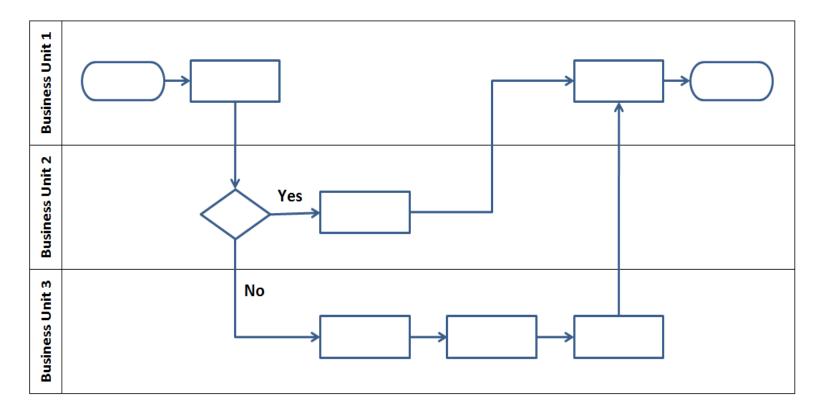
- Step 2: Define and sort the process steps with the flow.
 - Consult with process owners and SMEs or observe the process in action to understand how the process is actually performed.
 - Record the process steps and sort them according to the order of their occurrence.



- Step 3: Fill the step information into the appropriate process symbols and plot the diagram.
 - In the team meeting of process mapping, place the sticky notes with different colors on a white board so you can move them around while the map is under-construction.
 - The flow lines can be plotted directly on the white board.
 - Decision steps. Rotate the sticky note 45 degrees.
 - When the map is completed on the white board, record the map using Excel, PowerPoint, Visio, Quality Companion, or other preferred software.



- Step 3:
 - To illustrate the responsibility of different organizations involved in the process, use a Swim Lane Process Map.





- Step 4: Identify and record inputs/outputs and their corresponding specifications for each process step.
 - The process map helps in understanding and documenting Y = f(x) of a process, where Y represents the outputs and x represents the inputs.
 - The inputs of each process step can be controllable or non-controllable, standardized operational procedures, or noise.
 - Inputs are the source of variation in the process and need to be analyzed qualitatively and quantitatively.
 - The outputs of each process step can be products, information, services, etc. They are the little Y's within the process.

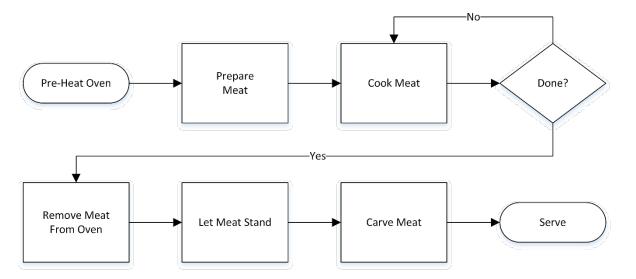


- Step 5: Evaluate the process map and adjust if needed.
 - If the process is too complicated to be covered in one single process map, you may create additional detailed sub-process maps for further information.
 - Number the process steps in the order of their occurrence for clarity.



High Level Process Map

- Most high-level process maps are also referred to as flow charts.
- The key to a high-level process map is to over-simplify the process being depicted so that it can be understood in its most generic form.
- As a general rule, high-level process maps should be no more than 4— 6 steps.
- Below is an oversimplified version of a high-level process map for cooking a 10lb prime rib for a dozen holiday guests.





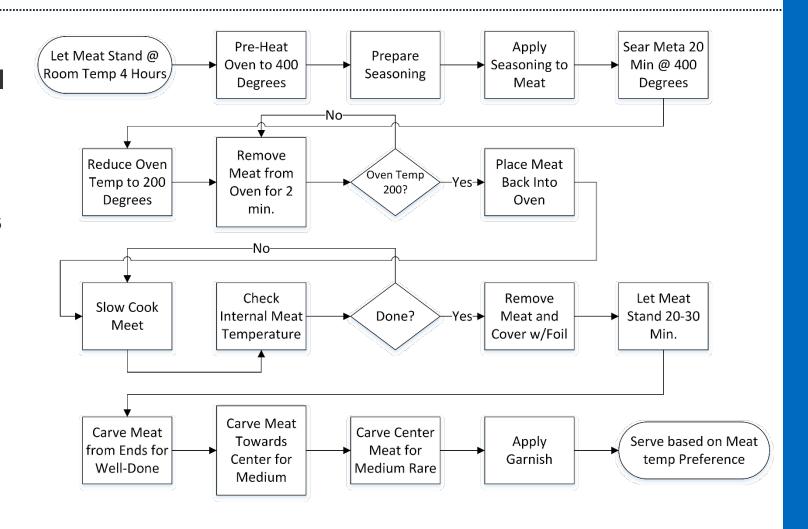
Detailed Process Map

- Detailed process maps or multi-level maps take the high-level map much further.
- Detailed maps can be 2–4 levels deeper than your high-level process map.
- A good guideline used to help create the second level is to take each step in the high level map and break it down into 2–4 steps (no more).
- Repeat this process (level 3, level 4 etc.) until reaching the desired level of detail.
- Some detailed maps are 2 or 3 levels deep, others can be 5–6 levels deep. Obviously, the deeper the levels, the more complex and the more burdensome.



Detailed Process Map

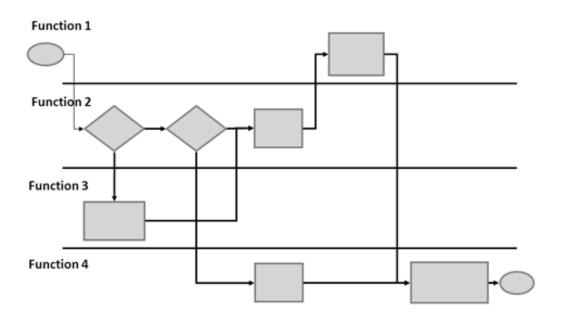
- At right is our prime rib cooking example at level
 2 detail.
- This process map has a few more decision points and process steps.
- You can see that going only one more level deep adds a fair amount of information to the process map.





Functional Process Map

- The functional map adds dimension to the high-level or detailed map.
- The dimension added is identifying which function or job performs the step or makes the decision.
- Below is a generic example of a functional map. Note that functions are identified in horizontal lanes and each process step is placed in the appropriate lane based on which function performs the step.





1.2.2 VOC and CTQs



Voice of the Customer

- VOC stands for "Voice of the Customer."
- Voice of the customer is a term used for a data-driven plan to discover customer wants and needs.
- VOC is an important component to a successful Six Sigma project.
- There are also other "Voices" that need to be heard when conducting projects. The 3 primary forms are:
 - VOC: Voice of the Customer
 - VOB: Voice of the Business
 - VOA: Voice of the Associate.





Gathering VOC

- Gathering VOC should be performed methodically.
- The two most popular methods of collecting VOC are
 - 1. Indirect
 - 2. Direct.
- 1. Indirect data collection for VOC involves passive information exchange:
 - Warranty claims
 - Customer complaints/compliments
 - Service calls
 - Sales reports.
- 2. Indirect methods are less effective, sometimes dated, require heavy interpretation, and are also more difficult to confirm.

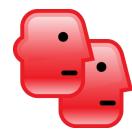


Gathering VOC

- Direct data collection methods for VOC are active and planned customer engagements:
 - Conducting interviews
 - Conducting customer surveys
 - Conducting market research
 - Hosting focus groups.



- Less need to interpret meaning
- Researchers can go a little deeper when interacting with customers
- Customers are aware of their participation and will respond better upon follow-up
- Researchers can properly plan engagements (questions, sample size, information collection techniques etc.).





Gathering VOC

- Gathering VOC requires consideration of many factors such as product or services types, customer segments, manufacturing methods or facilities etc.
- All this information will influence the sampling strategy.
- Consider which factors are important and build a sample size plan around them.
- Also, consider response rates and adjust the initial sample strategy to ensure adequate input is received.
- Once a sampling plan is in place, collect data via the direct and indirect methods discussed earlier.
- After gathering VOC it will be necessary to translate it into something meaningful: CTQs.



Critical to Quality: CTQ

- CTQ stands for Critical to Quality.
- CTQs are translated from VOC or "voice of the customer" feedback.
- VOC is often vague, emotional, or simply a generalization about products or services.



- CTQs are the quantifiable, measureable, and meaningful translations of VOC.
- Organizing VOC helps to identify CTQs.
- One effective way to organize VOC is to group or bucket it using an affinity diagram.
- Affinity diagrams are ideal for large amounts of soft data resulting from brainstorming sessions or surveys.

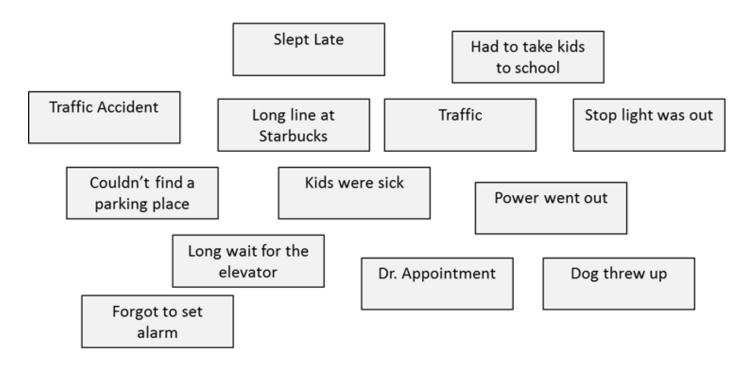


- Steps for conducting an Affinity Diagram exercise:
 - Step 1: Clearly define the question or focus of the exercise ("Why are associates late for work?").
 - Step 2: Record all participant responses on note cards or sticky notes (this is the sloppy part, record everything!).
 - Step 3: Lay out all note cards or post the sticky notes onto a wall.
 - Step 4: Look for and identify common themes.
 - Step 5: Begin moving the note cards or sticky notes into the themes until all responses are allocated.
 - Step 6: Re-evaluate and make adjustments.



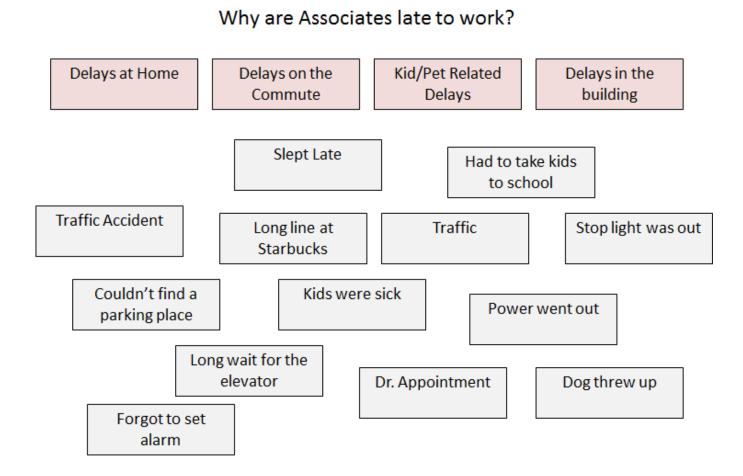
- Define the question or focus
- Record responses on note cards or sticky notes
- Display all note cards or sticky notes on a wall if necessary.

Why are Associates late to work?





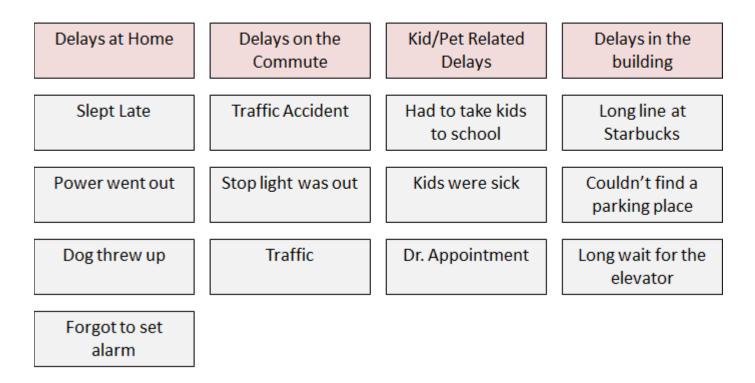
Look for and identify common themes within the responses.





- Group note cards or sticky notes into themes until all responses are allocated.
- Re-evaluate and make final adjustments.

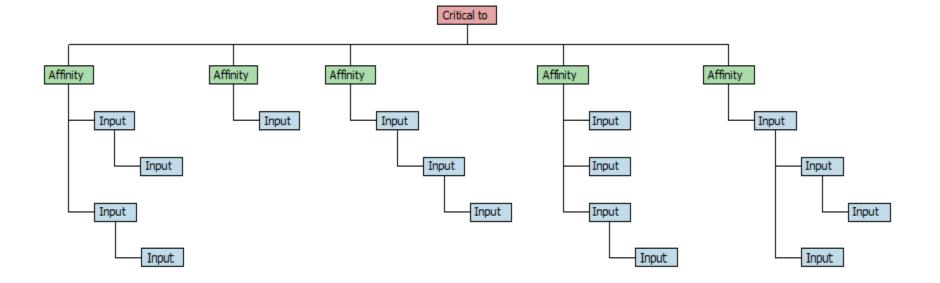
Why are Associates late to work?





CTQ Tree

 Example of a generic CTQ tree transposed from a white board to a software package.





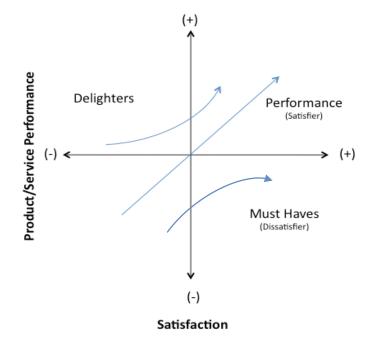
Kano

- Another VOC categorization technique is the Kano.
- The Kano model was developed by Noriaki Kano in the 1980s.

The Kano model is a graphic tool that further categorizes VOC and

CTQs into 3 distinct groups:

- Must Haves
- Performance Attributes
- Delighters.



 The Kano helps to identify CTQs that add incremental value vs. those that are simply requirements and having more is not necessarily better.



Validating VOC and CTQs

- After determining all CTQs, confirm them with the customer.
- Confirming can be accomplished by conducting surveys through one or more of the following methods:

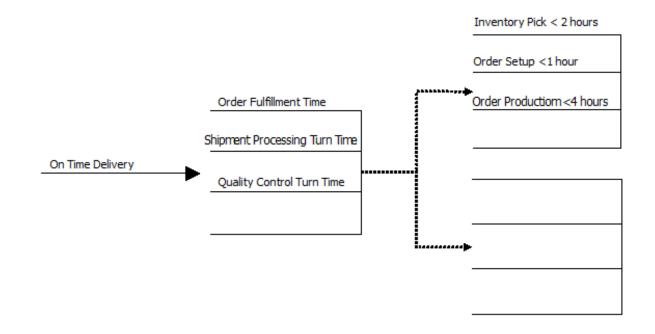


- Group sessions
- One-on-one meetings
- Phone interviews
- Electronic means (chat, email, social media etc.)
- Physical mail.
- Consider your confirming audience and try to avoid factors that may influence or bias responses such as inconvenience or overly burdensome time commitments.



Translating CTQs to Requirements

- Lastly, CTQs must be transformed into **specifics** that can be built upon in a process.
- A requirements tree translates CTQs to meaningful and measureable requirements for production processes and products.





1.2.3 Quality Function Deployment



History of QFD

 Developed by Shigeru Mizuno (1910–1989) and Yoji Akao (b. 1928) in Japan. Quality Function Deployment (QFD) aims to design products that assure customer satisfaction and value – the first time and every time.

 The QFD framework can be used for translating actual customer statements and needs ("The voice of the customer") into actions and designs to build and deliver a quality product.



What is QFD?

- Quality Function Deployment (QFD) is a construction methodology and quantification tool used to identify and measure customer's requirements and transform them into meaningful and measureable parameters.
- QFD helps to prioritize actions to advance process or product to meet customer's anticipations.
- QFD is an excellent tool for contact between cross-functional groups.



Purpose of QFD

The quality function deployment has many purposes. Among the most important are:

- Market analysis to establish needs and expectations
- Examination of competitors' abilities
- Identification of key factors for success
- Translation of key factors into product and process characteristics.



Phases of QFD

Four Key Phases of QFD

- Phase I: Product Planning Including the "House of Quality" (Requirements Engineering Life Cycle)
- Phase II: Product Design (Design Life Cycles)
- Phase III: Process Planning (Implementation Life Cycle)
- Phase IV: Process Control (Testing Life Cycle)

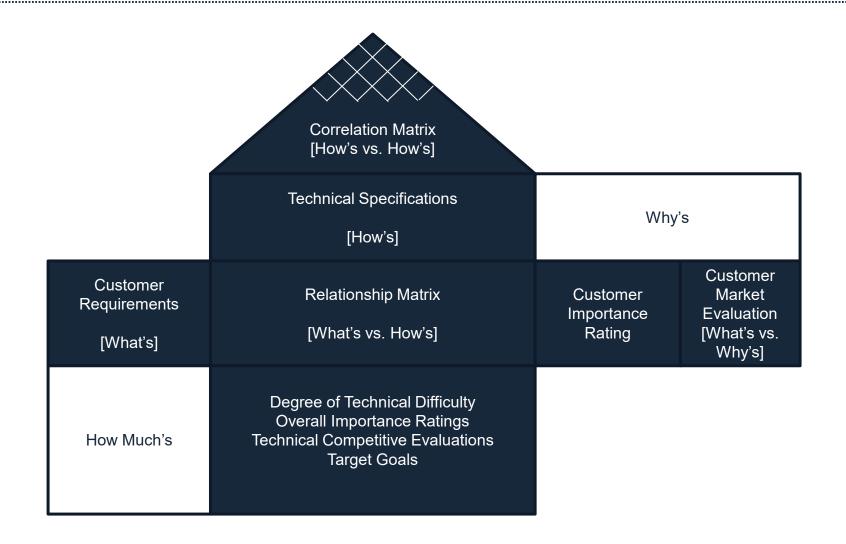


How to build a House of Quality

- Determine Customer Requirements (What's from VOC/CTQ)
- Technical Specifications/Design Requirements (How's)
- Develop Relationship Matrix (What's and How's)
- Prioritize Customer Requirements
- Conduct Competitive Assessments
- Develop Interrelationship (How's)
- Prioritize Design Requirements



House of Quality





Step 1: Determine Customer Requirements

• Identify the important customer requirements. These are the "What's" and are typically determined through the VOC/CTQ process.

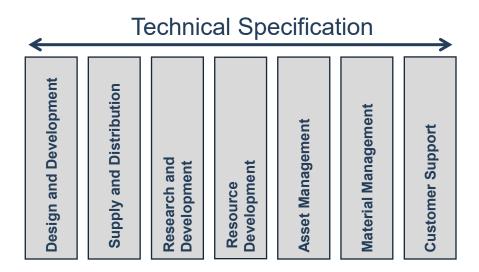
 Use the results from your requirements tree diagram as inputs for the customer requirements in your HOQ.

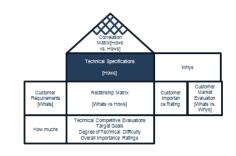




Step 2: Technical Specification

- Potential choices for product features
- Voice of Designers or Engineers
- Each "What" item must be refined to "How's"

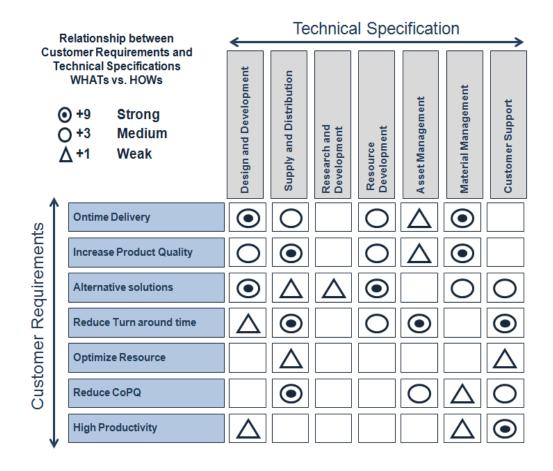


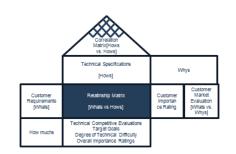




Step 3: Develop Relationship Matrix (What's & How's)

 This is the center portion of the house. Each cell represents how each technical specification relates to each customer requirement.

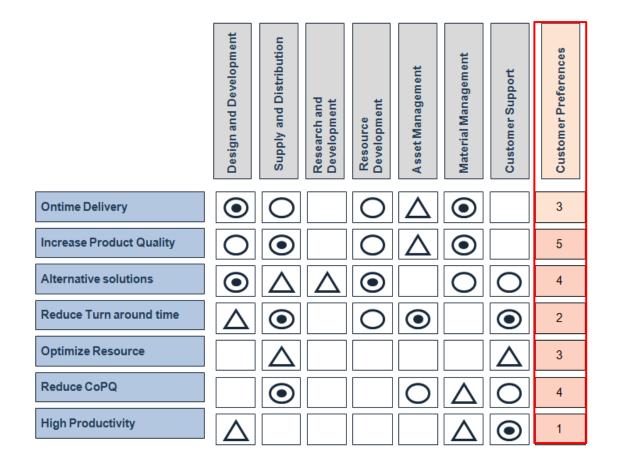


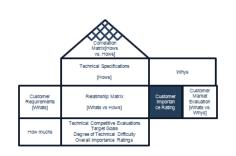




Step 4: Prioritize Customer Requirements

 This is the right portion of the house. Each cell represents customer requirements based on relative importance to customers and perceptions of competitive performance.

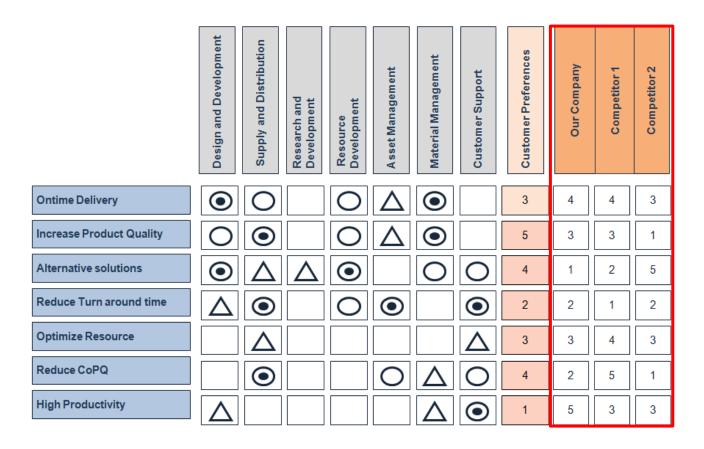


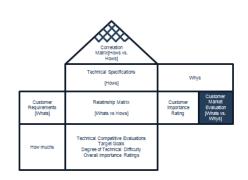




Step 5: Competitive Assessments

• This is the extreme right portion of the house. Comparison of the organization's product to competitors' products.

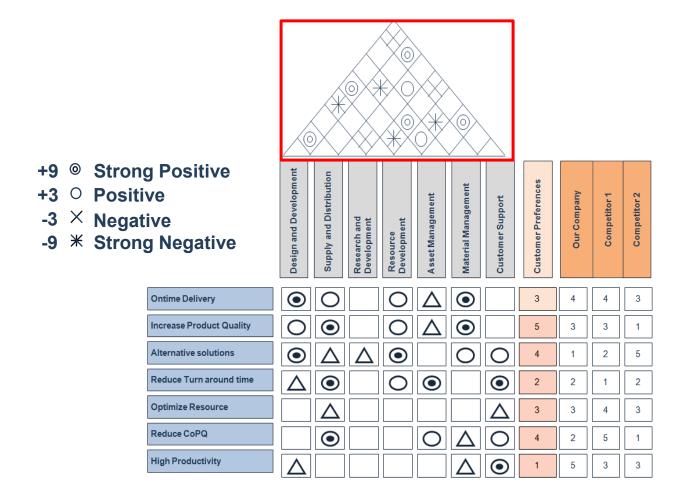


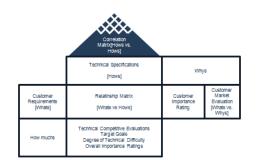




Step 6: Correlation Matrix

• This is the top portion of the house. It identifies the way "how" items either support (positive) or conflict (negative) with one another.



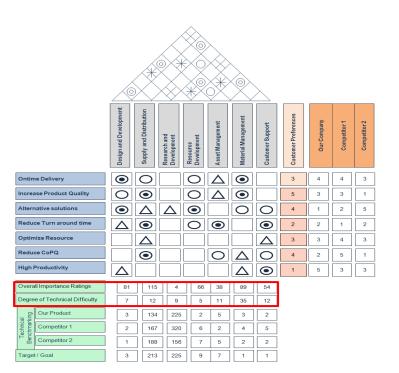




Step 7: Prioritize Design Requirements

- Overall Importance Ratings
- Function of relationship ratings and customer prioritization ratings
- Technical Difficulty Assessment
- Similar to customer market competitive evaluations but conducted by

the technical team



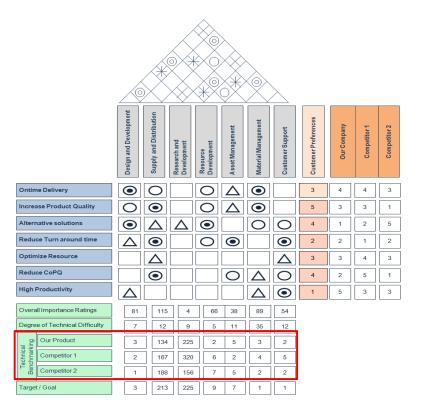


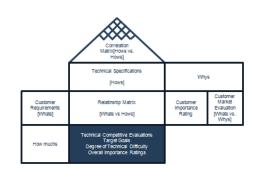
Technical Specifications

Relatinship Matri: [Whats vs Hows]

Step 7: Prioritize Design Requirements

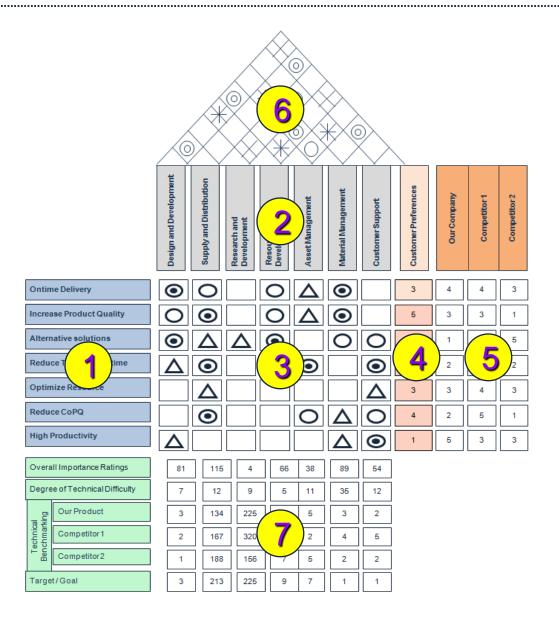
- Technical Specification Competitive Evaluation
- Helps to establish the feasibility and realization of each "how" item
- Target Goals
- How much is good enough to satisfy the customer







House of Quality





Pros of QFD

- Focuses the design of the product or process on satisfying customer's needs and wants.
- Improves the contact channels between customers, advertising, research and improvement, quality and production departments, which sustains better decision making.
- Reduces the new product development project period and cost.



Cons of QFD

- The relationship matrix can be too obscure with many process inputs and/or many customer constraints.
- It can be very complicated and difficult to implement without experience.
- If throughout the process new ideas, specifications, or requirements are not discovered, you run the risk of losing team members' trust in the process.



QFD Summary

When used properly, the quality function deployment is an extremely valuable approach to product/process design.

There are many benefits of QFD that can only be realized when each step of the process is completed thoroughly:

- Logical way of obtaining information and presenting it
- Smallest product development cycle
- Considerably condensed start-up costs
- Fewer engineering alterations
- Reduced chance of supervision during design process
- Collaborating environment
- Preserving everything in characters.



1.2.4 Cost of Poor Quality



Cost of Poor Quality

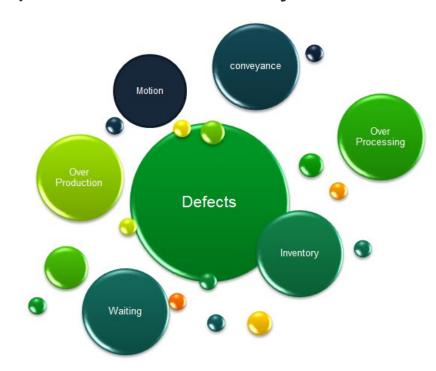


- Cost of Poor Quality (COPQ) is the expense incurred due to waste, inefficiencies, and defects.
- The COPQ has been proven to range from 5% to 30% of gross sales for most companies.
- The COPQ can be staggering when considering process inefficiencies, hidden factories, defective products, rework, scrap, etc.
- Understanding COPQ and where to look for it will help uncover process inefficiencies, defects, and hidden factories within your business.



Cost of Poor Quality

- There are 7 common forms of waste that are often referred to as the "7 deadly muda."
- Technically, there are more than 7 forms of waste but if you can remember these you will capture over 90% of your waste.
 - 1. Defects
 - 2. Overproduction
 - 3. Over-Processing
 - 4. Inventory
 - 5. Motion
 - 6. Transportation
 - 7. Waiting





Cost of Poor Quality

- The "7 deadly muda" are very important to understand. They are the best way to identify the COPQ.
- The presence of any muda causes many other forms of inefficiencies and hidden factories to manifest themselves.
- There are four key categories of costs related to muda:
 - 1. Costs Related to Production
 - 2. Costs Related to Prevention
 - 3. Costs Related to Detection
 - 4. Costs Related to Obligation





COPQ: Costs Related to Production

- Costs related to production are the direct costs of the presence of muda. These forms of COPQ are usually understood and easily observable. They are in fact the "7 deadly muda" themselves.
 - 1. Defects
 - 2. Overproduction
 - 3. Over-Processing
 - 4. Inventory
 - 5. Motion
 - 6. Transportation
 - 7. Waiting



COPQ: Costs Related to Prevention

- Costs related to the prevention of muda are those associated with trying to reduce or eliminate any of the "7 deadly muda."
 - Costs for error proofing methods or devices
 - Costs for process improvement and quality programs
 - Costs for training and certifications
 - etc.

 Any costs directly associated with the prevention of waste and defects should be included in the COPQ calculation.



COPQ: Costs Related to Detection

- Costs related to the detection of muda are those associated with trying to find or observe any of the "7 deadly muda."
 - Costs for sampling
 - Costs for quality control check points
 - Costs for inspection costs
 - Costs for cycle counts or inventory accuracy inspections
 - etc.
- Any costs directly associated with the detection of waste and defects should be included in the COPQ calculation.



COPQ: Costs Related to Obligation

- Costs related to obligation are those associated with addressing the muda that reaches a customer.
 - Repair costs
 - Warranty costs
 - Replacement costs
 - Customer returns and customer service overhead
 - etc.

 Any costs directly associated with customer obligations should be included in the COPQ calculation.



COPQ: Types of Cost

There are two types of costs to be considered when determining COPQ

1. Hard Costs

Tangible costs that can be traced to the income statement

2. Soft Costs

Intangible costs: avoidance, opportunity costs, lost revenue etc.

Calculating the COPQ

- 1. Determine the types of waste that are present in your process
- 2. Estimate the frequency of waste that occurs
- 3. Estimate the cost per event, item, or time frame
- 4. Do the math.



1.2.5 Pareto Charts and Analysis



Pareto Principle

- The Pareto principle is commonly known as the "law of the vital few" or "80:20 rule."
- It means that the majority (approximately 80%) of effects come from a few (approximately 20%) of the causes.
- This principle was first introduced in early 1900s and has been applied as a rule of thumb in various areas.
- Example of applying the Pareto principle:
 - 80% of the defects of a process come from 20% of the causes.
 - 80% of sales come from 20% of customers.



Pareto Principle

- The Pareto principle helps us to focus on the vital few items that have the most significant impact.
- In concept, it also helps us to prioritize potential improvement efforts.
- Since this 80:20 rule was originally based upon the works of Wilfried Fritz Pareto (or Vilfredo Pareto), the Pareto principle and references to it should be capitalized because Pareto refers to a person (proper noun).
 - Mr. Pareto is also credited for many works associated with the 80:20, some more loosely than others:
 - Pareto's Law
 - Pareto efficiency
 - Pareto distribution etc.



Pareto Charts

- A Pareto chart is a chart of descending bars with an ascending cumulative line on the top.
 - Sum or Count:

The descending bars on a Pareto chart may be set on a scale that represents the total of all bars or relative to the biggest bucket, depending on the software you are using.

- **Percent to Total**: A Pareto chart shows the percentage to the total for individual bars.
- Cumulative Percentage: A Pareto chart also shows the cumulative percentage of each additional bar. The data points of all cumulative percentages are connected into an ascending line on the top of all bars.



Pareto Charts

- Case study time!
 - Next we will use Minitab to run Pareto charts on exactly the same data set.
 - Open the Minitab data file labeled Pareto and follow the instructions over the next few pages to run Pareto charts in Minitab.

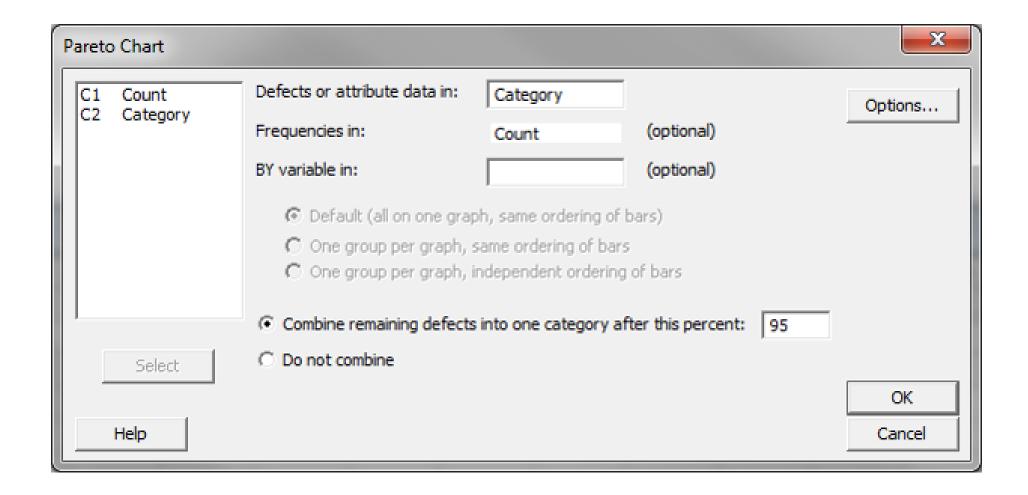


Create a Pareto Chart in Minitab

- Steps to generate a Pareto chart using Minitab:
 - 1. Open the spreadsheet with the count data for individual categories.
 - 2. Click on Stat → Quality Tools → Pareto Chart.
 - 3. A new window with the title "Pareto Chart" pops up.
 - 4. Select "Category" into the box "Defects or attribute data in" and "Count" into the box "Frequency in."
 - 5. Click "OK."
 - 6. The Pareto chart is created in a newly-generated window.

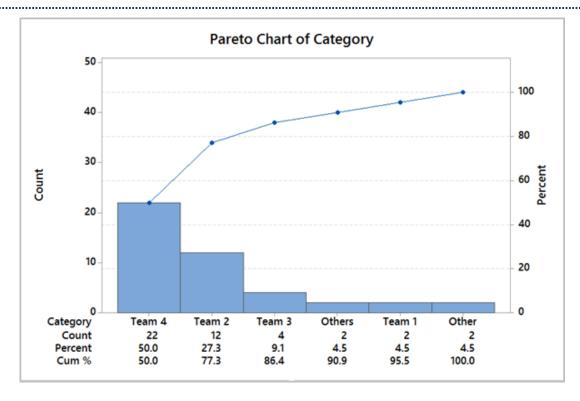


Create a Pareto Chart in Minitab





Create a Pareto Chart in Minitab



- The above Pareto chart generated in Minitab presents the count of defective products by team.
- The bars are descending on a scale with the peak at 50, which is approximately the total count of all defective products for all teams.
- The table below the chart contains counts, individual percentages, and cumulative percentages.
- The cumulative percentages are the red data points driving the red line that spans across the graphic.

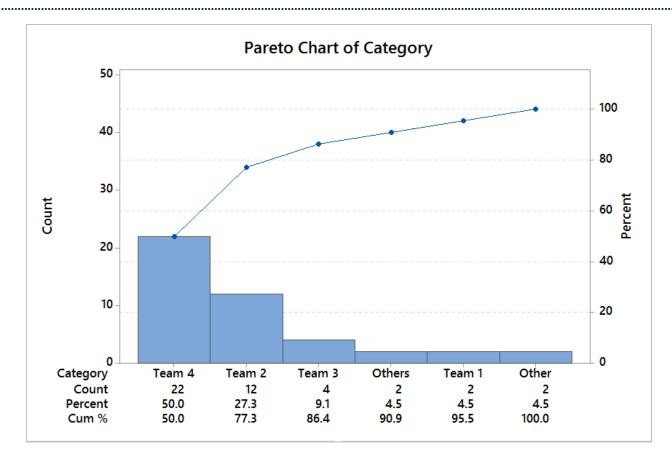


Pareto Analysis

- The Pareto analysis is used to identify the root causes by using multiple Pareto charts.
- In Pareto analysis, we drill down into the bigger buckets of defects and identify the root causes of defects that contribute heavily to total defects.
- This "drill down" approach effectively solves a significant portion of the problem.
- Next you will see an example of three-level Pareto analysis.
 - The second-level Pareto is a Pareto chart that is a subset of the tallest bar on the first Pareto.
 - The third-level Pareto is a subset of the tallest bar of the second-level Pareto.



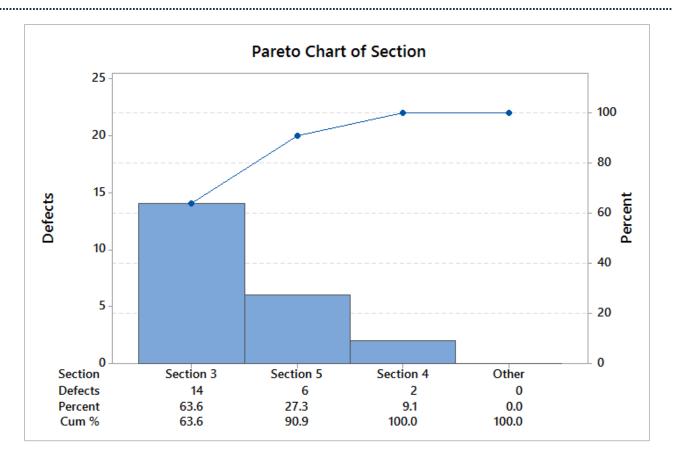
Pareto Analysis: First Level



- First-level Pareto
- Shows the count of defective items by team
- Next level will only show the defective items of team 4



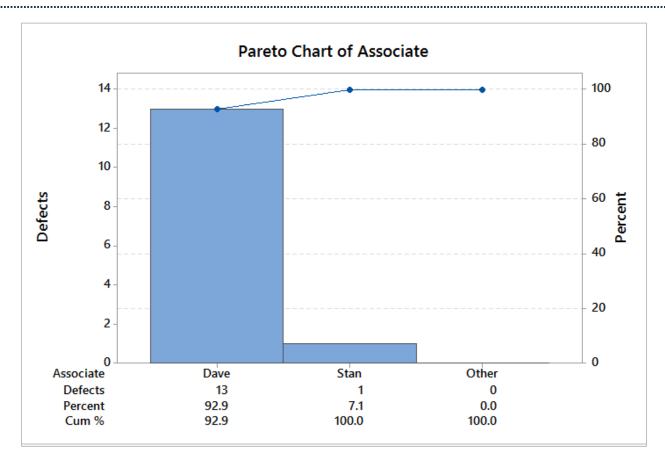
Pareto Analysis: Second Level



- Second-level Pareto
- Shows the count of the defective items by section for only team 4
- Next level will only show the defective items of section 3



Pareto Analysis: Third Level

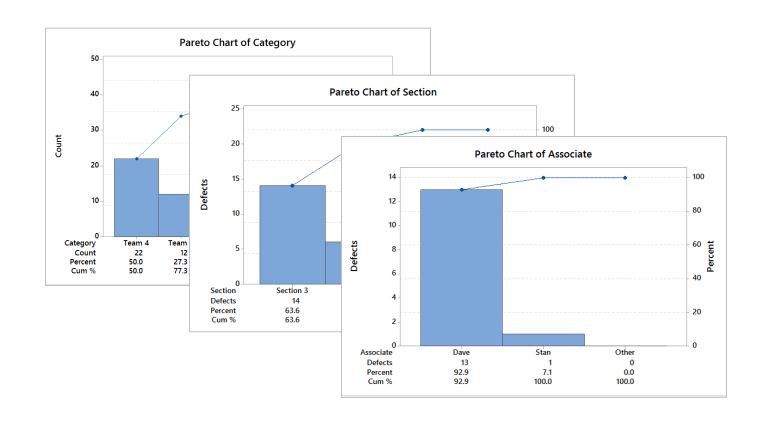


- Third-level Pareto
- Shows the count of defective items by associate for only section 3 of team 4
- Next level will only show the defective items of Dave



Pareto Analysis: Conclusion

- After drilling down three levels we find that most of the defective products are from Dave who is in Section 3 of Team 4.
- Determining what Dave might be doing differently and solving that problem can potentially fix about 30% of the entire defective products (13/44).





1.3 Six Sigma Projects



Black Belt Training: Define Phase

1.1 Six Sigma Overview

- 1.1.1 What is Six Sigma
- 1.1.2 Six Sigma History
- 1.1.3 Six Sigma Approach Y = f(x)
- 1.1.4 Six Sigma Methodology
- 1.1.5 Roles and Responsibilities

1.2 Six Sigma Fundamentals

- 1.2.1 Defining a Process
- 1.2.2 VOC and CTQs
- 1.2.3 QFD
- 1.2.4 Cost of Poor Quality (COPQ)
- 1.2.5 Pareto Analysis (80:20 rule)

1.3 Lean Six Sigma Projects

- 1.3.1 Six Sigma Metrics
- 1.3.2 Business Case and Charter
- 1.3.3 Project Team Selection
- 1.3.4 Project Risk Management
- 1.3.5 Project Planning

1.4 Lean Fundamentals

- 1.4.1 Lean and Six Sigma
- 1.4.2 History of Lean
- 1.4.3 The Seven Deadly Muda
- 1.4.4 Five-S (5S)



1.3.1 Six Sigma Metrics



Six Sigma Metrics

- There are many Six Sigma metrics and/or measures of performance used by Six Sigma practitioners.
- In addition to the ones we will cover here, several others (Sigma level, Cp, Cpk, Pp, Ppk, takt time, cycle time, utilization etc.) will be covered in other modules throughout this training.
- The Six Sigma metrics of interest here in the define phase are:
 - Defects per Unit (DPU)
 - Defects per Million Opportunities (DPMO)
 - Yield (Y)
 - Rolled Throughput Yield (RTY).



Defects per Unit: DPU

- DPU stands for "Defects per Unit"
- DPU is the basis for calculating DPMO and RTY, which we will cover in the next few pages.
- DPU is found by dividing total defects by total units.
 - DPU = D/U
- For example, if you have a process step that produces an average of 65 defects for every 598 units, then your DPU = 65/598 = 0.109.



- **DPMO** is one of the few important Six Sigma metrics that you should get comfortable with if you are associated with Six Sigma.
- In order to understand DPMO it is best if you first understand both the nomenclature and the nuances such as the difference between defect and defective.
- Nomenclature
 - Defects = D
 - Unit = U
 - Opportunity to have a defect = O



 In order to properly discuss DPMO, we must first explore the differences between "defects" and "defective."

Defective

- Defective suggests that the value or function of the entire unit or product has been compromised.
- Defective items will always have at least one defect. Typically, however, it takes
 multiple defects and/or critical defects to cause an item to be defective.

Defect

- A defect is an error, mistake, flaw, fault, or some type of imperfection that reduces the value of a product or unit.
- A single defect may or may not render the product or unit "defective" depending on the specifications of the customer.

Summary

- Defect means that part of a unit is bad.
- Defective means that the whole unit is bad.



 Now let us turn our attention to defining "opportunities" so that we can fully understand Defects per Million Opportunities (DPMO).

Opportunities

- Opportunities are the total number of possible defects.
- Therefore, if a unit has 6 possible defects, then each unit produced is equal to 6 defect opportunities.
- If we produce 100 units, then there are 600 defect opportunities.



- Calculating Defects per Million Opportunities
- The equation is DPMO = $(D/(U \times O)) \times 1,000,000$
- Example: Let us assume:
 - There are 6 defect opportunities per unit
 - There are an average of 4 defects every 100 units.

- Opportunities = $6 \times 100 = 600$
- Defect rate = 4/600
- DPMO = $4/600 \times 1,000,000 = 6,667$



- What is the reason or significance of 1,000,000?
- Converting defect rates to a per million value becomes necessary when the performance of your process approaches Six Sigma.
- When this happens, the number of defects shrinks to virtually nothing.
 In fact, if you recall from the "What is Six Sigma" module, sigma is 3.4 defects per million opportunities.
- By using 1,000,000 opportunities as the barometer we have the resolution in the measurement to count defects all the way up to Six Sigma.



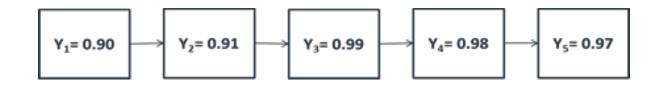
RTY: Rolled Throughput Yield

- Rolled Throughput Yield (RTY) is a process performance measure that provides insight into the cumulative effects of an entire process.
- RTY measures the yield for each of several process steps and provides the probability that a unit will come through that process defect-free.
- RTY allows us to expose the "hidden factory" by providing visibility into the yield of each process step.
- This helps us identify the poorest performing process steps and gives us clues into where to look to find the most impactful process improvement opportunities.



RTY: Rolled Throughput Yield

- Calculating RTY:
- RTY is found by multiplying the yields of each process step.
- Let us take the 5-step process below and calculate the RTY using the multiplication method mentioned above.



- The calculation is: RTY = 0.90 × 0.91 × 0.99 × 0.98 × 0.97 = 0.77
- Therefore, RTY = 77%.



RTY: Rolled Throughput Yield

- You may have noticed that in order to calculate RTY we must determine the yield for each process step.
- Before we get into calculating yield, there are a few abbreviations that need to be declared.
 - Abbreviations
 - Defects = **D**
 - Unit = **U**
 - Defects per Unit = DPU
 - Yield = **Y**
 - e = 2.71828 (mathematical constant)



RTY: Rolled Throughput Yield

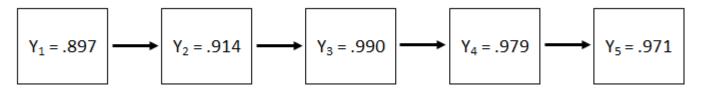
- Calculating Yield
- The Yield of a process step is the success rate of that step or the probability that the process step produces no defects.
- In order to calculate the yield, we need to know the DPU and then we can apply it to the yield equation below.

$$Yield = e^{-dpu}$$

- Example
 - Let us assume a process step has a DPU of 0.109 (65/598)

- Yield = 2.718 ^ -0.109 = 0.8967. Rounded, Yield = 90%.

RTY: Rolled Throughput Yield



- Below is a table using the above process yield data that we used in the earlier RTY calculation.
- This table allows us to see the DPU and yield of each step as well as the RTY for the whole process.

Process Step	Defects	Units	DPU	Yield	RTY
1	65	598	0.109	0.897	90%
2	48	533	0.090	0.914	82%
3	5	485	0.010	0.990	81%
4	10	480	0.021	0.979	79%
5	14	471	0.030	0.971	77%



RTY: Using an Estimate of Yield

- Calculating RTY using Yield estimation
 - It is possible to "estimate" yield by taking the inverse of DPU or simply subtracting DPU from 1.
 - Yield Estimation = 1 DPU
 - Example Yield Estimate for process step 1: 1 − 0.10870 = 0.90
 - RTY using the Yield Estimation Method
 - RTY = $0.891 \times 0.91 \times 0.99 \times 0.979 \times 0.97 = 0.76 = 76\%$

Process Step	Defects	Units	DPU	Yield _{estimate}	RTY _{estimate}
1	65	598	0.109	0.891	89%
2	48	533	0.090	0.910	81%
3	5	485	0.010	0.990	80%
4	10	480	0.021	0.979	79%
5	14	471	0.030	0.970	76%

NOTE: This method assumes max DPU is 1. Although this is not technically true, in practice its extremely rare to find processes with DPU greater than 1. If it is the case this method will not work.



RTY: Using First Time Yield (FTY)

- In the previous discussion we calculated RTY using Yields calculated based on "Defects" per unit. What if we're counting "Defectives"?
 - First Time Yield (FTY) is the probability of producing a product or service defect free before rework
 - $FTY = \frac{\# Non-Defective\ Units\ Produced\ Before\ Rework}{\#\ Units\ Inspected}$
- FTY is also a good measure of process performance. It accounts for the hidden factory or rework in the process. Its simply based on defective units instead of defects per unit.
 - You can still calculate RTY using First Time Yield:
 - RTY= FTY₁ * FTY₂ * FTY₃ * FTY₄ * FTY₅



1.3.2 Business Case and Charter



Business Case and Project Charter

- Earlier we stated that DMAIC is a structured and rigorous methodology designed to be repeatedly applied to <u>any</u> process in order to achieve Six Sigma.
- We also stated that DMAIC was a methodology that refers to 5 phases of a project.
 - Define, Measure, Analyze, Improve, and Control
- Given that the premise of the DMAIC methodology is project-based, we must take the necessary steps to define and initiate a project, hence the need for. . .
 - Project Charters



Project Charter

- The purpose of a project charter is to provide vital information about a project in a quick and easy-to-comprehend manner.
- Project charters are used to get approval and buy-in for projects and initiatives as well as declaring:
 - The scope of work
 - Project teams
 - Decision authorities
 - Project lead
 - Success measures



Project Charter Template

				Project Title:				
				Black Belt	Project Champion	Executive	Sponsor	MBB/Mentor
Primary M	letric				Secondary Metric			
Problem S	tatement				Business Case			
High Leve	l Project T	imeline	Constraint	ts & Dependencies	Project Risks		Other Dia	gnostics
Phase	Start	Finish		·				
Define								
Measure]					
Analyze								
Improve								
Control	· · ·					в т	O CMEI	
Approval/Steering Committee Stakeholders & Advis								
Name		Organizat	ion	Name	Organization	Name		Organization
I								



Project title, leader and sponsor are critic					
components. Identifying a lead helps kee order and accountability. Without a spons	· IIII a cla II a lt	Project Champion	Executive Sponsor	MBB/Mentor	
and approval there is no project					
rimary Metric		Secondary Metric			
The primary metric is vital to u you are measuring project so being able to establish a ba	uccess as well as	Sometimes you'll	tric is a measure that help find improvement of a me econdary metrics helps br	tric to the detriment of	
oblem Statement		Business Case			
The problem statement is a conc declaring the projects baseline, go closure. It should also be exp	oal, gap to goal and gap olicit about timing.	project is	ess case describes s important to the bu	isiness.	
gh Level Project Timeline Const	raints & Dependencies	Project Risks	Other Di	agnostics	
Be clear about the project timeline.	Identify constraints and project dependencies.	Identify proj risks that the sponsor shou aware of.	he Ild be	entify any other elevant project measures.	
pproval/Steering Committee	Stakeholders & Adv	visors	Project Team & SMI	's	
ame Organization	Name	Organization	Name	Organization	
All formal project require approvals. Identify people who have a vested interest in the project's success. Organization Name Organization Declare project team members and subject matter experts.					



Title

- Projects should have a name, title, or some reference that identifies them.
- Branding can be an important ingredient in the success of a project so be sure your project has a reference name or title.

Leader

- Any projects needs a declared leader or someone who is responsible for project's execution and success.
- You may hear references to RACI throughout in your Six Sigma journey.
- RACI stands for Responsible, Accountable, Consulted, Informed and identifies the people that play those roles.
- Every project must have declared leaders indicating who is responsible and who is accountable.



- Business Case
 - A **business case** is the quantifiable reason why the project is important.
 - Business cases help shed light on problems. They explain why a business should care.

- Business cases must be quantified and stated succinctly.
- COPQ is a key method of quantification for any business case.



- Problem Statement and Objective
 - A properly written problem statement has an objective statement woven into it.
 - There should be no question as to the current state or the goal.
 - A gap should be declared, the gap being the difference between the present state and the goal state.
 - The project objective should be to close the gap or reduce the gap by some reasonable amount.
 - Valuation or COPQ is the monetary value assigned to the gap.
 - Lastly, a well-written problem statement refers to a timeline expected to be met.



Project Charter: Problem Statement Examples

- Currently, process defect rates are 17% with a goal of 2%. This represents a gap of 15%, costing the business \$7.4 million dollars. The goal of this project is to reduce this gap by 50% before Nov 2010 putting process defect rates at 9.5% and saving \$3.7MM.
- Process cycle time has averaged 64 minutes since Q1 2009. However, production requirements put the cycle time goals at 48 min. This 16-min gap is estimated to cost the business \$296,000. The goal of this project is reduce cycle time by 16 min. by Q4 2010 and capture all \$296,000 cost savings.



- Metrics
 - A measure of success is an absolute for any project.
 - Metrics give clarity to the purpose of the work.
 - Metrics establish how the initiative will be judged.
 - Metrics establish a baseline or "starting point."
 - For Six Sigma projects...metrics are mandatory!



- Primary Metric
 - The primary metric is a generic term for a Six Sigma project's most important measure of success. The primary metric is defined by the Black Belt, GB, MBB, or Champion.
 - A primary metric is an absolute MUST for any project and it should not be taken lightly. Here are a few characteristics of good primary metrics.
 - Primary metrics should be:
 - tied to the problem statement
 - measureable
 - expressed with an equation
 - aligned to business objectives
 - tracked at the proper frequency (hourly, daily, weekly, monthly etc.)
 - expressed pictorially over time with a run chart, time series, or control chart
 - validated with an MSA.



- The primary metric is the reason for your work.
- It is the success indicator.
- It is your beacon.
- The primary metric is of utmost importance and should be improved,
 but not at the expense of your secondary metric.



- Secondary Metric
 - The **secondary metric** is the thing you do not want sacrificed on behalf of a primary improvement.
 - A secondary metric is one that makes sure problems are not just "changing forms" or "moving around."
 - The secondary metric keeps us honest and ensures we are not sacrificing too much for our primary metric.
 - If your primary metric is a cost or speed metric, then your **secondary metric** should probably be some quality measure.
 - Example: If you were accountable for saving energy in an office building and your primary metric was energy consumption then you *could* shut off all the lights and the HVAC system and save tons of energy. . .except that your secondary metrics are probably comfort and functionality of the work environment.



- Elements of a Good Project Charters (continued)
 - Scope Statement defined by high-level process map
 - Stakeholders Identified who is affected by the project
 - Approval Authorities Identified who makes the final call
 - Review Committees Defined who is on the review team
 - Risks and Dependencies Highlighted identify risks and critical path items
 - Project Team Declared declare team members
 - Project Timeline Estimated set high-level timeline expectations.



1.3.3 Project Team Selection



 Six Sigma project team selection is the cornerstone of a successful Six Sigma project.

- Teams and Team Success
 - A team is a group of people who share complementary skills and experience.
 - A team will be dedicated to consistent objectives.
 - Winning teams share similar and coordinated goals.
 - Teams often execute common methods or approaches.
 - Team members hold each other accountable for achieving shared goals.



- What makes a team successful?
 - Shared goals
 - Commitment
 - Leadership
 - Respect
 - Effective communication
 - Autonomy
 - Diversity (capabilities, knowledge, skills, experience etc.)
 - Adequate resources.





- Keys to Team Success
 - Agreed focus on the goal or the problem at hand
 - Focus on problems that have meaning to the business
 - Focus on solvable problems within the scope of influence; a successful team does not seek unattainable solutions.
 - Team Selection
 - Selected teammates have proper skills and knowledge
 - Adequately engaged management
 - Appropriate support and guidance from their direct leader
 - Successful teams use reliable methods
 - Follow the prescribed DMAIC methodology
 - · Manage data, information, and statistical evidence
 - Successful teams always have exceeds players
 - Winning teams typically
 - Have unusually high standards.
 - Have greater expectations of themselves and each other.
 - Do not settle for average or even above average results.



- Principles of Team Selection:
 - Select team members based on
 - Skills required to achieve the objective
 - Experience (Subject Matter Expertise)
 - Availability and willingness to participate
 - Team size (usually 4–8 members)
 - Don't go at it alone!
 - Don't get too many cooks in the kitchen!
 - Members' ability to navigate
 - The process
 - The company
 - The political landscape
 - Be sure to consider the inputs of others
 - Heed advice
 - Seek guidance



- All teams experience the following four stages of development. It is helpful to understand these phases so that you can anticipate what your team is going to experience.
- The four stages of team development process:
 - Forming
 - Storming
 - Norming
 - Performing
- Teammates seek something different at each stage:
 - In the forming stage they seek inclusion
 - In the storming stage they seek direction and guidance
 - In the norming stage they seek agreement
 - In the performing stage they seek results.



- Patterns of a team in the Forming stage:
 - Roles and responsibilities are unclear
 - Process and procedures are ignored
 - Scope and parameter setting is loosely attempted
 - Discussions are vague and frustrating
 - There is a high dependence on leadership for guidance
- Patterns of a team in the **Storming** stage:
 - Attempts to skip the research and jump to solutions
 - Impatience for some team members regarding lack of progress
 - Arguments about decisions and actions of the team
 - Team members establish their position
 - Subgroups or small teams form
 - Power struggles exist and resistance is present



- Patterns of a team in the Norming stage:
 - Agreement and consensus start to form
 - Roles and responsibilities are accepted
 - Team members' engagement increases
 - Social relationships begin to form
 - The leader becomes more enabling and shares authority
- Patterns of a team in the **Performing** stage:
 - Team is directionally aware and agrees on objectives
 - Team is autonomous
 - Disagreements are resolved within the team
 - Team forms above average expectations of performance



- Well-structured and energized project teams are the essential components of any successful Six Sigma project.
- To have better chances of executing the project successfully, you will need to understand and effectively manage the team development process.



1.3.4 Project Risk Management



Risk

 Risk is defined as a future event that can impact the task/project if it occurs.





What is Project Risk Management?

- The main purpose of **risk management** is to foresee potential risks that may inhibit the project deliverables from being delivered on time, within budget, and at the appropriate level of quality, and then to mitigate these risks by creating, implementing, and monitoring *contingency plans*.
- Risk management is concerned with identifying, assessing, and monitoring project risks before they develop into issues and impact the project.
- Risk analysis helps to identify and manage potential problems that could impact key business initiatives or project goals.



Three Basic Parameters of Risk Analysis

Risk Assessment:

 The process of identifying and evaluating risks, whether in absolute or relative terms.

Risk Management:

• Project risk management is the effort of responding to risks throughout the life of a project and in the interest of meeting project goals and objectives.

Risk Communication:

 Communication plays a vital role in the risk analysis process because it leads to a good understanding of risk assessment and management decisions.



Why is Risk Analysis Necessary?

What can happen if you omit the risk analysis?

- Vulnerabilities cannot be detected
- Mitigation plans are introduced without proper justification
- Customer dissatisfaction
- Not meeting project goals
- Remake the whole system
- Huge cost and time loss





Project Risk Analysis Steps

The project risk analysis process consists of the following steps that evolve through the life cycle of a project.

- Risk Identification:
 - Identify risks and risk categories, group risks, and define ownership.
- Risk Assessment:
 - Evaluate and estimate the possible impacts and interactions of risks.
- Response Planning:
 - Define mitigation and reaction plans.
- Mitigation Actions:
 - Implement action plans and integrate them into the project.
- Tracking and Reporting:
 - Provide visibility to all risks.
- Closing:
 - Close the identified risk.



Risk Identification

The first action of risk management is the identification of individual events that the project may encounter during its lifecycle.



The identification step comprises:

- Identify the risks
- Categorize the risks
- Match the identified risks to categories
- Define ownership for managing the risks.



Risk Identification

- Source of Risk:
- Identification of risk sources provides a basis for systematically examining changing situations over time to uncover circumstances that impact the ability of the project to meet its objectives.



Risk Identification

Source of Risk	Description				
Human Resources	The risks originated from human resources (e.g., availability, skill etc.)				
Physical Resources	The risks originated from physical resources (e.g., hardware or software, availability of the required number at the right time etc.)				
Technology	The risks originated from technology (e.g., development environment, new or complex technologies, performance requirements, tools etc.)				
Suppliers	The risks are associated with a supplier (e.g., delays in supplies, capability of suppliers etc.)				
Customer	The risks derived from the customer (e.g., unclear requirements, requirement volatility, change in project scope, delays in response etc.)				
Security	The risks are associated with information security, security of personnel, security of assets, and security of intellectual property				
Legal	The risks are associated with legal issues that may impact the project				
Project management	The risks are associated with project management processes, organizational maturity, and ability				



Risk Identification

Risk Parameters:

Parameters for evaluating, categorizing, and prioritizing risks include the following:

- Risk likelihood (i.e., probability of risk occurrence)
- Risk consequence (i.e., impact and severity of risk occurrence)
- Thresholds to trigger management activities.



Risk Assessment

The risk assessment consists of evaluating the range of possible impacts should the risk occur.

Follow these steps when assessing risks:

- 1. Define the various impacts of each risk
- 2. Rate each impact based on a logical severity level
- 3. Sort and evaluate risks by severity level
- 4. Determine if any controls already exist
- 5. Define potential mitigation actions.



Risk Mitigation Planning

The risk owners are responsible for planning and implementing mitigation actions with support from the project team.

- All team members, inclusive of partners and suppliers, may be requested to identify and develop mitigation measures for identified risks.
- The project core team members are responsible for identifying an appropriate action owner for each identified risk.
- After mitigation actions are defined, the project core team will review the actions.
- The risk owner must track all mitigation actions and expected completion dates.
- The risk owner and the project core team members must hold all action owners accountable for the risk mitigation planning.



Risk Mitigation Action Implementation

- The action implementation is the responsibility of the risk owner.
- The action owners are responsible for the execution of the tasks or activities necessary to complete the mitigation action and eliminate or minimize the risk.
- The risk owner or the project manager will monitor completion dates of the mitigation action implementation.



Risk Occurrence and Contingency Plans

- Whenever any risk occurs, the project team should implement contingency plans to ensure that project deliverables can be met.
- The details of each occurrence should be recorded in the risk register or other tracking tool.
- The risk register or risk management plan (see next slide) will be maintained by the project manager and reviewed on a regular basis.



Risk Tracking and Reporting

- Risk tracking and reporting provides critical visibility to all risks.
- Risk owners must report on the status of their mitigation actions.
- Depending on the risk severity, project managers need to report the risk status of each category of risk to senior management.

Risk Management Plan

<u>Company</u>		Project/Program Name	<u>P</u>	roject Lead	Project Sponsor/Champion			<u>Last Updated</u>	
Risk ID	Risk Category	Risk Description	Risk Description			Impact Rating Mitigation Action		Responsible	Status



Risk Closure

- The risk owners are responsible for recommending the risk closure to the project manager.
- A risk is closed only when the item is not considered a risk to the project anymore.
- When a risk is closed, the project manager needs to update the risk status in the lessons learned document.



Risk Analysis Features

The risk analysis should be:

- Systematic
- Comprehensive
- Data driven
- Adherent to evidence
- Logically sound
- Practically acceptable
- Open to critique
- Easy to understand.



Project Risk Analysis Advantages

- Helps strategic and business planning
- Meets customer requirements
- Reduces schedule slips and cost overruns
- Promotes an effective usage of resources
- Promotes continuous improvement
- Helps to achieve project goals
- Minimizes surprises from customers and stakeholders
- Allows a quick grasp of new opportunities
- Enhances communication
- Reassures stakeholders that the project stays on track.



1.3.5 Project Planning



What is Project Management?

- **Project management** is the process of defining, planning, organizing, managing, leading, securing, and optimizing the resources to achieve a set of planned goals and objectives.
- It is the application of knowledge, skills, tools, and techniques to project activities in order to meet project requirements.



What is a Project Plan?

- A **project plan** is a crucial step in project management for achieving a project's goals.
- A project plan is a formal approved document used to guide and execute project tasks.
- It provides an overall framework for managing project tasks, schedules, and costs.
- A project plan is a coordinating tool and communication device that helps teams, contractors, customers, and organizations define the crucial aspects of a project or program.



Project Planning Stages

- 1. Determine project scope and objectives: Explore opportunities, identify and prioritize needs, consider project solutions.
- 2. Plan the project: Identify input and resources requirements such as human resources, materials, software, hardware, and budgets.
- 3. Prepare the project proposal: Based on stakeholder feedback, plan the necessary resources, timeline, budget etc.
- 4. Implement the project: Implement the project by engaging responsible resources and parties. Ensure execution and compliance of the defined plans.
- **5. Evaluate the project:** Regularly review progress and results. Measure the project's effectiveness against *quantifiable* requirements.



Planning and Scheduling Objectives

- To optimize the use of resources (both human and other resources).
- To increase productivity.
- To achieve desired schedules and deliverables.
- To establish an approach to minimize long-term maintenance costs.
- To minimize the chaos and productivity losses resulting from planned production schedules, priority changes, and non-availability of resources.
- To assess current needs and future challenges.



- Statement of work (SOW)
- Work breakdown structure (WBS)
- Resource estimation plan
- Project schedule
- Budget or financial plan
- Communication plan
- Risk management plan



Statement of Work (SOW) is a formal document often accompanying a contract that outlines specific expectations, limitations, resources and work guidelines intended to define the work or project.

SOW's:

- Define the scope of the project.
- Establish expectations and parameters of the project.
- Identify technical requirements for the project
- Provide guidance on materials to be used.
- Establish timeline expectations.
- Etc.

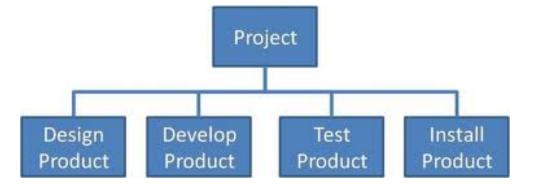




Work Breakdown Structure (WBS) is a decomposition of project components into small and logical bodies of work or tasks.

Work Breakdown Structures:

- Identify all required components of a project.
- Cascades components into sub-components and tasks.
- WBS's are not by themselves project plans or schedules but they are a necessary step to help establish project plans and timelines.
- WBS's also enable logical reporting and summarizations of project progress





Work Breakdown Structure Example: below is an example of a WBS for the assembly of a bicycle.

Level 1	Level 2	Level 3				
1.0 Bicycle	1.1 Frame	1.1.1 Set Body Frame				
		1.1.2 Assemble Handlebars				
		1.1.3 Install Seat & Seat Post				
		1.1.4 Assemble Wheel Bearings & Axle				
		1.1.5 Install Wheels				
		1.1.6 Attach Brake System to Frame Set				
	1.2 Wheels	1.2.1 Stage Rims				
		1.2.2 Install Spokes				
		1.2.3 Insert Tubes in Tires				
		1.2.4 Assemble Tires on Rims				
		1.2.5 Install Reflectors				
	1.3 Brake System	1.3.1 Connect Brake Cables To Hand Levers				
		1.3.2 Connect Brake Cables to Brake Harness				
		1.3.3 Attach Brake Pads to Brake Harness				
		1.3.4 Adjust/Calibrate Brake System				



Project Planning Activities: Resource Planning

Resource Estimation Plan

- Estimate Resource Requirements (Use Your WBS)
 - Parts, Hardware, Software, Human Resources etc.
- Plan resources
 - Establish who is responsible for what and when.
 - Determine quantity requirements and delivery dates etc.

Level 3	Parts Required	Quantity	Cost	Part Inventory	Responsible
1.1.1 Set Body Frame	Body Frame	1	\$9.00	4	John
1.1.2 Assemble Handlebars	Handlebars	1	\$4.00	4	John
1.1.3 Install Seat & Seat Post	Seat & Post	1 Each	\$3.00	4 Each	John
1.1.4 Assemble Wheel Bearings & Axle	Bearing & Axle	2 Each	\$1.00	8 Each	John
1.1.5 Install Wheels	Wheels	2	\$0.50	8	John
1.1.6 Attach Brake System to Frame Set					John
1.2.1 Stage Rims	Wheel Rim	2	\$0.30	12	Cathy
1.2.2 Install Spokes	Spokes	48	\$0.03	72	Cathy
1.2.3 Insert Tubes in Tires	Tubes & Tires	2 Each	\$0.33	0 Tubes 8 Tires	Cathy
1.2.4 Assemble Tires on Rims					Cathy
1.2.5 Install Reflectors	Reflectors	4	\$0.10	18	Cathy
1.3.1 Connect Brake Cables To Hand Levers	Cables & Levers	2 Each	\$0.90	12 Each	Lisa
1.3.2 Connect Brake Cables to Brake Harness	Brake Harness	2	\$0.40	8	Lisa
1.3.3 Attach Brake Pads to Brake Harness	Brake Pads	4	\$0.15	12	Lisa
1.3.4 Adjust/Calibrate Brake System					Lisa



Project Planning Activities: Project Scheduling

Project Scheduling:

- Assign beginning times to each activity in the WBS (days are used for start and durations times in example below).
- Assign duration times to each activity in the WBS.
- Identify People Responsible and set completion dates.
- Represent schedules as Gantt charts or network diagrams.
- Identify critical dependencies between tasks.

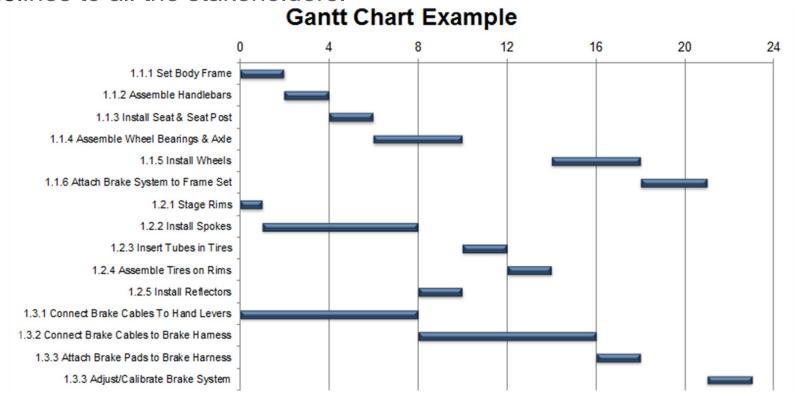
Level 2	Level 3	Task Start	Task Duration	Responsible
1.1 Frame	1.1.1 Set Body Frame	0	2	John
	1.1.2 Assemble Handlebars	2	2	John
	1.1.3 Install Seat & Seat Post	4	2	John
	1.1.4 Assemble Wheel Bearings & Axle	6	4	John
	1.1.5 Install Wheels	14	4	John
	1.1.6 Attach Brake System to Frame Set	18	3	John
1.2 Wheels	1.2.1 Stage Rims	0	1	Cathy
	1.2.2 Install Spokes	1	7	Cathy
	1.2.3 Insert Tubes in Tires	10	2	Cathy
	1.2.4 Assemble Tires on Rims	12	2	Cathy
	1.2.5 Install Reflectors	8	2	Cathy
1.3 Brake System	1.3.1 Connect Brake Cables To Hand Levers	0	8	Lisa
	1.3.2 Connect Brake Cables to Brake Harness	8	8	Lisa
	1.3.3 Attach Brake Pads to Brake Harness	16	2	Lisa
	1.3.4 Adjust/Calibrate Brake System	21	2	Lisa



Project Planning Activities: Project Scheduling

Project Schedule – Gantt Chart:

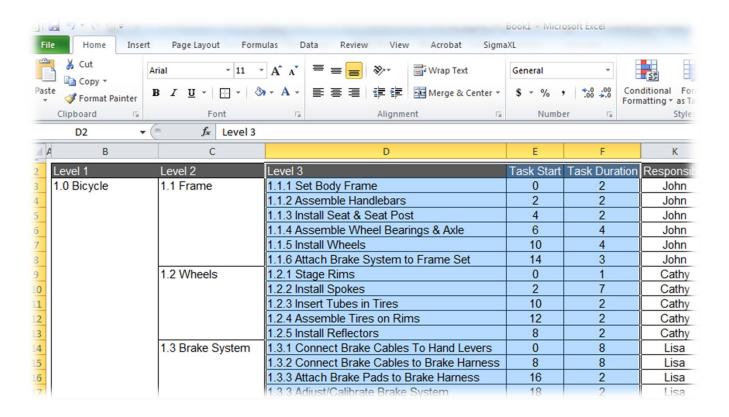
- The advantage of a Gantt chart is its ability to display the status of each task/activity at a glance.
- Because it is a graphic representation, it is easy to demonstrate the schedule of tasks and timelines to all the stakeholders.





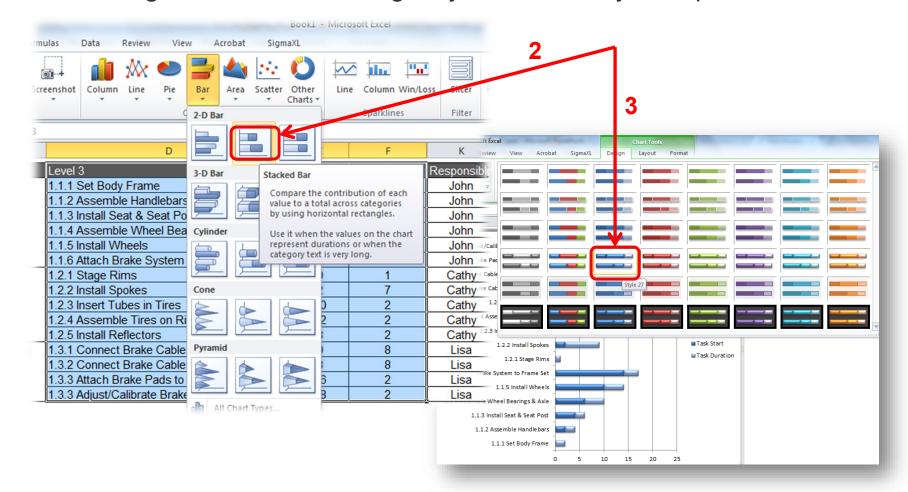
There are many tools available to create Gantt charts. Some of the most common and compatible for businesses are: Microsoft Project, Visio and Excel. In Microsoft Excel we use the bicycle example:

Select rows & columns of your tasks, start times and durations.



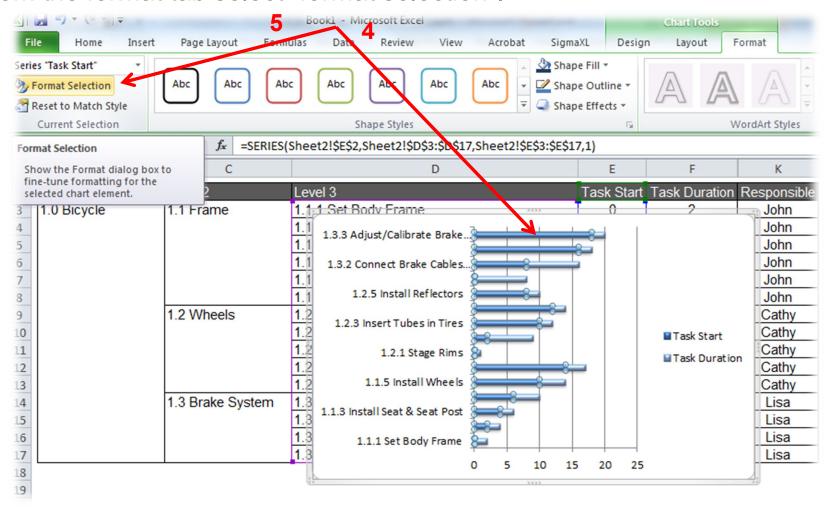


- On the Insert tab select Bar > Stacked Bar.
- 3. On the Design tab choose a design style from the style dropdown.



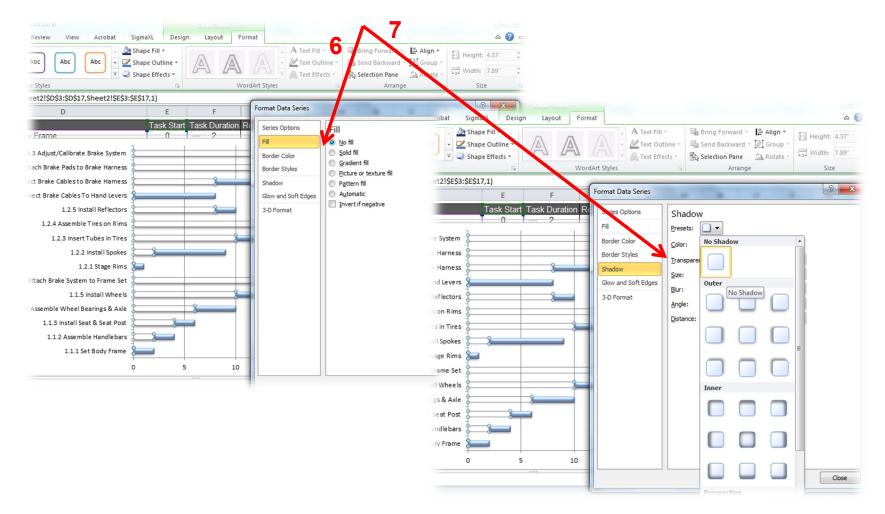


- 4. Activate the "start tasks" on your graph.
- From the format tab select "format selection".



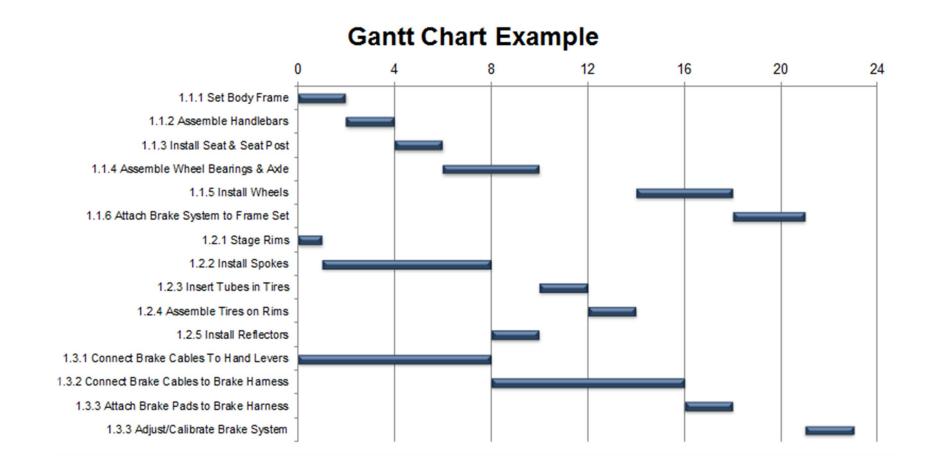


- 6. Select Fill > No Fill.
- Select Shadow > Presets > No Shadow > Close.





8. Format your new Gantt chart to your own preferences.





Critical Path Method (CPM) is a project modeling technique used to identify the set of activities that are most influential to a projects completion timeline.

- Critical Path tasks are those that others are dependent upon.
- Project timelines cannot be shortened without shortening the tasks or activities that are identified as critical path items.

Steps To Using The Critical Path Method

- 1. List all activities necessary to complete the project.
- 2. Determine the time or duration of each activity.
- 3. Identify the dependencies between the activities.

Use Your Work Breakdown Structure (WBS).



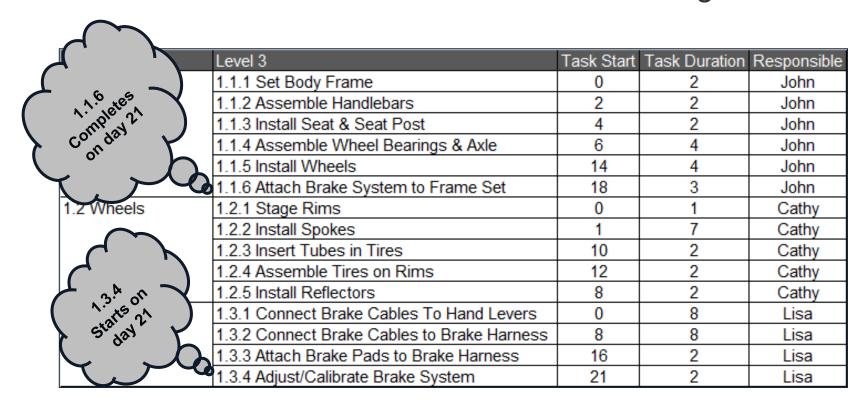
Critical Path Method Example: Let's continue with our bicycle example. Note that the tasks in the WBS are already set to start based on their dependencies.

Level 2	Level 3	Task Start	Task Duration	Responsible
1.1 Frame	1.1.1 Set Body Frame	0	2	John
	1.1.2 Assemble Handlebars	2	2	John
	1.1.3 Install Seat & Seat Post	4	2	John
	1.1.4 Assemble Wheel Bearings & Axle	6	4	John
	1.1.5 Install Wheels	14	4	John
	1.1.6 Attach Brake System to Frame Set	18	3	John
1.2 Wheels	1.2.1 Stage Rims	0	1	Cathy
	1.2.2 Install Spokes	1	7	Cathy
	1.2.3 Insert Tubes in Tires	10	2	Cathy
	1.2.4 Assemble Tires on Rims	12	2	Cathy
	1.2.5 Install Reflectors	8	2	Cathy
1.3 Brake System	1.3.1 Connect Brake Cables To Hand Levers	0	8	Lisa
	1.3.2 Connect Brake Cables to Brake Harness	8	8	Lisa
	1.3.3 Attach Brake Pads to Brake Harness	16	2	Lisa
	1.3.4 Adjust/Calibrate Brake System	21	2	Lisa

• For example, Installing the wheels in step 1.1.5 requires the whole set of steps in 1.2 to be completed so 1.1.5 starts on day 14 when the wheels are ready.



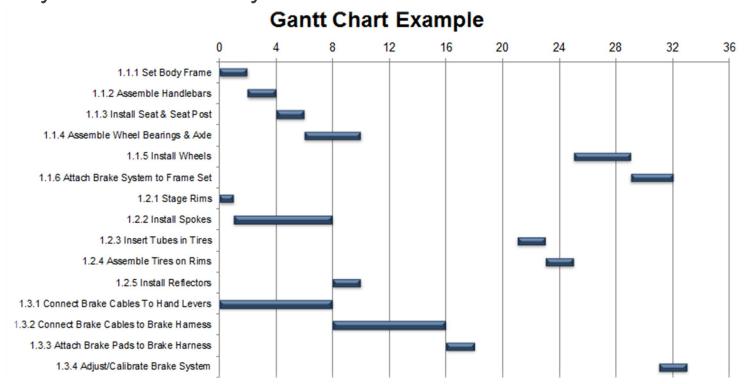
Continuing this example; the last task, 1.3.4 Adjusting the Brake System, completes on day 23 and can't begin until day 21 because the brake system must be attached to the bike frame before being calibrated.





Earlier if you noticed when we conducted our resource plan that we were out of tire tubes then you might have managed your resources differently.

Let's assume the tubes take 21 days to receive from the supplier. Below is the adjustment to our Gantt chart. Notice that the project timeline is pushed out by 8 days, finishing on day 31 instead of day 23.





Project Scheduling: PERT

Program Evaluation & Review Technique (PERT): is a method of evaluation that can be applied to time or cost.

- PERT provides a weighted assessment of time or cost
- PERT uses 3 parameters for estimation:
 - Optimistic or Best Case Scenario represented by 'O'.
 - · Pessimistic or Worst Case Scenario represented by 'P'.
 - Most Likely Scenario represented by 'ML'.
- PERT Equation:

$$(O + (4*ML) + P) / 6$$

 The PERT equation provides for heavier weighting of the most likely scenario but also considers the best and worst cases. In the event that a best or worst case scenario is an extreme situation, the PERT will account for it in a weighted manner.



Project Scheduling: PERT

Let's apply the PERT formula to our estimates of receiving our needed tire tubes from the supplier and then change our schedule based on the result.

- Your experience with this supplier tells you that they typically over estimate the time required to deliver. Therefore, their 21 day lead time on the tubes should be considered a "worst case scenario".
- Procurement has indicated that tubes have arrived in as few as 6 days from this supplier. So, your best case scenario is 6 days.
- The most likely scenario you decide will be the median of this suppliers delivery time which is 10 days.
- Therefore:



Project Scheduling: PERT

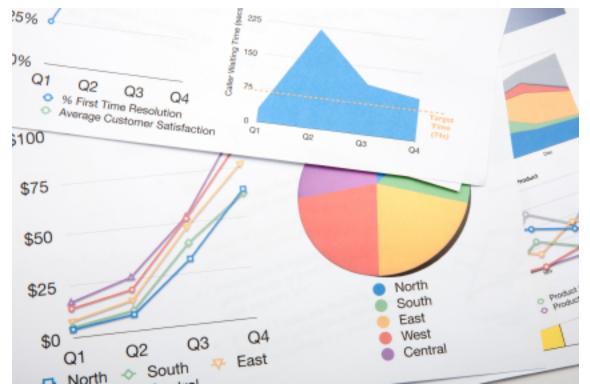
Using 11 days in your project plan instead of 21 gives you a more confident estimate of time and actually removes tire tubes as the primary detracting critical path item.

Level 3	Task Start	Task Duration								
1.1.1 Set Body Frame	0	2								
1.1.2 Assemble Handlebars	2	2								
1.1.3 Install Seat & Seat Post	4	2								
1.1.4 Assemble Wheel Bearings & Axle	6	4								
1.1.5 Install Wheels	15	4								
1.1.6 Attach Brake System to Frame Set	19	3								
1.2.1 Stage Rims	0	1	Ca	ntt C	har	4 Eva	mnla			
1.2.2 Install Spokes	1	7	Ga	nu C	·IIai	t Exa	ilibie			
1.2.3 Insert Tubes in Tires	11	2			1	8	12	16	20	24
1.2.4 Assemble Tires on Rims	13	2	1		+		12	10	20	24
1.2.5 Install Reflectors	8	2	t Body Frame	_						
1.3.1 Connect Brake Cables To Hand Levers	0	8	e Handlebars	_	_					
1.3.2 Connect Brake Cables to Brake Harness	8	8	at & Seat Post			.				
1.3.3 Attach Brake Pads to Brake Harness	16	2	-							
1.3.4 Adjust/Calibrate Brake System	21	2	arings & Axle				1			
		1.1.5	Install Wheels						_	
	1.1.6 Att	ach Brake Systen	n to Frame Set							
		1.2	2.1 Stage Rims	_						
		1.2.2	Install Spokes							
		1.2.3 Insert	Tubes in Tires							
		1.2.4 Assemble	Tires on Rims					_		
		1.2.5 In	stall Reflectors			_	. -	_		
	1.3.1 Connec	t Brake Cables T	-							
		Brake Cables to	-							
1			-							
	1.3.3 Attac	h Brake Pads to I	Brake Harness						4	
	1.3.4	Adjust/Calibrate	Brake System						_	_



Budget or Financial Plan

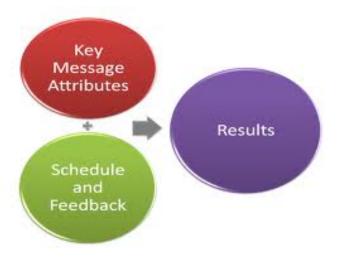
- Planned expenses
- Planned revenues
- Budget forecast





Communication Plan

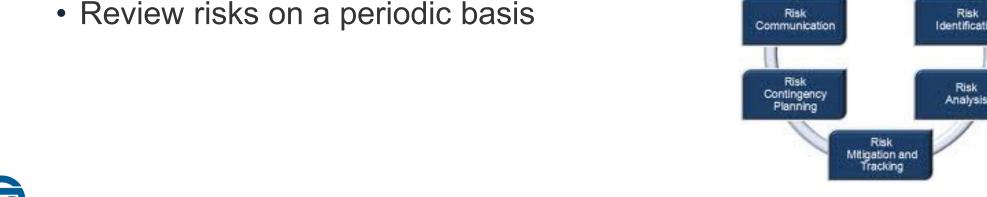
- Establish communication procedures among management, team members, and relevant stakeholders.
- Determine the communication schedule.
- Define the acceptable modes of communication.





Risk Management Plan

- Identify the sources of project risks and estimate the effects of those risks.
- Risks might arise from new technology, availability of resources, lack of inputs from customers, business risks etc.
- Assess the impact of risk to the customers/stakeholders.
- Calculate the probability of risk occurrence based on previous similar projects or industry benchmarks
- Initiate mitigation and contingency plans
- Review risks on a periodic basis





Project Planning Tools Advantages

- Project planning tools are very useful to organize and communicate project plans, status, and projections.
- They help link tasks and sub-tasks or other work elements to get a whole view of what needs to be accomplished.
- They allow a more objective comparison of alternative solutions and provide consistent coverage of responsibilities.
- They allow for effective scope control and change management.
- They facilitate effective communication with all project participants and stakeholders.
- They help define management reviews.
- They act as an effective monitoring mechanism for the project.
- They establish project baselines for progress reviews and control points.



1.4 Lean Fundamentals



Black Belt Training: Define Phase

1.1 Six Sigma Overview

- 1.1.1 What is Six Sigma
- 1.1.2 Six Sigma History
- 1.1.3 Six Sigma Approach Y = f(x)
- 1.1.4 Six Sigma Methodology
- 1.1.5 Roles and Responsibilities

1.2 Six Sigma Fundamentals

- 1.2.1 Defining a Process
- 1.2.2 VOC and CTQs
- 1.2.3 QFD
- 1.2.4 Cost of Poor Quality (COPQ)
- 1.2.5 Pareto Analysis (80:20 rule)

1.3 Lean Six Sigma Projects

- 1.3.1 Six Sigma Metrics
- 1.3.2 Business Case and Charter
- 1.3.3 Project Team Selection
- 1.3.4 Project Risk Management
- 1.3.5 Project Planning

1.4 Lean Fundamentals

- 1.4.1 Lean and Six Sigma
- 1.4.2 History of Lean
- 1.4.3 The Seven Deadly Muda
- 1.4.4 Five-S (5S)



1.4.1 Lean and Six Sigma



What is Lean?

- A lean enterprise intends to eliminate waste and allow only value to be pulled through its system.
- Lean manufacturing is characterized by:
 - Identifying and driving value
 - Establishing flow and pull systems
 - Creating production availability and flexibility
 - Zero waste
- Waste Elimination
 - Waste identification and elimination is critical to any successful lean enterprise.
 - Elimination of waste enables flow, drives value, cuts cost, and provides flexible and available production.





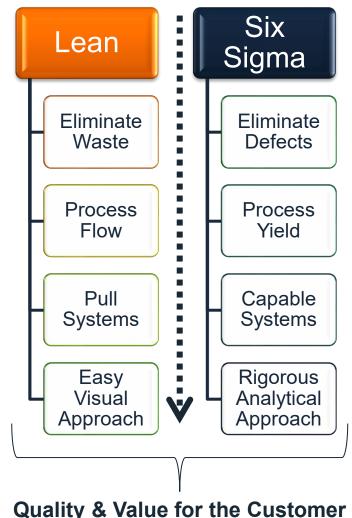
The 5 Lean Principles

- The following 5 principles of lean are taken from the book Lean Thinking (1996) by James P. Womack and Daniel T. Jones.
 - 1. Specify value desired by customers.
 - 2. Identify the value stream.
 - Make the product flow continuous.
 - 4. Introduce pull systems where continuous flow is possible.
 - 5. Manage toward perfection so that the number of steps and the amount of time and information needed to serve the customer continually falls.



Lean & Six Sigma

- Lean and Six Sigma both have the objectives of producing high value (quality) at lower costs (efficiency).
- They approach these objectives in somewhat different manners but in the end, both Lean and Six Sigma drive out waste, reduce defects, improve processes, and stabilize the production environment.
- Lean and Six Sigma are a perfect combination of tools for improving quality and efficiency.



Quality & Value for the Customer Efficiency for the Business



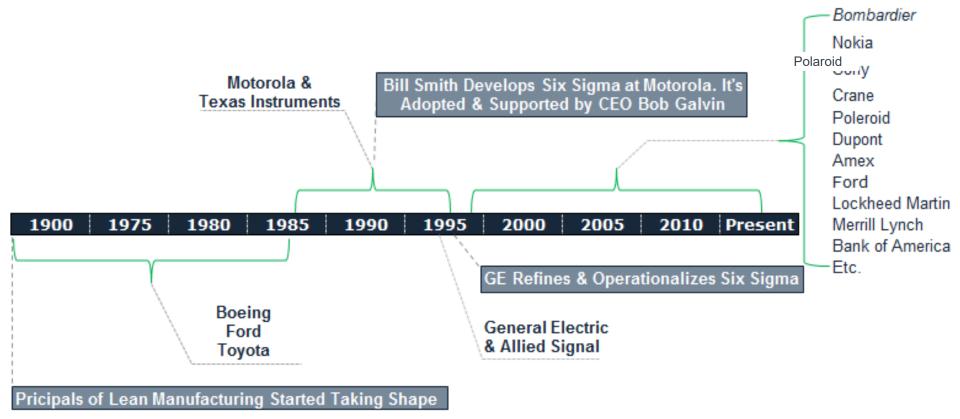
1.4.2 History of Lean



History of Lean

Lean Six Sigma

History & Timeline





History of Lean

- Lean thinking originated, as far as is known, the 1400s.
- Henry Ford established the first mass production system in 1913 by combining standard parts, conveyors, and work flow.
- Decades later, Kiichiro Toyoda and Taiichi Ohno at Toyota improved and implemented various new concepts and tools (e.g., value stream, takt time, kanban etc.) based on Ford's effort.
- Toyota developed what is known today as the Toyota Production System (TPS) based on lean principles.



History of Lean

- Starting in the mid 1990s, Lean became extensively recognized and implemented when more and more Fortune 100 companies began to adopt Lean and Six Sigma.
- The term "Lean manufacturing" was introduced by James Womack in the 1990s.
- Lean and Six Sigma share similar objectives, work hand in hand, and have benefited from one another in the past 30 years.

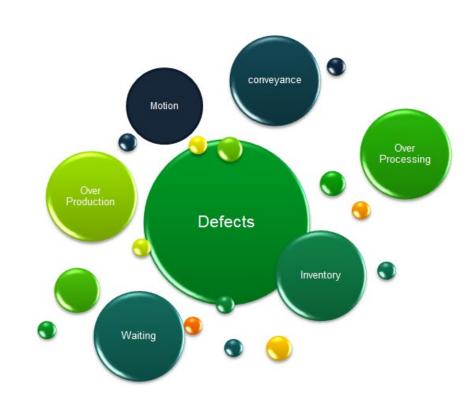


1.4.3 Seven Deadly Muda



The 7 Deadly Muda

- The Japanese word for waste is "muda."
- There are 7 commonly recognized forms of waste, often referred to as the "7 deadly muda."
 - 1. Defects
 - 2. Overproduction
 - 3. Over-Processing
 - 4. Inventory
 - 5. Motion
 - 6. Transportation
 - 7. Waiting





The 7 Deadly Muda: Defects

 Defects or defectives are an obvious waste for any working environment or production system.



- Defects require rework during production and/or after the product is returned from an unhappy customer.
- Some defects are difficult to solve and they create "workarounds" and hidden factories.
- Eliminating defects is a sure way to improve product quality, customer satisfaction, and production costs.



The 7 Deadly Muda: Overproduction

 Overproduction is wasteful because your system expends energy and resources to produce more materials than the customer or next function requires.

- Overproduction is one of the most detrimental of the seven deadly muda because it leads to many others:
- Inventory
- Transportation
- Waiting etc.



The 7 Deadly Muda: Over-processing



Over-processing occurs any time more work is done than is required by the next process step, operation, or consumer.

- Over-processing also includes being over capacity (scheduling more workers than required or having more machines than necessary).
- Another form of over processing can be buying tools or software that are overkill (more precise, complex, or expensive than required).



The 7 Deadly Muda: Inventory



- **Inventory** is an often overlooked waste. Look at the picture above and imagine all the time, materials, and logistics that went into establishing such an abundance of inventory.
- If this were your personal business, and inventory velocity was not matched with production, how upset would you be?



The 7 Deadly Muda: Motion



- **Motion** is another form of waste often occurring as a result of poor setup, configuration, or operating procedures.
- Wasted motion can be experienced by machines or humans.
- Wasted motion is very common with workers who are unaware of the impact of small unnecessary movements in repetitive tasks.
- Wasted motion is exaggerated by repetition or recurring tasks.



The 7 Deadly Muda: Transportation



- Transportation is considered wasteful because it does nothing to add value or transform the product.
- Imagine for a moment driving to and from work twice before getting out of your car to go into work...
- That is waste in the form of transportation.
- The less driving you have to do, the better.
- In a similar way, the less transportation a product has to endure, the better. There would be fewer opportunities for delay, destruction, loss, damage etc.



The 7 Deadly Muda: Waiting



- Waiting is an obvious form of waste and is typically a symptom of an upstream problem.
- Waiting is usually caused by inefficiency, bottlenecks, or poorlydesigned work flows within the value stream.
- Waiting can also be caused by inefficient administration.
- Reduction in waiting time will require thoughtful applications of lean and process improvement.



1.4.4 Five-S (5S)



What is 5S?

- 5S is systematic method to organize, order, clean, and standardize a workplace...and to keep it that way!
 - 5S is a methodology of organizing and improving the work environment.
- 5S is summarized in five Japanese words all starting with the letter S:
 - Seiri (sorting)
 - Seiton (straightening)
 - Seiso (shining)
 - Seiketsu (standardizing)
 - Shisuke (sustaining)
- 5S was originally developed in Japan and is widely used to optimize the workplace to increase productivity and efficiency.

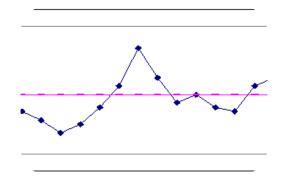


Five-S (5S)









Sustain



Goals of 5S

- Reduced waste
- Reduced cost
- Establish a work environment that is:
 - self-explaining
 - self-ordering
 - self-regulating
 - self-improving.
 - Where there is/are no more:
 - Wandering and/or searching
 - Waiting or delaying
 - Secret hiding spots for tools
 - Obstacles or detours
 - Extra pieces, parts, materials etc.
 - Injuries
 - Waste.



Benefits of 5S Systems

- Reduced changeovers
- Reduced defects
- Reduced waste
- Reduced delays
- Reduced injuries
- Reduced breakdowns
- Reduced complaints
- Reduced red ink
- Higher quality
- Lower costs
- Safer work environment
- Greater associate and equipment capacity.



Reported Results of 5S Systems

 Cut in floor space: 	60%
 Cut in flow distance: 	80%
 Cut in accidents: 	70%
 Cut in rack storage: 	68%
 Cut in number of forklifts: 	45%
 Cut in machine changeover time: 	62%
 Cut in annual physical inventory time: 	50%
 Cut in classroom training requirements: 	55%
 Cut in nonconformance in assembly: 	96%

 α

50%

15%

Late deliveries:

• Increase in throughput:

Increase in test yields:



Sorting (Seiri)



- Go through all the tools, parts, equipment, supply, and material in the workplace.
- Categorize them into two major groups: needed and unneeded.
- Eliminate the unneeded items from the workplace. Dispose of or recycle those items.
- Keep the needed items and sort them in the order of priority. When in doubt...throw it out!



Straightening (Seiton)



- Straightening in 5S is also called setting in order.
- Label each needed item.
- Store items at their best locations so that the workers can find them easily whenever they needed any item.
- Reduce the motion and time required to locate and obtain any item whenever it is needed.
- Promote an efficient work flow path.
- Use visual aids like the tool board image on this page.



Shining (Seiso)



- Clean the workplace thoroughly.
- Maintain the tidiness of the workplace.
- Make sure every item is located at the specific location where it should be.
- Create the ownership in the team to keep the work area clean and organized.





Standardizing (Seiketsu)

- Standardize the workstation and the layout of tools, equipment and parts.
- Create identical workstations with a consistent way of storing the items at their specific locations so that workers can be moved around to any workstation any time and perform the same task.



Sustaining (Shisuke)

- Sustaining in 5S is also called self-discipline.
- Create the culture in the team to follow the first four S's consistently.
- Avoid falling back to the old ways of cluttered and unorganized work environment.
- Keep the momentum of optimizing the workplace.
- Promote innovations of workplace improvement.
- Sustain the first fours S's using:
 - 5S Maps
 - 5S Schedules
 - 5S Job cycle charts
 - Integration of regular work duties
 - 5S Blitz schedules
 - Daily workplace scans.



Simplified Summary of 5S

- 1. Sort "when in doubt, move it out."
- 2. Set in Order Organize all necessary tools, parts, and components of production. Use visual ordering techniques wherever possible.
- 3. Shine Clean machines and/or work areas. Set regular cleaning schedules and responsibilities.
- **4. Standardize** Solidify previous three steps, make 5S a regular part of the work environment and everyday life.
- **5. Sustain** Audit, manage, and comply with established 5S guidelines for your business or facility.



Five-S (5S)

- A few words about 5S and the Lean Enterprise
 - As a method, 5S generates immediate improvements.
 - 5S is one of many effective lean methods that create observable results.
 - It is tempting to implement 5S alone without considering the entire value stream.
 - However, it is advisable to consider a well-planned lean manufacturing approach to the entire production system.



2.0 Measure Phase



Black Belt Training: Measure Phase

2.1 Process Definition

- 2.1.1 Cause and Effect Diagrams
- 2.1.2 Cause and Effects Matrix
- 2.1.3 Process Mapping
- 2.1.4 FMEA: Failure Modes & Effects Analysis
- 2.1.5 Theory of Constraints

2.2 Six Sigma Statistics

- 2.2.1 Basic Statistics
- 2.2.2 Descriptive Statistics
- 2.2.3 Distributions and Normality
- 2.2.4 Graphical Analysis

2.3 Measurement System Analysis

- 2.3.1 Precision and Accuracy
- 2.3.2 Bias, Linearity, and Stability
- 2.3.3 Gage R&R
- 2.3.4 Variable and Attribute MSA

2.4 Process Capability

- 2.4.1 Capability Analysis
- 2.4.2 Concept of Stability
- 2.4.3 Attribute and Discrete Capability
- 2.4.4 Monitoring Techniques



2.1 Process Definition



Black Belt Training: Measure Phase

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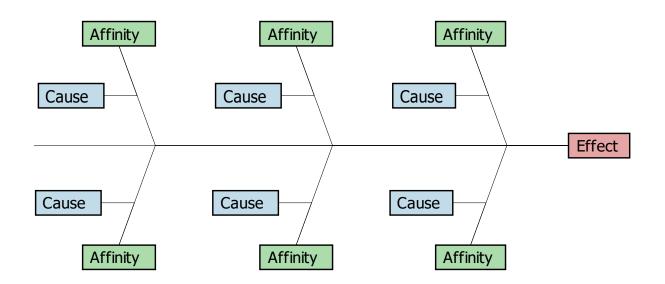


2.1.1 Cause and Effect Diagram



What is a Cause and Effect Diagram?

- A cause and effect diagram is also called a *Fishbone Diagram* or *Ishikawa Diagram*. It was created by Kaoru Ishikawa and is used to identify, organize, and display the potential causes of a specific effect or event in a graphical way similar to a fishbone.
- It illustrates the relationship between one specified event (output) and its categorized potential causes (inputs) in a visual and systematic way.





Major Categories of Potential Causes

P4ME

- People: People who are involved in the process
- Methods: How the process is completed (e.g., procedures, policies, regulations, laws)
- Machines: Equipment or tools needed to perform the process
- Materials: Raw materials or information needed to do the job
- Measurements: Data collected from the process for inspection or evaluation
- Environment: Surroundings of the process (e.g., location, time, culture).



- Step 1: Identify and define the effect/event being analyzed.
 - Clearly state the operational definition of the effect/event of interest.
 - The event can be the positive outcome desired or negative problem targeted to solve.
 - Enter the effect/event in the end box of the Fishbone diagram and draw a spine pointed to it.



Step 1

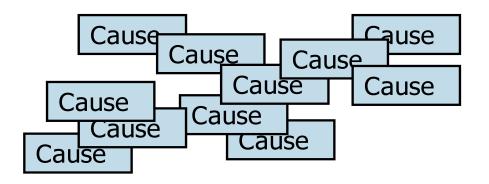
Effect or Event Being Analyzed



- Step 2: Brainstorm the potential causes or factors of the effect/event occurring.
 - Identify any factors with a potential impact on the effect/event and include them in this step.
 - Put all the identified potential causes aside for use later.



• Step 2

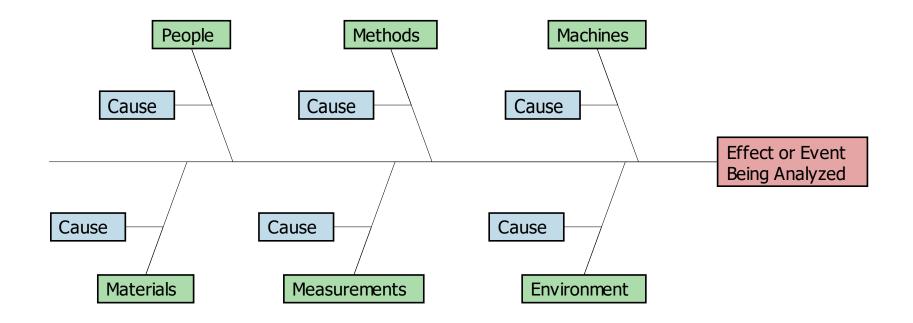




- Step 3: Identify the main categories of causes and group the potential causes accordingly.
 - Besides P4ME (i.e., people, methods, machines, materials, measurements, and environment), you can group potential causes into other customized categories.
 - Below each major category, you can define sub-categories and then classify them to help you visualize the potential causes.
 - Enter each cause category in a box and connect the box to the spine. Link each potential cause to its corresponding cause category.



• Step 3





- Step 4: Analyze the cause and effect diagram.
 - A cause and effect diagram includes all the possible factors of the effect/event being analyzed.
 - Use a Pareto chart to filter causes the project team needs to focus on.
 - Identify causes with high impact that the team can take action upon.
 - Determine how to measure causes and effects quantitatively. Prepare for further statistical analysis.



Benefits to Using Cause and Effect Diagram

- Helps to quickly identify and sort the potential causes of an effect.
- Provides a systematic way to brainstorm potential causes effectively and efficiently.
- Identifies areas requiring data collection for further quantitative analysis.
- Locates "low-hanging fruit."



Limitation of Cause and Effect Diagrams

- A cause and effect diagram only provides qualitative analysis of correlation between each cause and the effect.
- One cause and effect diagram can only focus on one effect or event at a time.
- Further statistical analysis is required to quantify the relationship between various factors and the effect and identify the root causes.



- Case study:
 - A real estate company is interested to find the root causes of high energy costs of its properties.
 - The cause and effect diagram is used to identify, organize, and analyze the potential root causes.

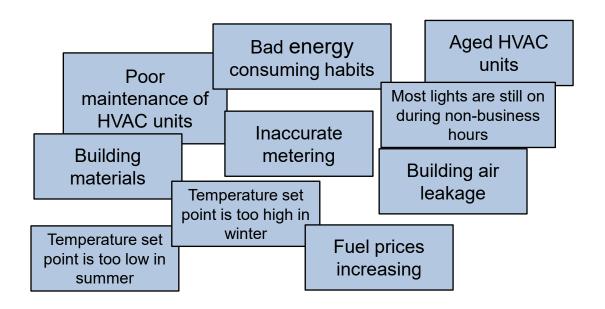


 Step 1: Identify and define the effect/event being analyzed: high energy costs of buildings.

High Energy
Cost of Buildings

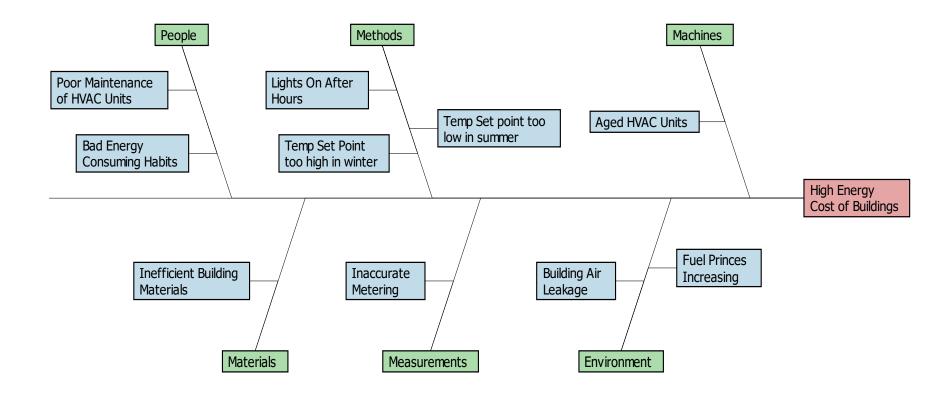


 Step 2: Brainstorm the potential causes or factors of the high energy costs.





 Step 3: Identify the main categories of causes and group the potential causes accordingly.





- Step 4: Analyze the cause and effect (C&E) diagram.
 - After completing the C&E diagram, the real estate company conducts further research on each potential root cause.
 - It is discovered that:
 - The utility metering is accurate
 - The building materials are fine and there is not significant amount of air leakage from the building
 - The fuel prices increased recently but were negligible
 - Most lights are off during the non-business hours except that some lights have to be on for security purposes
 - The temperature set points in the summer and winter are both adequate and reasonable
 - The high energy costs are probably caused by the poor HVAC maintenance on aged units and the wasteful energy consuming habits.
- Next, the real estate company needs to collect and analyze the data to check whether
 root causes identified in the C&E diagram are statistically the causes of the high
 energy costs.



2.1.2 Cause and Effects Matrix



What is a Cause and Effect Matrix?

- The cause and effect matrix (XY Matrix) is a tool to help subjectively quantify the relationship of several X's to several Y's.
- Among the Y's under consideration, two important ones should be the primary and secondary metrics of your Six Sigma project.
- The X's should be derived from your cause and effect diagram. Let us take a peek as what it looks like on the next page.



Cause and Effects Matrix

		Lea	an Si	x Sig	ma λ	(Y Ma	atrix				
Date:											
Project:											
XY Matrix Owner:											
Output Measures (Y's)*	Y ₁	Y ₂	Y ₃	Y ₄	Y ₅	Υ ₆	Y ₇	Y ₈	Υ ₉	Y ₁₀	
Weighting (1-10):											
Input Variables (X's)#		For	each X, sc	ore its impa	act on eacl	n Y listed a	bove (use a	0,3,5,7 sc	ale)		Score
X ₁											0
X ₂											0
X ₃											0
X ₄											0

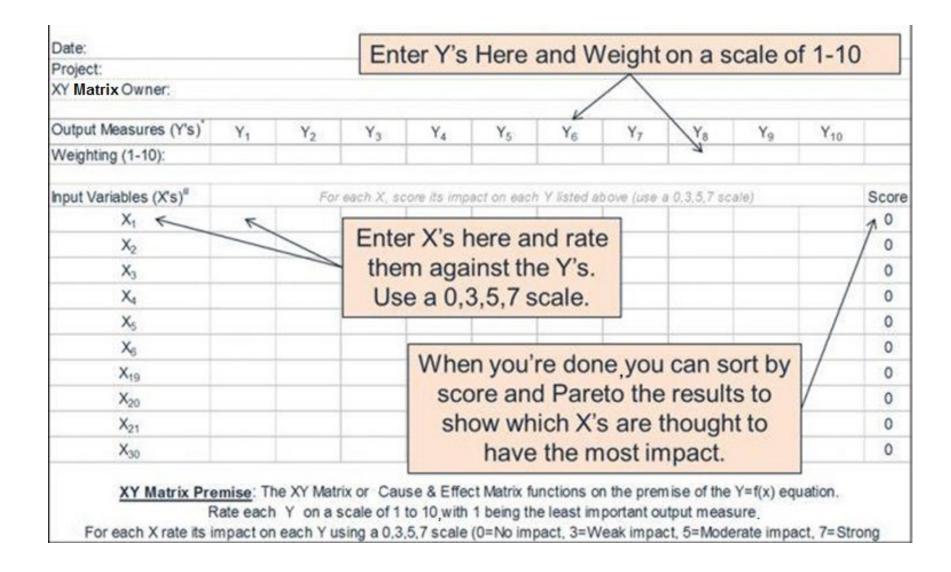


How to Use a Cause and Effect Matrix

- Across the top enter your output measures. These are the Y's that are important to your project.
- 2. Next, give each Y a weight. Use a 1–10 scale, 1 being least important and 10 most important.
- 3. Below, in the leftmost column, enter all the variables you identified with your cause and effect diagram.
- 4. Within the matrix itself, rate the strength of the relationship between the X in the row and the corresponding Y in that column. Use a scale of 0, 3, 5, and 7.
- 5. Lastly, sort the "Score" column to order the most important X's first.



Cause and Effect Matrix Notes





After You Have Completed the C&E Matrix

After you have completed your cause and effects matrix, build a strategy for validating and/or eliminating the x's as significant variables to the Y=f(x) equation.

- Build a data collection plan
- Prepare and execute planned studies
- Perform analytics
- Review results with SMEs
- etc.



2.1.3 Process Mapping



What is a Process Map?

- Process mapping is an important aspect to begin understanding a process.
 It provides a visual display of the steps required to produce a good or service.
- There are various forms of process mapping techniques that are used in the DMAIC methodology. We have already covered three common and useful maps in our "Process Definition" section of the Define phase:
 - High Level Process Map
 - Detailed Process Map
 - Functional Process Map
- Now we will explore a few other value process mapping techniques:
 - SIPOC
 - Value Stream
 - Spaghetti Diagram
 - Thought Process Map



What is SIPOC?

- A SIPOC (Suppliers-Input-Process-Output-Customers) is a high-level visualization tool to help identify and link the different components in a process.
- It is usually applied in the Measure phase in order to better understand the current state of the process and define the scope of the project.



Key Components of a SIPOC

- Suppliers: vendors who provide the raw material, services, and information. Customers can also be suppliers sometimes.
- Input: the raw materials, information, equipment, services, people, environment involved in the process.
- Process: the high-level sequence of actions and decisions that results in the services or products delivered to the customers.
- Output: the services or products delivered to the customers and any other outcomes of the process.

Customers: the end users or recipients of the services or products.

How to Plot a SIPOC Diagram

- The first method:
 - Step 1: Create a template that can contain the information of the five key components in a clear way.
 - Step 2: Plot a high-level process map that covers five steps at maximum.
 - Step 3: Identify the outputs of the process.
 - Step 4: Identify the receipt of the process.
 - Step 5: Brainstorm the inputs required to run each process step.
 - Step 6: Identify the suppliers who provide the inputs.



How to Plot a SIPOC Diagram

- The second method:
 - Step 1: Create a template that can contain the information of the five key components in a clear way.
 - Step 2: Identify the receipt of the process.
 - Step 3: Identify the outputs of the process.
 - Step 4: Plot a high-level process map that covers five steps at maximum.
 - Step 5: Brainstorm the inputs required to run each process step.
 - Step 6: Identify the suppliers who provide the inputs.



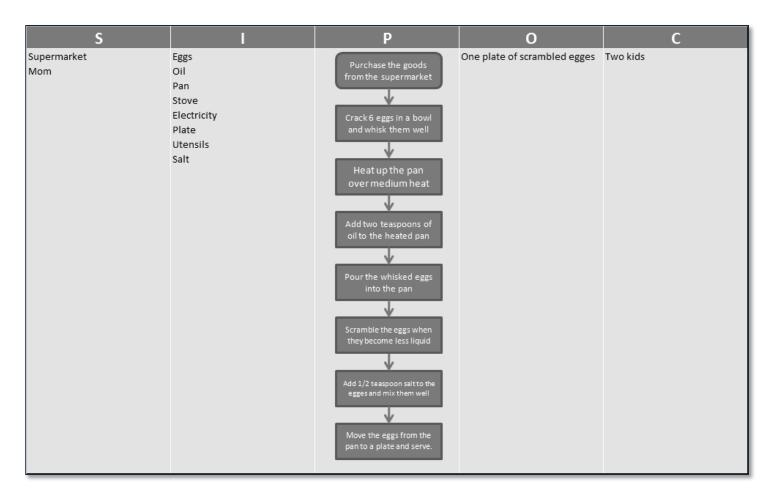
Benefits of SIPOC Diagrams

- A SIPOC diagram provides more detailed information than process maps and it demonstrates how each component gets involved in the process.
- It helps visualize and narrow the project scope.
- It serves as a great communication tool to help different process owners understand the entire process, their specific roles and responsibilities.



SIPOC Diagram Example

 Example of plotting a SIPOC diagram for Mom cooking scrambled eggs for two kids





- Step 1: Vertically List High-Level Process
 - If you followed the general rules for a high-level process map, then you should have no more than 4–6 steps for your process.
 - List those steps in a vertical manner as depicted below.

SUPPLIERS	INPUTS	PROCESS	OUTPUTS	CUSTOMERS
		Start		
		Step 1		
		Step 2		
		Step 3		
		Last Step		



Step 2: List Process Outputs

SUPPLIERS	INPUTS	PROCESS	OUTPUTS	CUSTOMERS
		Start		
		Step 1	Enter Step 1 Outputs	
		Step 2	Enter Step 2 Outputs	
		Step 3	Enter Step 3 Outputs	
		Last Step	Enter Step 4 Outputs	



Step 3: List Output Customers

SUPPLIERS	INPUTS	PROCESS	OUTPUTS	CUSTOMERS
		Start		
		Step 1	Enter Step 1 Outputs	Enter Step 1 Customers
		Step 2	Enter Step 2 Outputs	Enter Step 2 Customers
		Step 3	Enter Step 3 Outputs	Enter Step 3 Customers
		Last Step	Enter Step 4 Outputs	Enter Step 4 Customers



Step 4: List Process Inputs

SUPPLIERS	INPUTS	PROCESS	OUTPUTS	CUSTOMERS
		Start		
	Enter Step 1 Inputs	Step 1	Enter Step 1 Outputs	Enter Step 1 Customers
	Enter Step 2 Inputs	Step 2	Enter Step 2 Outputs	Enter Step 2 Customers
	Enter Step 3 Inputs	Step 3	Enter Step 3 Outputs	Enter Step 3 Customers
	Enter Step 4 Inputs	Last Step	Enter Step 4 Outputs	Enter Step 4 Customers



Creating a SIPOC

Step 5: List Suppliers of Inputs

SUPPLIERS	INPUTS	PROCESS	OUTPUTS	CUSTOMERS
		Start		
Enter Step 1	Enter Step 1	Step 1	Enter Step 1	Enter Step 1
Suppliers	Inputs		Outputs	Customers
Enter Step 2	Enter Step 2	Step 2	Enter Step 2	Enter Step 2
Suppliers	Inputs		Outputs	Customers
Enter Step 3	Enter Step 3	Step 3	Enter Step 3	Enter Step 3
Suppliers	Inputs		Outputs	Customers
Enter Step 4	Enter Step 4	Last Step	Enter Step 4	Enter Step 4
Suppliers	Inputs		Outputs	Customers



SIPOC Benefits

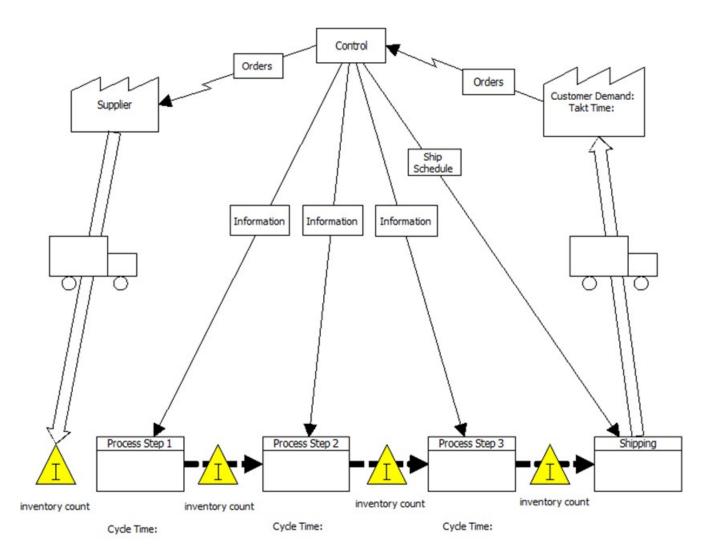
- Visually communicate project scope
- Identify key inputs and outputs of a process
- Identify key suppliers and customers of a process
- Verify:
 - Inputs match outputs for upstream processes
 - Outputs match inputs for downstream processes.
- This type of mapping is effective for identifying opportunities for improvement of your process.
- If you have completed your high-level process map, follow the outlined steps to create a process map of Suppliers, Inputs, Process, Outputs, and Customer.



What is Value Stream Mapping?

• Value stream mapping is a method to visualize and analyze the path of how information and raw materials are transformed into products or services customers receive.

 It is used to identify, measure, and decrease the non-value-adding steps in the current process.





Non-Value-Added Activities

- Non-value-adding activities are activities in a process that do not add any other value to the products or services customers demand.
- Example of non-value-adding activities:
 - Rework
 - Overproduction
 - Excess transportation
 - Excess stock
 - Waiting
 - Unnecessary motion.
- Not all non-value-adding activities are unnecessary.



How to Plot a Value Stream Map

- Plot the entire high-level process flow from when the customer places the order to when the customer receives the products or services in the end.
- A value stream map requires more detailed information for each step than the standard process map.
 - Cycle time
 - Preparation time
 - Actual working time
 - Available time
 - Scrap rate
 - Rework rate
 - Number of operators
- Assess the value stream map of current process, identity and eliminate the waste.



Spaghetti Chart

- A spaghetti chart is a graphical tool to map out the physical flow of materials, information, and people involved in a process. It can also reflect the distances between multiple workstations the physical flow has been through.
- A process that has not been streamlined has messy and wasteful movements of materials, information, and people, resembling a bowl of cooked spaghetti.

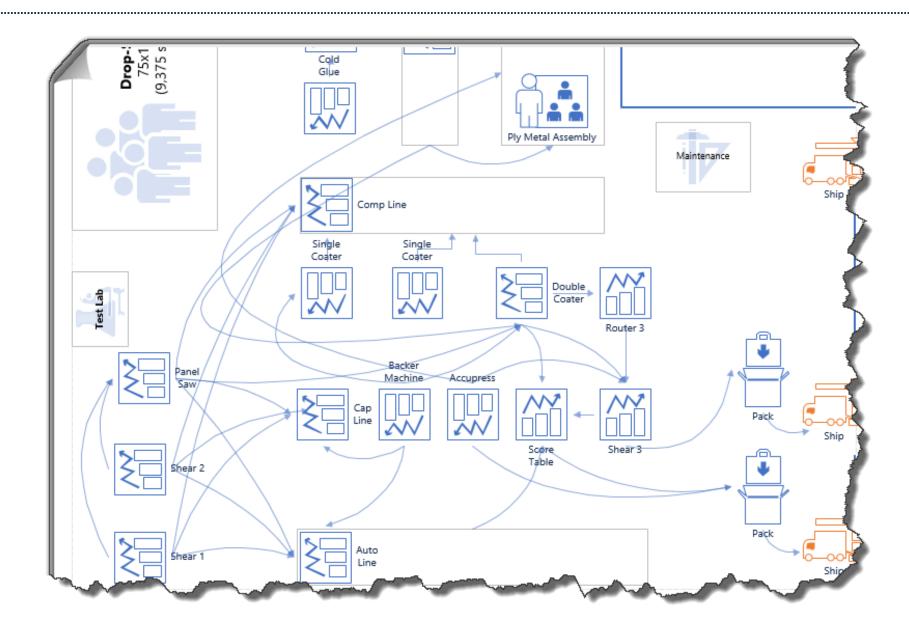


How to Plot a Spaghetti Chart

- Step 1: Create a map of the work area layout.
- Step 2: Observe the current work flow and draw the actual work path from the very beginning of work to the end when products exit the work area.
- Step 3: Analyze the spaghetti chart and identify improvement opportunities.



Spaghetti Chart Example





Thought Process Mapping

- A **thought process map** is a graphical tool to help brainstorm, organize, and visualize the information, ideas, questions, or thoughts regarding reaching the project goal.
- It's a popular tool often used at the beginning of a project in order to:
 - identify knowns and unknowns
 - communicate assumptions and risks
 - discover potential problems and solutions
 - identify resources, information, and actions required to meet the goal
 - present relationship of thoughts.



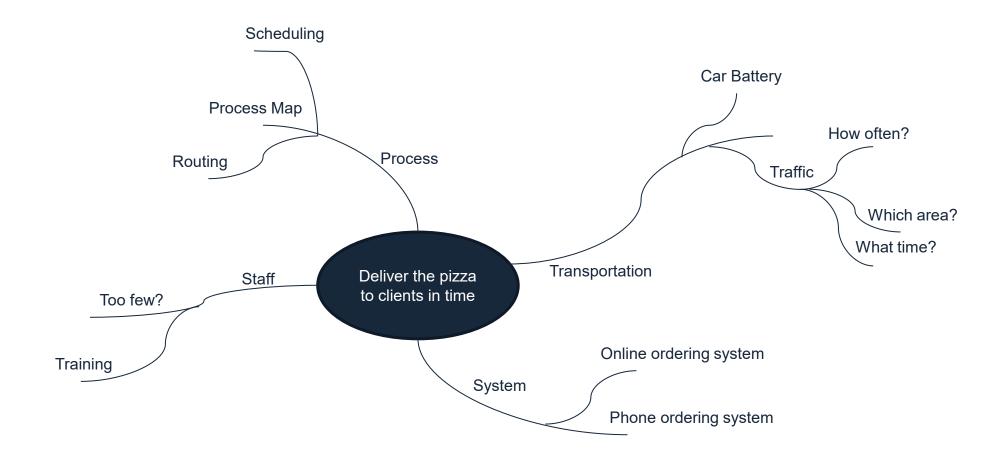
How to Plot a Thought Process Map

- Step 1: Define the project goal.
- Step 2: Brainstorm knowns and unknowns about the project.
- Step 3: Brainstorm questions and group the unknowns and questions into five phases (Define, Measure, Analyze, Improve, and Control).
- Step 4: Sequence the questions below the project goal and link the related questions.
- Step 5: Identify tools or methods that would be used to answer the questions.



Step 6: Repeat steps 3 to 5 as the project continues.

Thought Process Map Example





2.1.4 FMEA



What is FMEA?

 The FMEA (Failure Modes and Effects Analysis) is an analysis technique to identify, evaluate, and prioritize a potential deficiency in a process so that the project team can design action plans to reduce the probability of the failure/deficiency occurring.

Failure Modes & Effects Analysis - FMEA

Product or Process Step	Potential Failure Mode	Potential Failure Effects	S	Potential Causes	O	Current Controls	D	R P N	Recommended Actions	Responsible	Actions Taken	S	o d	R P N
								0						0
								0						0
								0						0
								0						0
								0						0
								0						0
								0						0

• FMEA is completed in cross-functional brainstorming sessions in which attendees have a good understanding of the entire process or of a segment of it.



Basic FMEA Terms

Process Functions

 Process steps depicted in the process map. FMEA is based on a process map and one step/function is analyzed at a time.

Failure Modes

 Potential and actual failure in the process function/step. It usually describes the way in which failure occurs. There might be more than one failure mode for one process function.

Failure Effects

 Impact of failure modes on the process or product. One failure mode might trigger multiple failure effects.

Failure Causes

 Potential defect of the design that might result in the failure modes occurring. One failure mode might have multiple potential failure causes.

Current Controls

 Procedures currently conducted to prevent failure modes from happening or to detect the failure mode occurring.



Basic FMEA Terms

Severity Score

- The seriousness of the consequences of a failure mode occurring.
- Ranges from 1 to 10, with 10 indicating the most severe consequence.

Occurrence Score

- The frequency of the failure mode occurring.
- Ranges from 1 to 10, with 10 indicating the highest frequency.

Detection Score

- How easily failure modes can be detected.
- Ranges from 1 to 10, with 10 indicating the most difficult detection.



Basic FMEA Terms

RPN (Risk Prioritization Number)

- The product of the severity, occurrence, and detection scores.
- Ranges from 1 to 1000.
- The higher RPN is, the more focus the particular step/function needs.

Recommended Actions

 The action plan recommended to reduce the probability of failure modes occurring.



How to Conduct an FMEA

- Step 1: List the critical functions of the process based on the process map created.
- Step 2: List all potential failure modes that might occur in each function.
 One function may have multiple potential failures.
- Step 3: List all potential failure effects that might affect the process or product.
- Step 4: List all possible causes that may lead to the failure mode happening.

Step 5: List the current control procedures for each failure mode.

How to Conduct an FMEA

- Step 6: Determine the severity rating for each potential failure mode.
- Step 7: Determine the occurrence rating for each potential failure cause.
- Step 8: Determine the detection rating for each current control procedure.
- Step 9: Calculate RPN (Risk Prioritization Number).
- Step 10: Rank the failures using RPN and determine the precedence of problems or critical inputs of the process. A Pareto chart might help to focus on the failure modes with high RPNs. The higher the RPN, the higher the priority the correction action plan.



How to Conduct an FMEA

- Step 11: Brainstorm and create recommended action plans for each failure mode.
- Step 12: Determine and assign the task owner and projected completion date to take actions.
- Step 13: Determine the new severity rating if the actions are taken.
- Step 14: Determine the new occurrence rating if the actions are taken.
- Step 15: Determine the new detection rating if the actions are taken.
- Step 16: Update the RPN based on new severity, occurrence, and detection ratings.



Case study:

- Joe is trying to identify, analyze, and eliminate the failure modes he
 experienced in the past when preparing his work bag before heading to
 the office every morning. He decides to run an FMEA for his process of
 work bag preparation.
- There are only two steps involved in the process.
 - Putting the work files in the bag
 - Putting a water bottle in the bag.



Step 1: List the critical functions of the process based on the process map created.

Product or Process Step	Potential Failure Mode	Potential Failure Effects
Place files in bag		
Put water bottle in bag		

Step 2: List all the potential failure modes for each function.

Product or Process Step	Potential Failure Mode	Potential Failure Effects
Place files in bag	Incorrect files put in the bag	
Put water bottle in bag	Water leaks	

Step 3: List potential failure effects that might affect the process.

Product or Process Step	Potential Failure Mode	Potential Failure Effects
Place files in bag	Incorrect files put in the bag	Work is delyed
Put water bottle in bag	Water leaks	Files in bag damaged



• Step 4: List all possible causes to the failure mode.

Potential Failure Effects	S	Potential Causes	o
Work is delyed		Files are not organized well	
Files in bag damaged		Cap on water bottle not tight	

Step 5: List any control procedures for each failure mode.

Potential Causes	O	Current Controls	D
Files are not organized well		Check if files are needed	
Cap on water bottle not tight		Check bottle cap before inserting	

Step 6: Determine the severity rating for each failure mode.

Potential Failure Effects	S	Potential Causes	o	Current Controls
Work is delyed	9	Files are not organized well		Check if files are needed
Files in bag damaged	7	Cap on water bottle not tight		Check bottle cap before inserting



• Step 7: Determine the occurrence rating for each failure cause.

Potential Failure Effects	S	Potential Causes	o	Current Controls
Work is delyed	9	Files are not organized well	3	Check if files are needed
Files in bag damaged	7	Cap on water bottle not tight	5	Check bottle cap before inserting

Step 8: Determine the detection rating for each control.

S	Potential Causes	o	O Current Controls		R P N
9	Files are not organized well	3	Check if files are needed	5	135
7	Cap on water bottle not tight	5	Check bottle cap before inserting	5	175

• Step 9: Calculate the RPN (Risk Prioritization Number).

S	Potential Causes	o	Current Controls	D	R P N
9	Files are not organized well	3	Check if files are needed	5	135
7	Cap on water bottle not tight	5	Check bottle cap before inserting	5	175



 Step 10: Rank the failures using the RPN and determine the precedence of problems or critical inputs of process.

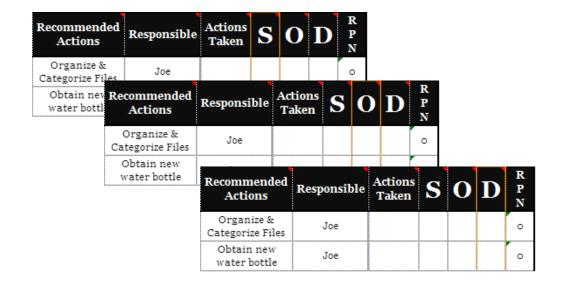
Current Controls	D	R P N	Recommended Actions
Check bottle cap before inserting	5	175	Organize & Categorize Files
Check if files are needed	5	135	Obtain new water bottle

- Step 11: Brainstorm and create recommended action plans.
- Step 12: Determine and assign owners with completion dates.

D	R P N	Recommended Actions	Responsible
5	175	Organize & Categorize Files	Joe
5	135	Obtain new water bottle	Joe



• Steps 13-15: Determine new severity, occurrence and detection ratings if actions are taken.



Step 16: Update RPN based on new ratings.

Recommended Actions	Responsible	Actions Taken	S	o d	R P N
Organize & Categorize Files	Joe				0
Obtain new water bottle	Joe				0



2.1.5 Theory of Constraints



What is the Theory of Constraints?

- Processes, systems, and organizations are vulnerable to their weakest part.
- Any manageable system is limited by constraints in its ability to produce more (and there is always at least one constraint).





Performance Measures

Making sound financial decisions based on these three measures is a critical requirement.

- Throughput rate at which a system generates money through sales.
- Operational Expense money spent by the system to turn inventory into throughput.
- Inventory money the system has invested in purchasing things it intends to sell.





Five Focusing Steps

Objective: To ensure ongoing improvement efforts are focused on the constraints of a system.



- 1. Identify the system's constraint(s).
- 2. Decide how to exploit the constraint(s).
- 3. Subordinate everything else to the decision in step 2.
- 4. Elevate the constraint(s).
- 5. If in previous steps a constraint has been broken, return to step 1, but do not allow inertia to cause a system's constraint.



Logical Thinking Processes

	Focusing Step	Thinking Process	Tools				
1	Identify the system's constraint(s)	Identify the problemsFind the root causes	Cause and effect diagram				
2	Decide how to exploit the constraint(s)	Develop a solution	Future reality tree				
3	Subordinate everything else to the decision in step 2	Identify the conflict preventing the solutionRemove the conflict	Evaporating cloud				
4	Elevate the constraint	Construct and execute an implementation plan	Prerequisite tree Transition tree				
5	· · · · · · · · · · · · · · · · · · ·	If in previous steps a constraint has been broken, return to step 1, but do not allow inertia to cause a system's constraint					



Simulation Exercise

Resources needed:

- 3 "production line" participants
- 1 timer per each "production line" participant
- 5 small boxes of 15 widgets each (paperclips, pens/pencils, candy, etc.)

Widget Value Chain:

Get box of widgets and empty it

Participant 1

Count widgets and fill box

Participant 2

Return box to instructor

Participant 3



Simulation Exercise

Widget Value Chain:

Get box of widgets and empty it

Participant 1

Count widgets and fill box

Participant 2

Return box to instructor

Participant 3

Five focusing steps:

- 1. Identify the system's constraint(s).
- 2. Decide how to exploit the constraint(s). Quick improvements. Use what you have.
- 3. Subordinate everything else to the decision in step 2.
- 4. Elevate the constraint(s). If more improvement is needed, what actions are required?
- 5. If in previous steps a constraint has been broken, return to step 1, but do not allow inertia to cause a system's constraint.



2.2 Six Sigma Statistics



Black Belt Training: Measure Phase

2.1 Process Definition

- 2.1.1 Cause and Effect Diagrams
- 2.1.2 Cause and Effects Matrix
- 2.1.3 Process Mapping
- 2.1.4 FMEA: Failure Modes & Effects Analysis
- 2.1.5 Theory of Constraints

2.2 Six Sigma Statistics

- 2.2.1 Basic Statistics
- 2.2.2 Descriptive Statistics
- 2.2.3 Distributions and Normality
- 2.2.4 Graphical Analysis

2.3 Measurement System Analysis

- 2.3.1 Precision and Accuracy
- 2.3.2 Bias, Linearity, and Stability
- 2.3.3 Gage R&R
- 2.3.4 Variable and Attribute MSA

2.4 Process Capability

- 2.4.1 Capability Analysis
- 2.4.2 Concept of Stability
- 2.4.3 Attribute and Discrete Capability
- 2.4.4 Monitoring Techniques



2.2.1 Basic Statistics



What is Statistics?

- **Statistics** is the science of collection, analysis, interpretation, and presentation of data.
- In Six Sigma, we apply statistical methods and principles to quantitatively measure and analyze the process performance to reach statistical conclusions and help solve business problems.



Types of Statistics

- Descriptive Statistics
 - Describing what was going on
- Inferential Statistics
 - Making inferences from the data at hand to more general conditions



Descriptive Statistics

- **Descriptive statistics** is applied to describe the main characteristics of a collection of data.
- Descriptive statistics summarizes the features of the data quantitatively.
- Descriptive statistics is descriptive only and it does not make any generalizations beyond the data at hand.
- The data used for descriptive statistics are for the purpose of representing or reporting.



Inferential Statistics

- Inferential statistics is applied to infer the characteristics or relationships of the populations from which the data are collected.
- Inferential statistics draws statistical conclusions about the population by analyzing the sample data subject to random variation.
- A complete data analysis includes both descriptive statistics and inferential statistics.



Statistics vs. Parameters

- The word *statistic* refers to a numeric measurement calculated using a sample data set, for example, sample mean or sample standard deviation. Its plural is *statistics* (the same spelling as "statistics" which refers to the scientific discipline).
- The *parameter* refers to a numeric metric describing the population, for example, population mean and population standard deviation. Unless you have the full data set of the population, you will not be able to know the population parameters.



Continuous Variable vs. Discrete Variable

- Continuous Variable
 - Measured
 - There is an infinite number of values possible
 - Examples: temperature, height, weight, money, time
- Discrete Variable
 - Counted
 - There is a finite number of values available
 - Examples: count of people, count of countries, count of defects, count of defectives



Types of Data

- Nominal
 - Categorical data
 - Examples: a set of colors, the social security number
- Ordinal
 - Rank-ordering data
 - Examples: the first, second place in a race, scores of exams
- Interval
 - Equidistant data
 - Examples: temperature with Fahrenheit or Celsius scale
- Ratio
 - The ratio between the magnitude of a continuous value and the unit value of the same category
 - Examples: weight, length, time



2.2.2 Descriptive Statistics



Basics of Descriptive Statistics

- Descriptive statistics provides a quantitative summary for the data collected.
- It summarizes the main features of the collection of data.
 - Shape
 - Location
 - Spread
- It is a presentation of data collected and it does *not* provide any inferences about a more general condition.



Shape of the Data

- **Distribution** is used to describe the shape of the data.
- Distribution (also called frequency distribution) summarizes the frequency of an individual value or a range of values of a variable (either continuous or discrete).
- Distribution is depicted as a table or graph.



Shape of the Data

- Simple example of distribution
 - We are tossing a fair die. The possible value we obtain from each tossing is a value between 1 and 6.
 - Each value between 1 and 6 has a 1/6 chance to be hit for each tossing.
 - The distribution of this game describes the relationship between every possible value and the percentage of times the value is being hit (or count of times the value is being hit).



Shape of the Data

- Examples of continuous distribution
 - Normal Distribution
 - T distribution
 - Chi-square distribution
 - F distribution
- Examples of discrete distribution
 - Binomial distribution
 - Poisson distribution



Location of the Data

- The **location** (i.e. central tendency) of the data describes the value where the data tend to cluster around.
- There are multiple measurements to capture the location of the data:
 - Mean
 - Median
 - Mode.



Mean

• The **mean** is the arithmetic average of a data set.

$$\frac{-}{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$$

n is the number of values in the data set

• For example, we have a set of data: 2, 3, 5, 8, 5, and 9. The arithmetic mean of the data set is

$$\frac{2+3+5+8+5+9}{6} = 5.33$$



Median

- The median is the middle value of the data set in numeric order.
- It separates the finite set of data into two parts: one with values higher that the median and the other with values lower than the median.
- For example, we have a set of data: 45, 32, 67, 12, 37, 54, and 28. The median is 37 since it is the middle value of the sorted list of values (i.e., 12, 28, 32, 37, 45, 54, and 67).



Mode

- The mode is the value that occurs most often in the data set.
- If no number is repeated, there is no mode for the data set.
- For example, we have a data set: 55, 23, 45, 45, 68, 34, 45, 55. The mode is 45 since it occurs most frequently.



Spread of the Data

- The spread (i.e. variation) of the data describes the degree of data dispersing around the center value.
- There are multiple measurements to capture the spread of the data:
 - Range
 - Variance
 - Standard Deviation.



Range

- The **range** is the numeric difference between the greatest and smallest values in a data set.
- Only two data values (i.e., the greatest and the smallest values) are accounted for calculating the range.
- For example, we have a set of data: 34, 45, 23, 12, 32, 78, and 23. The range of the data is 78–12 = 66.



Variance

- The **variance** measures how far on average the data points spread out from the mean.
- It is the average squared deviation of each value from its mean.
- It is the square of the sample standard deviation.
- All the data points are accounted for calculating the variance.

$$s^{2} = \frac{1}{n} \sum_{i=1}^{n} (x_{i} - \overline{x})^{2}$$

where

n is the number of values in the data set

$$\overline{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$$



Standard Deviation

- Standard deviation describes how far the data points spread away from the mean.
- It has the same measurement units as the data itself.
- It is simply the square root of the variance.
- All the data points are accounted for calculating the standard deviation.

$$S = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (x_i - \bar{x})^2}$$

where

n is the number of values in the data set

$$\overline{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$$



2.2.3 Normal Distribution & Normality



What is Normal Distribution?

- The **normal distribution** is a probability distribution of a continuous random variable whose values spread symmetrically around the mean.
- A normal distribution can be completely described by using its mean (μ) and variance (σ^2) .
- When a variable x is normally distributed, we denote $x \sim N(\mu, \sigma^2)$.



Z Distribution

- The **Z** distribution is the simplest normal distribution with the mean equal to zero and the variance equal to one.
- Any normal distribution can be transferred to a Z distribution by applying

$$z = \frac{x - \mu}{\sigma}$$

where

$$x \sim N(\mu, \sigma^2)$$
 $\sigma \neq 0$



Z Score

- The **Z Score** is the measure of how many standard deviations an observation is above or below the mean.
- Positive Z Scores indicate the observation is above the mean or "right of the mean".
- Negative Z Scores indicate the observation is below the mean of "left of the mean"
- Calculate Z Score using the formula below:

$$z = \frac{x - \mu}{\sigma}$$

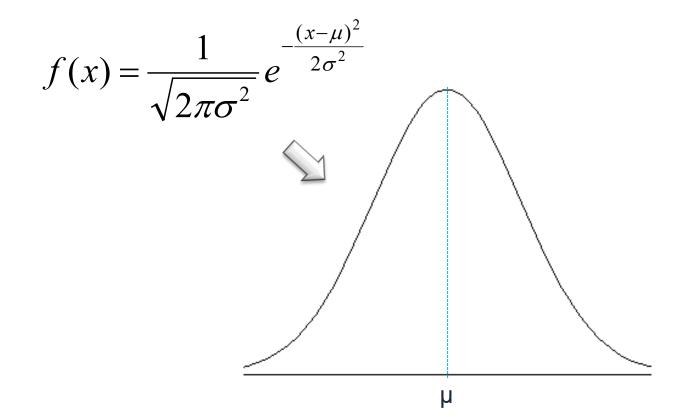
where

x is the observation y is the mean of the population z is the standard deviation of the population



Shape of Normal Distribution

- The probability density function curve of normal distribution is bellshaped.
- Probability density function of normal distribution





Location of Normal Distribution

- If a variable is normally distributed, the mean, the median, and the mode have the same value.
- The probability density curve of normal distribution is symmetric around a center value which is the mean, the median, and the mode at the same time.



Spread of Normal Distribution

- The spread or variation of the normally-distributed data can be described using the variance or the standard deviation.
- The smaller the variance or the standard deviation, the less variability in the data set.

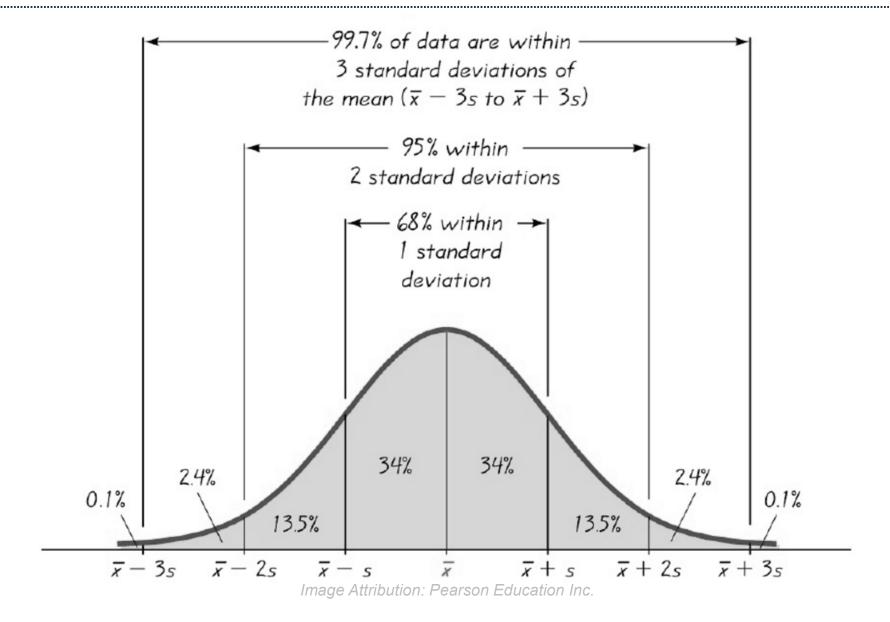


Empirical Rule

- The 68-95-99.7 rule or the *empirical rule in statistics* states that for a normal distribution:
 - About 68% of the data fall within one standard deviation of the mean
 - About 95% of the data fall within two standard deviations of the mean
 - About 99.7% of the data fall within three standard deviations of the mean.



Empirical Rule (68-95-99.7 Rule)





Normality

- Not all the distributions with a bell shape are normal distributions.
- To check whether a group of data points are normally distributed, we need to run a normality test.
- There are different normality tests available:
 - Anderson-Darling test
 - Sharpiro-Wilk test
 - Jarque-Bera test.
- More details of normality test will be introduced in the Analyze module.



Normality Testing

- To check whether the population of our interest is normally distributed, we need to run normality test.
 - Null Hypothesis (H₀): The data points *are* normally distributed.
 - Alternative Hypothesis (H_a): The data points are *not* normally distributed.
- There are many normality tests available. For example, Anderson-Darling test, Sharpiro-Wilk test, Jarque-Bera test, and so on.



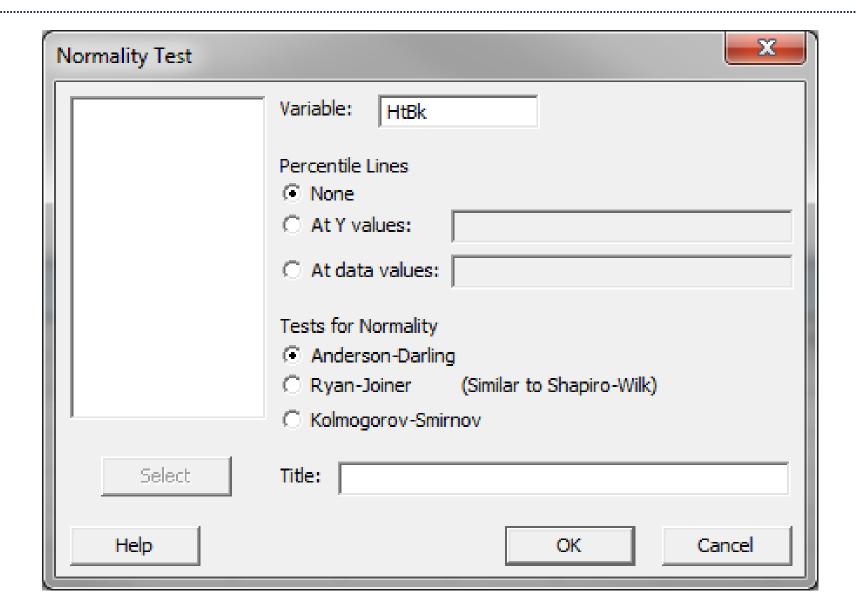


- Case study: we are interested to know whether the height of basketball players is normally distributed.
 - Data File: "One Sample T-Test" tab in "Sample Data.xlsx"
- Null Hypothesis (H0): the height of basketball players is normally distributed.
- Alternative Hypothesis (Ha): the height of basketball players is not normally distributed.



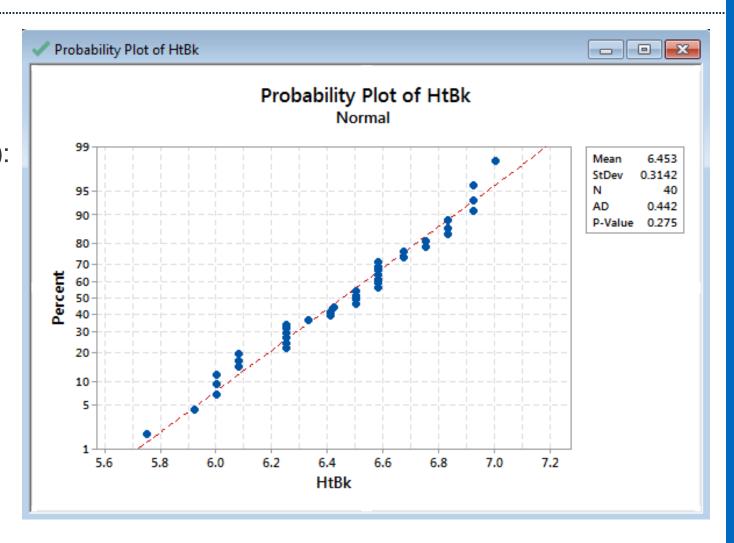
- Steps to run a normality test in Minitab
 - 1) Click Stat → Basic Statistics → Normality Test.
 - 2) A new window named "Normality Test" pops up.
 - 3) Select "HtBk" as the "Variable."
 - 4) Click "OK."
 - 5) The normality test results appear in the new window.







- Null Hypothesis (H₀): The data are normally distributed.
- Alternative Hypothesis (H_a): The data are not normally distributed.
- Since the p-value of the normality is 0.275 greater than alpha level (0.05), we fail to reject the null and claim that the data are normally distributed.





2.2.4 Graphical Analysis



What is Graphical Analysis?

- In statistics, **graphical analysis** is a method to visualize the quantitative data.
- Graphical analysis is used to discover the structure and patterns in the data, explaining and presenting the statistical conclusions.
- A complete statistical analysis includes both quantitative analysis and graphical analysis.



Graphical Analysis Example

- There are various graphical analysis tools available. Here are four most commonly used examples:
 - Box Plot
 - Histogram
 - Scatter Plot
 - Run Chart.

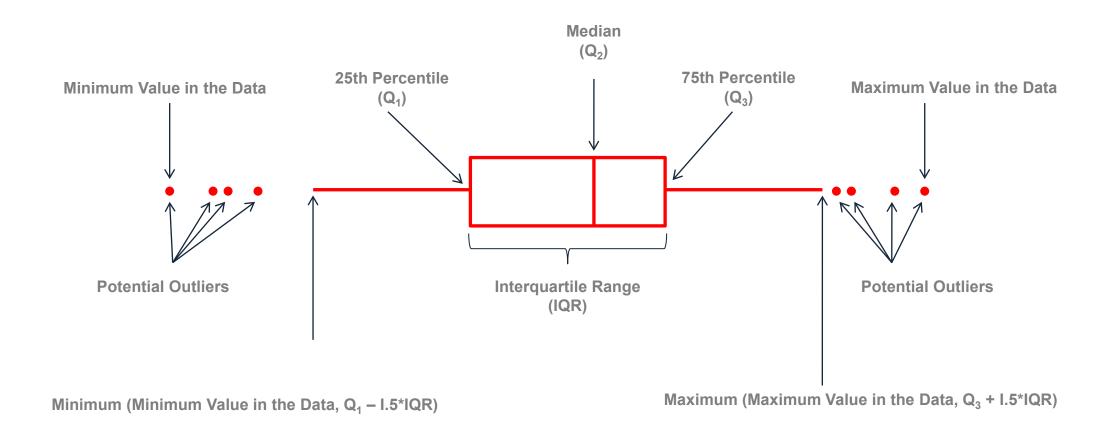


Box Plot

- A **box plot** is a graphical method to summarize a data set by visualizing the minimum value, 25th percentile, median, 75th percentile, the maximum value, and potential outliers.
- A percentile is the value below which a certain percentage of data fall. For example, if 75% of the observations have values lower than 685 in a data set, then 685 is the 75th percentile of the data.



Box Plot



Interquartile Range = 75th Percentile – 25th Percentile

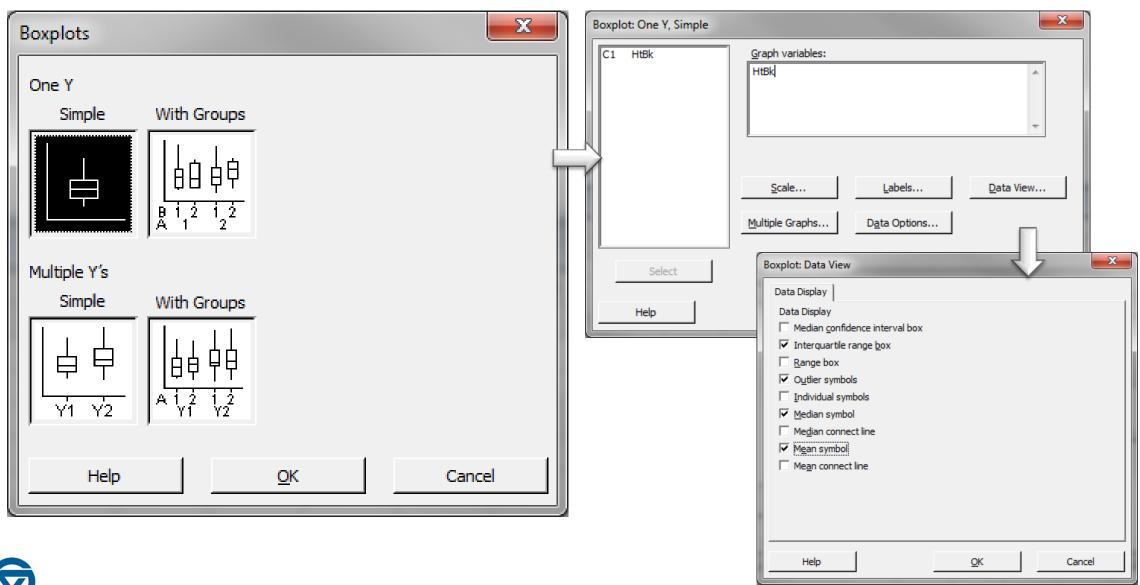


How to Use Minitab to Generate a Box Plot

- Data File: "Box Plot" tab in "Sample Data.xlsx"
- Steps to render a Box Plot in Minitab
 - 1) Click Graph → Boxplot.
 - 2) A new window named "Boxplots" pops up.
 - 3) Click "OK" in the window "Boxplots."
 - 4) Another new window named "Boxplot One Y, Simple" pops up.
 - 5) Select "HtBk" as the "Graph Variables."
 - 6) Click the "Data View" button and a new window named "Boxplot Data View" pops up.
 - 7) Check the boxes "Median symbol" and "Mean symbol."
 - 8) Click "OK" in the window "Boxplot Data View."
 - Click "OK" in the window "Boxplot One Y, Simple."
 - 10) The box plot appears automatically in the new window.

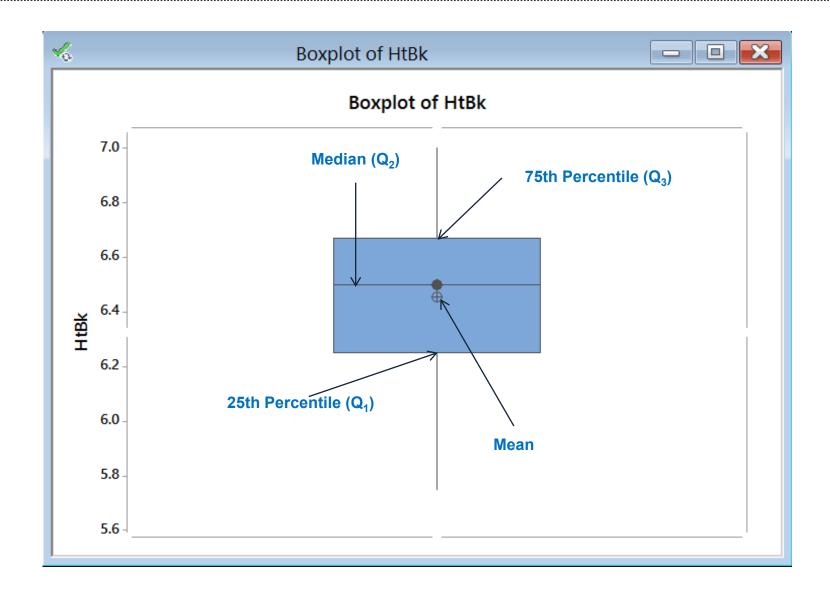


How to Use Minitab to Generate a Box Plot





How to Use Minitab to Generate a Box Plot





Histogram

- A histogram is a graphical tool to present the distribution of the data.
- The X axis represents the possible values of the variable and the Y axis represents the frequency of the value occurring.
- A histogram consists of adjacent rectangles erected over intervals with heights equal to the frequency density of the interval.
- The total area of all the rectangles in a histogram is the number of data values.

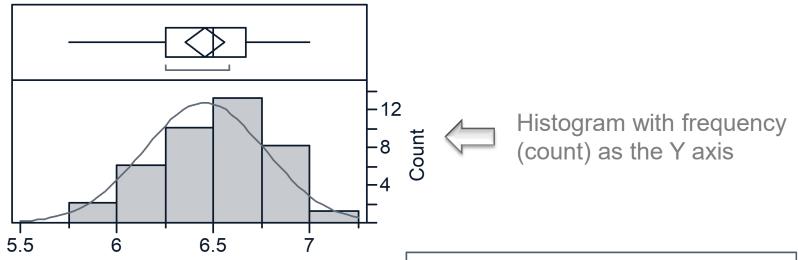


Histogram

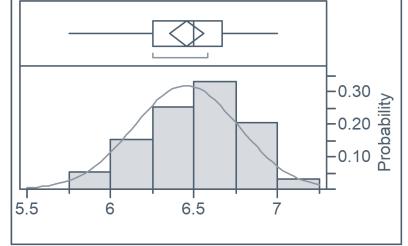
- A histogram can also be normalized. In this case, the X axis still represents the possible values of the variable, but the Y axis represents the percentage of observations that fall into each interval on the X axis.
- The total area of all the rectangles in a normalized histogram is 1.
- With the histogram, we have a better understanding of the shape, location, and spread of the data.



Histogram



Normalized histogram with proportion (probability) as the Y axis



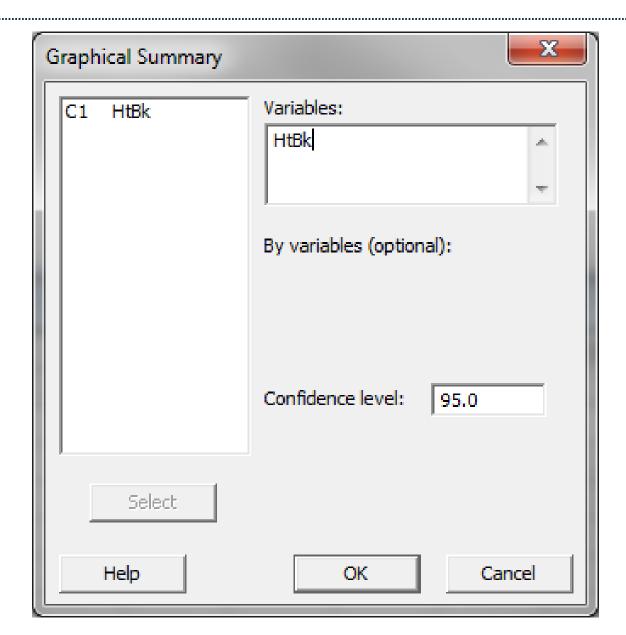


How to Use Minitab to Generate a Histogram

- Data File: "Histogram" tab in "Sample Data.xlsx"
- Steps to render a histogram in Minitab
 - 1) Click Stat → Basic Statistics → Graphical Summary.
 - 2) A new window named "Graphical Summary" pops up.
 - 3) Select "HtBk" as the "Variables."
 - 4) Click "OK."
 - 5) The histogram appears in the new window.

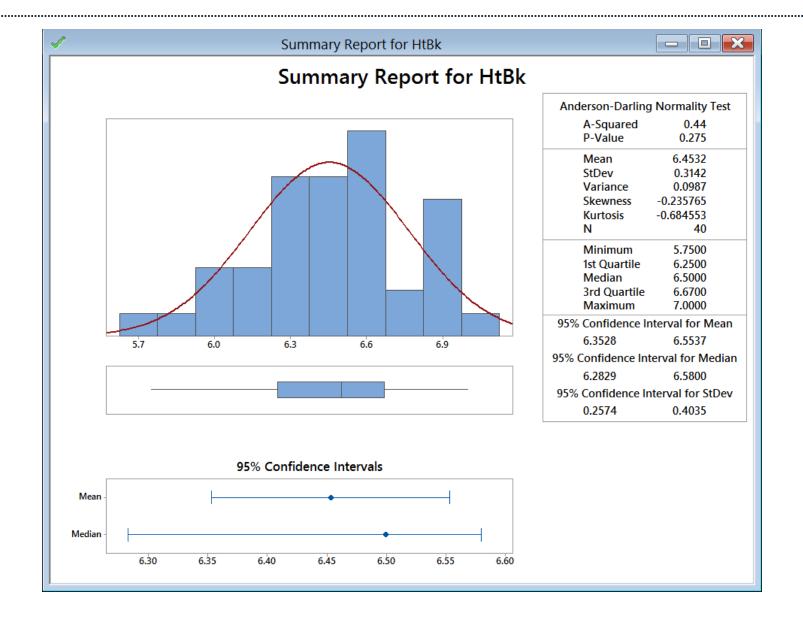


How to Use Minitab to Generate a Histogram





How to Use Minitab to Generate a Histogram





Scatter Plot

- A **scatter plot** is a diagram to present the relationship between two variables of a data set.
- A scatter plot consists of a set of data points.
- On the scatter plot, a single observation is presented by a data point with its horizontal position equal to the value of one variable and its vertical position equal to the value of the other variable.



Scatter Plot

- A scatter plot helps to understand:
 - Whether the two variables are related to each other or not
 - How is the strength of their relationship
 - What is the shape of their relationship
 - What is the direction of their relationship
 - Whether outliers are present.

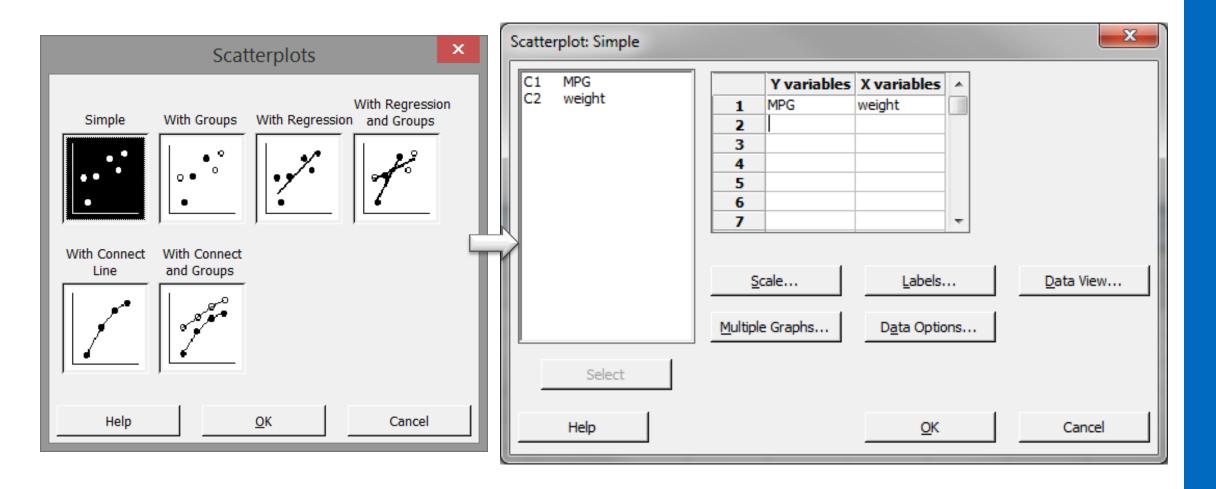


How to Use Minitab to Generate a Scatter Plot

- Data File: "Scatter Plot" tab in "Sample Data.xlsx"
- Steps to render a histogram in Minitab:
 - 1) Click Graph → Scatterplot.
 - 2) A new window named "Scatterplots" pops up.
 - 3) Click "OK" in the window "Scatterplots" and another window named "Scatterplot Simple" pops up.
 - 4) Select "MPG" as the "Y variables."
 - 5) Select "weight" as the "X variables."
 - 6) Click "OK."
 - 7) The scatter plot appears in the new window.

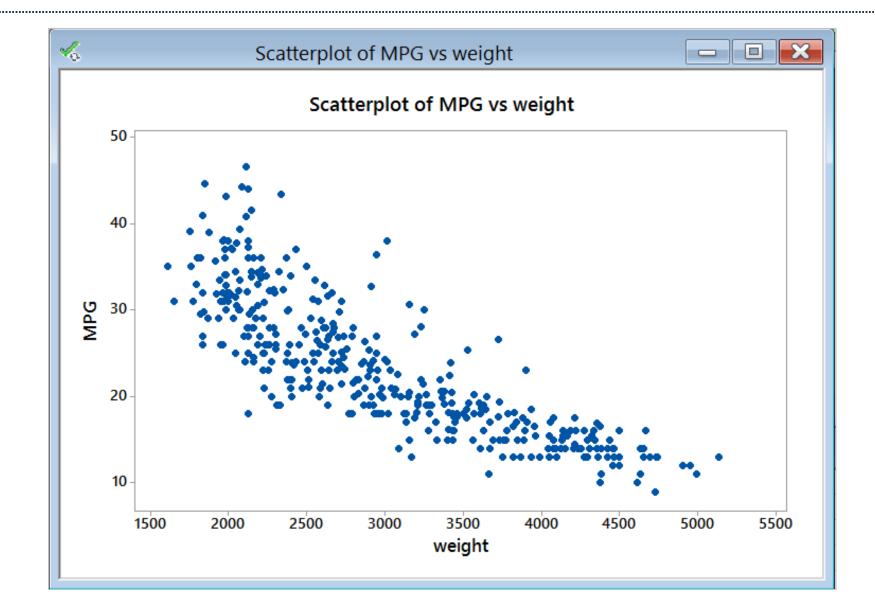


How to Use Minitab to Generate a Scatter Plot





How to Use Minitab to Generate a Scatter Plot





Run Chart

- A **run chart** is a chart used to present the data in time order. It captures the process performance over time.
- The X axis of the run chart indicates the time and the Y axis indicates the observed values.
- Run chart looks similar to control charts except that a run chart does not have control limits plotted. It is easier to produce a run chart than a control chart.
- It is often used to identify the anomalies in the data and discover the pattern of data changing over time.

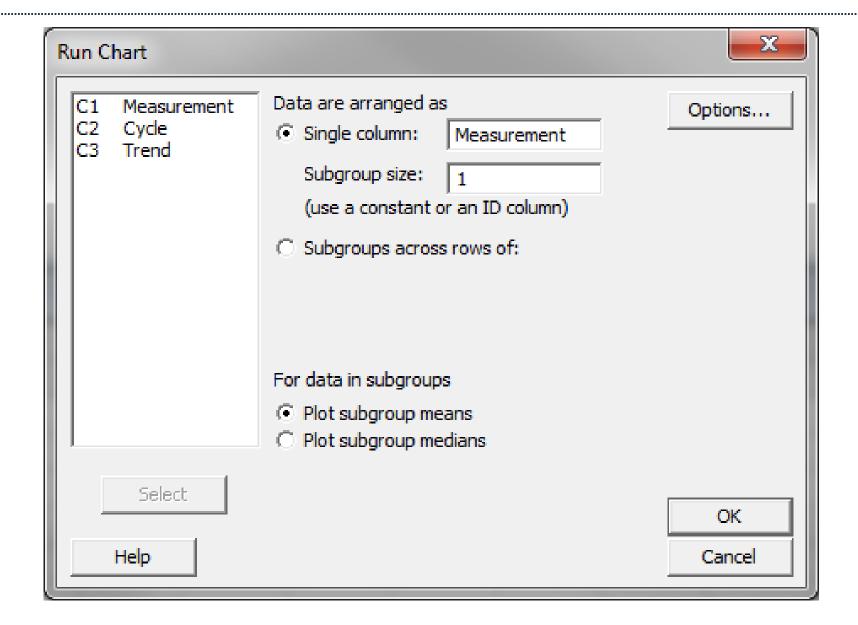


How to Plot a Run Chart in Minitab

- Steps to plot a run chart in Minitab:
 - 1) Data File: "Run Chart" tab in "Sample Data.xlsx."
 - 2) Click Stat → Quality Tools → Run Chart.
 - 3) A new window named "Run Chart" pops up.
 - 4) Select "Measurement" as the "Single Column."
 - 5) Enter "1" as the "Subgroup Size."
 - 6) Click "OK."
 - 7) The run chart appears automatically in the new window.



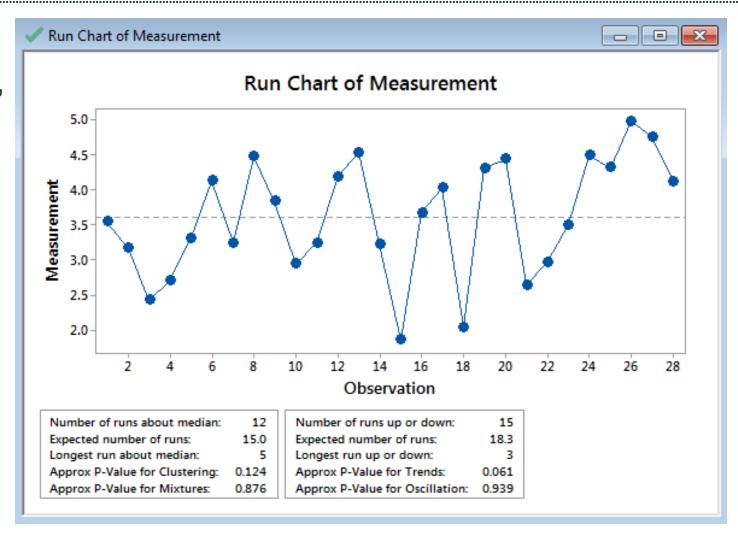
How to Plot a Run Chart in Minitab





How to Plot a Run Chart in Minitab

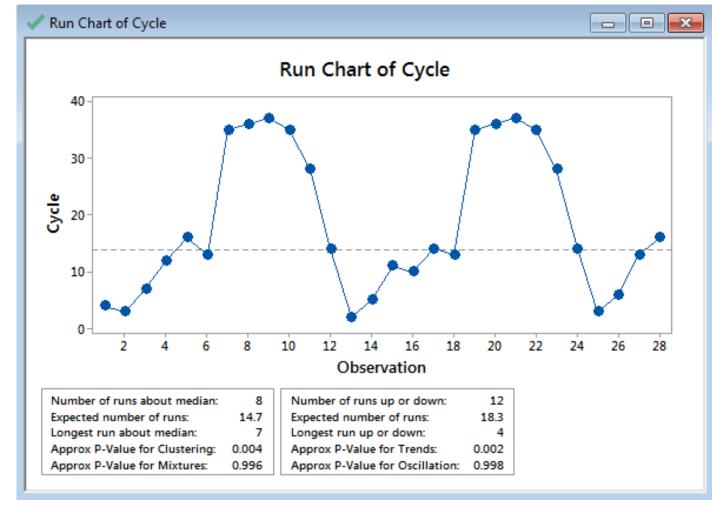
- Run charts are used to identify the trend, cycle, seasonal pattern, or abnormality in the data.
 - The time series in this chart appears stable.
 - There are no extreme outliers, trending or seasonal patterns.





Run Chart Example

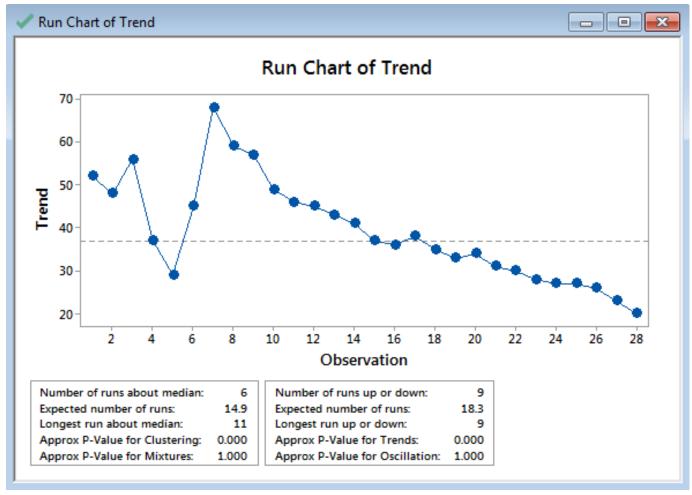
- Following the previous "Run Chart" instructions, create a new chart using the data listed in the column "Cycle" in the "Run Chart" tab of "Sample Data.xlsx."
- Notice the repeating pattern of the data. This could suggest something cyclical or seasonal.





Run Chart Example

- Create another run chart using the data listed in column "Trend" in the "Run Chart" tab of "Sample Data.xlsx."
- Notice the trending pattern represented by the data points successively declining.





2.3 MSA (Measurement System Analysis)



Black Belt Training: Measure Phase

2.1 Process Definition

- 2.1.1 Cause and Effect Diagrams
- 2.1.2 Cause and Effects Matrix
- 2.1.3 Process Mapping
- 2.1.4 FMEA: Failure Modes & Effects Analysis
- 2.1.5 Theory of Constraints

2.2 Six Sigma Statistics

- 2.2.1 Basic Statistics
- 2.2.2 Descriptive Statistics
- 2.2.3 Distributions and Normality
- 2.2.4 Graphical Analysis

2.3 Measurement System Analysis

- 2.3.1 Precision and Accuracy
- 2.3.2 Bias, Linearity, and Stability
- 2.3.3 Gage R&R
- 2.3.4 Variable and Attribute MSA

2.4 Process Capability

- 2.4.1 Capability Analysis
- 2.4.2 Concept of Stability
- 2.4.3 Attribute and Discrete Capability
- 2.4.4 Monitoring Techniques



2.3.1 Precision and Accuracy



What is Measurement System Analysis

- Measurement System Analysis (MSA) is a systematic method to identify and analyze the variation components in the measurement.
- It is a mandatory step in any Six Sigma project to ensure the data are reliable before making any data-based decisions.
- The MSA is the check point of data quality before we start any further analysis and draw any conclusions from the data.



Data-Based Analysis

- Here are some examples of data-based analysis where MSA is the prerequisite:
 - Correlation analysis
 - Regression analysis
 - Hypothesis testing
 - Analysis of variance
 - Design of experiments
 - Multivariate analysis
 - Statistical process control.



Measurement System

A measurement system is a process to obtain data.

Inputs (Xs) Measurement System Output (Y)

- Y (output of the measurement system)
 - Observed values
- X's (inputs of the measurement system)
 - True values
 - Measurement errors



Observed Value = True Value + Measurement Error

True Value

- The actual value we are interested to measure
- It reflects the true performance of the process we are measuring

Measurement Error

The errors brought in by measurement system

Observed Value

The observed/measured value obtained by the measurement system



- Types of Observed Values:
 - Continuous measurements
 - Weight
 - Height
 - Money
 - Discrete measurements
 - Red/Yellow/Green
 - Yes/No
 - Ratings of 1–10
- A variable MSA is designed for continuous measurements and an attribute MSA is for discrete measurements.



- Sources of measurement errors:
 - Human
 - Environment
 - Equipment
 - Sample
 - Process
 - Material
 - Method.
- Fishbone diagrams can help to brainstorm the potential factors affecting the measurement system.



- The more errors the measurement system brings in, the less reliable the observed values are.
- A valid measurement system brings in minimum amount of measurement errors.
- The goal of MSA is to qualify the measurement system by quantitatively analyzing its characteristics.



Characteristics of a Measurement System

- Any measurement systems can be characterized by two aspects:
 - Accuracy (location related)
 - Precision (variation related).
- A valid measurement system is both accurate and precise.
 - Being accurate does not guarantee the measurement system is precise.
 - Being precise does not guarantee the measurement system is accurate.



Accuracy vs. Precision

Accuracy:

- The level of closeness between the average observed value and the true value
- How well the observed value reflects the true value.

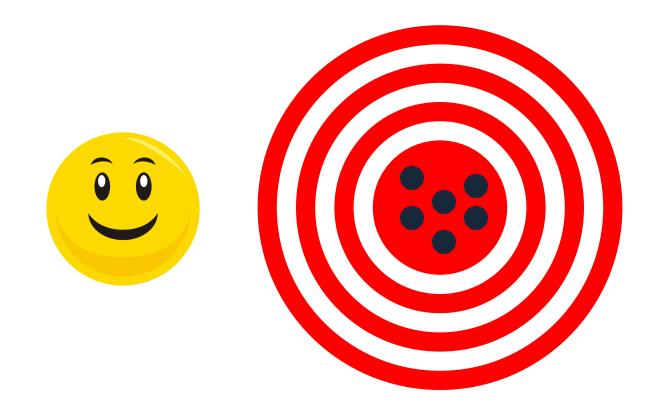
Precision:

- The spread of measurement values
- How consistent the repeated measurements deliver the same values under the same circumstances.



Accuracy vs. Precision

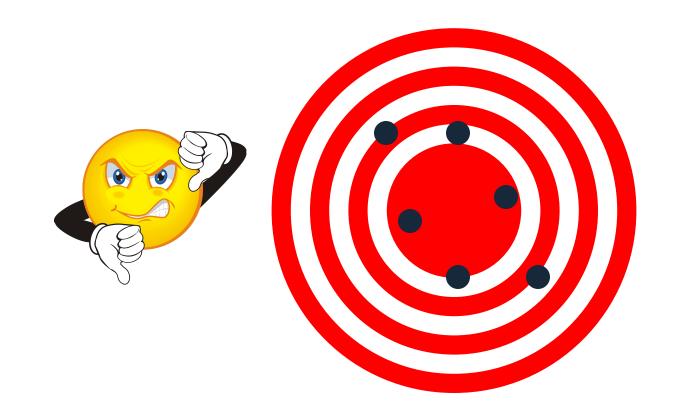
- Accurate and precise
 - high accuracy and high precision





Accuracy vs. Precision

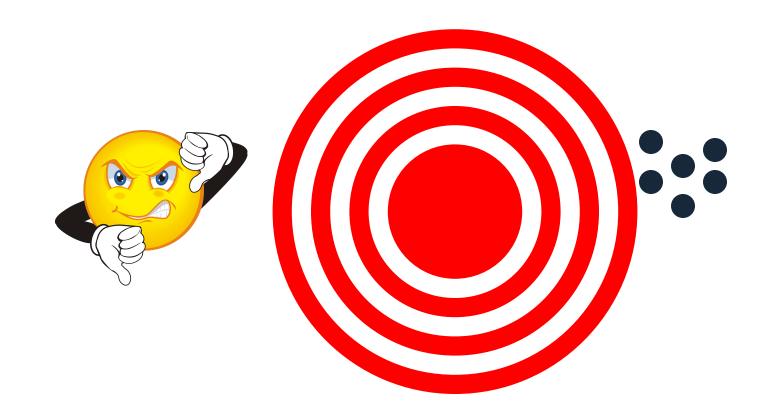
- Accurate and not precise
 - high accuracy and low precision





Accuracy vs. Precision

- Precise and not accurate
 - high precision and low accuracy





Accuracy vs. Precision

- Not accurate and not precise
 - low accuracy and low precision





MSA Conclusions

- If the measurement system is considered both accurate and precise, we can start the data-based analysis or decision making.
- If the measurement system is either not accurate or not precise, we need to identify the factor(s) affecting it and calibrate the measurement system until it is both accurate and precise.



Stratification of Accuracy and Precision

- Accuracy
 - Bias
 - Linearity
 - Stability
- Precision
 - Repeatability
 - Reproducibility



2.3.2 Bias, Linearity and Stability

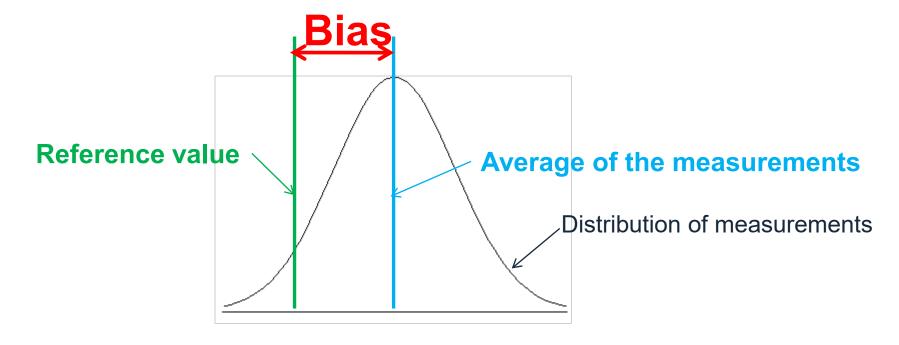


- **Bias** is the difference between the observed value and the true value of a parameter or metric being measured.
- It is calculated by subtracting the reference value from the average value of the measurements.

Bias = Grand Mean - Reference Value

where the reference value is a standard agreed upon





- The closer the average of all measurements is to the reference level, the smaller the bias.
- The reference level is the average of measurements of the same items using the master or standard instrument.



• To determine whether the difference between the average of observed measurement and the reference value is statistically significant (we will explain more details about statistical significance in the Analyze module), we can either conduct a *hypothesis testing* or *compare* the reference value against the confidence intervals of the average measurements.

• If the reference value falls into the confidence intervals, the bias is not statistically significant and can be ignored. Otherwise, the bias is statistically significant and must be fixed.

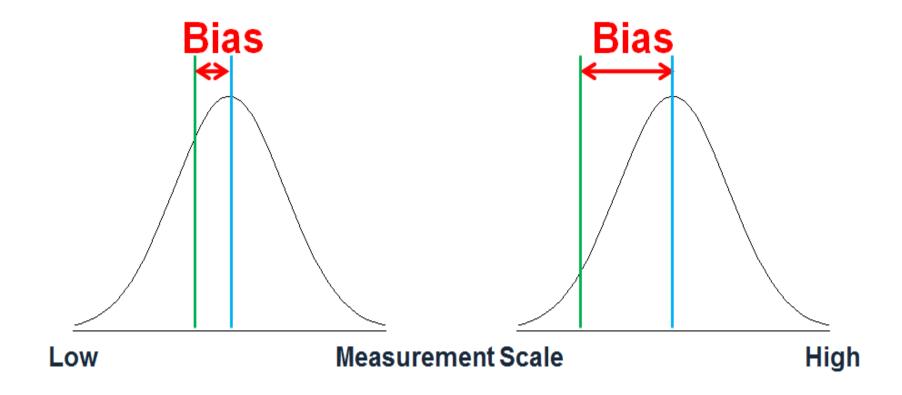


- Potential causes of bias:
 - Errors in measuring the reference value
 - Lack of proper training for appraisers
 - Damaged equipment or instrument
 - Measurement instrument not calibrated precisely
 - Appraisers read the data incorrectly.



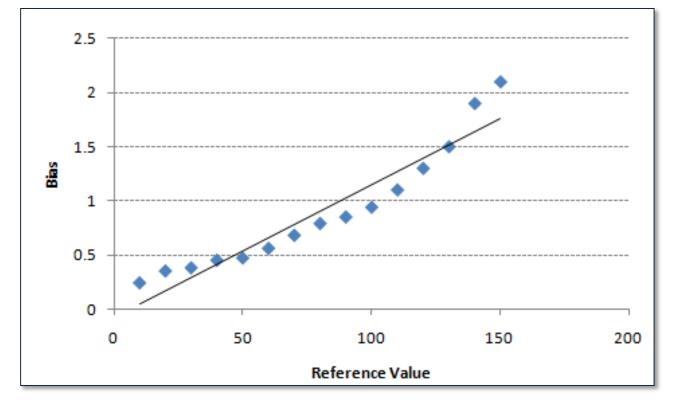
- **Linearity** is the degree of the consistency of bias over the entire expected measurement range.
- It quantifies how the bias changes over the range of measurement.
- For example, a scale is off by 0.01 pounds when measuring an object of 10 pounds. However, it is off by 10 pounds when measuring an object of 100 pounds. The scale's bias is not linear.







- Create a scatter plot for bias (Y-axis) and reference level (X-axis).
- Find a best fit linear regression line and compute the slope of the line.
- The closer the slope is to zero, the better the measurement system performs.





Formula of the linearity of a measurement system:

Linearity = |Slope| × Process Variation

where

$$Slope = \frac{\sum_{i=1}^{n} x_{i} y_{i} - \frac{1}{n} \sum_{i=1}^{n} x_{i} \sum_{i=1}^{n} y_{i}}{\sum_{i=1}^{n} (x_{i}^{2}) - \frac{1}{n} \left(\sum_{i=1}^{n} x_{i}\right)^{2}}$$

 x_i is the reference value; y_i is the bias at each reference level; n is the sample size.



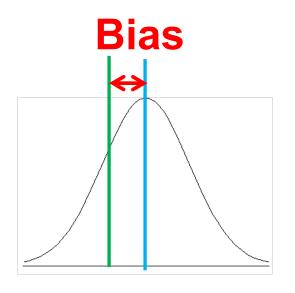
- Potential causes of linearity:
 - Errors in measuring the lower end or higher end of the reference value
 - Lack of proper training for appraisers
 - Damaged equipment or instrument
 - Measurement instrument not calibrated correctly at the lower or higher end of the measurement scale
 - Innate nature of the instrument.

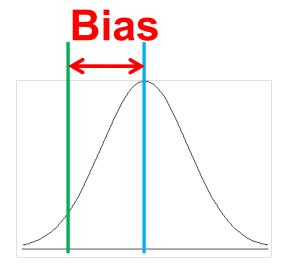


- **Stability** is the consistency level needed to obtain the same values when measuring the same objects over an extended period of time.
- A measurement system that has low bias and linearity close to zero but cannot consistently perform well would not deliver reliable data.
- Stability is evaluated using control charts.



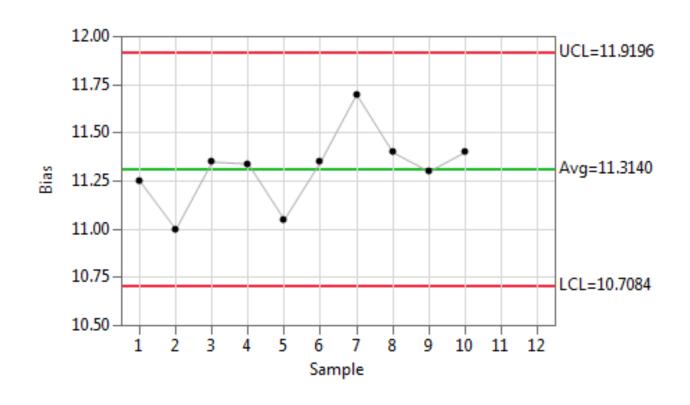
Time







- Control charts are used to evaluate the stability of a measurement system.
- When there are no data points out of control, the measurement system is considered stable.





- Potential causes of instability:
 - Inconsistent training for appraisers
 - Damaged equipment or instrument
 - Worn equipment or instrument
 - Measurement instrument not calibrated
 - Appraisers do not follow the procedure consistently.



2.3.3 Gage R&R



Repeatability

- Repeatability evaluates whether the same appraiser can obtain the same value multiple times when measuring the same object using the same equipment under the same environment.
- It refers to the level of agreement between the repeated measurements of the same appraiser under the same condition.
- Repeatability measures the inherent variation of the measurement instrument.



Reproducibility

- Reproducibility evaluates whether different appraisers can obtain the same value when measuring the same object independently.
- It refers to the level of agreement between different appraisers.
- It is not caused by the inherent variation of the measurement instrument. It reflects the variability caused by different appraisers, locations, gauges, environments etc.



- Gauge R&R (i.e. Gauge Repeatability & Reproducibility) is a method to analyze the variability of a measurement system by partitioning the variation of the measurements using ANOVA (Analysis of Variance).
- Gauge R&R only addresses the precision of a measurement system.



- Data collection of a gauge R&R study:
 - let *k* appraisers measure *n* random samples independently and repeat the process *p* times.
- Different appraisers perform the measurement independently.
- The order of measurement (e.g., sequence of samples and sequence of appraisers) is randomized.



The potential sources of variance in the measurement:

• Appraisers: $\sigma_{appraisers}^2$

• Parts: σ_{parts}^2

• Appraisers × Parts: $\sigma_{appraisers \times parts}^2$

• Repeatability: $\sigma_{repeatability}^2$

Variance Components

$$\sigma_{total}^2 = \sigma_{appraisers}^2 + \sigma_{parts}^2 + \sigma_{appraisers \times parts}^2 + \sigma_{repeatability}^2$$



 A valid measurement system has low variability in both repeatability and reproducibility so that the total variability observed can reflect the true variability in the objects (parts) being measured.

$$\sigma_{total}^2 = \sigma_{reproducibility}^2 + \sigma_{repeatability}^2 + \sigma_{parts}^2$$

where

$$\sigma_{reproducibility}^2 = \sigma_{appraisers}^2 + \sigma_{appraisers \times parts}^2$$

 Gauge R&R variance reflects the precision level of the measurement system.

$$\sigma_{R\&R}^2 = \sigma_{repeatability}^2 + \sigma_{reproducibility}^2$$



Variation Components

$$Variation_{total} = Z_0 \times \sigma_{total}$$

$$Variation_{repeatability} = Z_0 \times \sigma_{repeatability}$$

$$Variation_{reproducibility} = Z_0 \times \sigma_{reproducibility}$$

$$Variation_{parts} = Z_0 \times \sigma_{parts}$$

where

$$\sigma_{total}^2 = \sigma_{reproducibility}^2 + \sigma_{repeatability}^2 + \sigma_{parts}^2$$

 Z_0 is a sigma multiplier that assumes a specific confidence level in the spread of the data.



 The percentage of variation R&R contributes to the total variation in the measurement:

$$Contribution\%_{R\&R} = \frac{Variation_{R\&R}}{Variation_{total}} \times 100\%$$

where
$$Variation_{R\&R} = Z_0 imes \sqrt{\sigma_{repeatability}^2 + \sigma_{reproducibility}^2}$$

Measurement System	% Study Var.	% Contribution	Distinct Categories
Acceptable	10% or less	1% or Less	5 or Greater
Marginal	10% - 30%	1% - 9%	
Unacceptable	30% or Greater	9% or Greater	Less than 5



2.3.4 Variable and Attribute MSA



Variable Gage R&R

- Whenever something is measured repeatedly or by different people or processes, the results of the measurements will vary. Variation comes from two primary sources:
 - 1. Differences between the parts being measured
 - 2. The measurement system.
- We can use a gage R&R to conduct a measurement system analysis to determine what portion of the variability comes from the parts and what portion comes from the measurement system.
- There are key study results that help us determine the components of variation within our measurement system.



Key Measures of a Variable Gage R&R

- %Contribution: The percent of contribution for a source is 100 times the variance component for that source divided by the total variance.
- %Study Var (6*SD): The percent of study variation for a source is 100 times the study variation for that source divided by the total variation.
- %Tolerance (SV/Tolerance): The percent of spec range taken up by the total width of the distribution of the data based on variation from that source.
- Distinct Categories: The number of distinct categories of parts that the measurement system is able to distinguish. If a measurement system is not capable of distinguishing at least five types of parts, it is probably not adequate.



Variable Gage R&R Guidelines (AIAG)

Percent Tolerance and Percent Study Variation

- 10% or less Acceptable
- 10% to 30% Marginal
- 30% or greater Unacceptable

Percent Contribution

- 1% or less Acceptable
- 1% to 9% Marginal
- 9% or greater Unacceptable

Distinct Categories

• Look for five or more distinct categories to indicate that your measurement system is acceptable.



Guidelines for Distinct Categories

• Distinct categories is the number of categories of parts that your measurement system can distinguish. If it is below five, it is likely not able to distinguish between parts.

Number of Categories	Conclusion	
Distinct Categories = 1	Measurement system cannot discriminate between parts	
Distinct Categories = 2	Measurement system can only distinguish between high/low or big/small	
Distinct Categories = 3 or 4	Measurement system is of little or no value	
Distinct Categories = 5+	According to AIAG, the measurement system can acceptably discriminate parts	

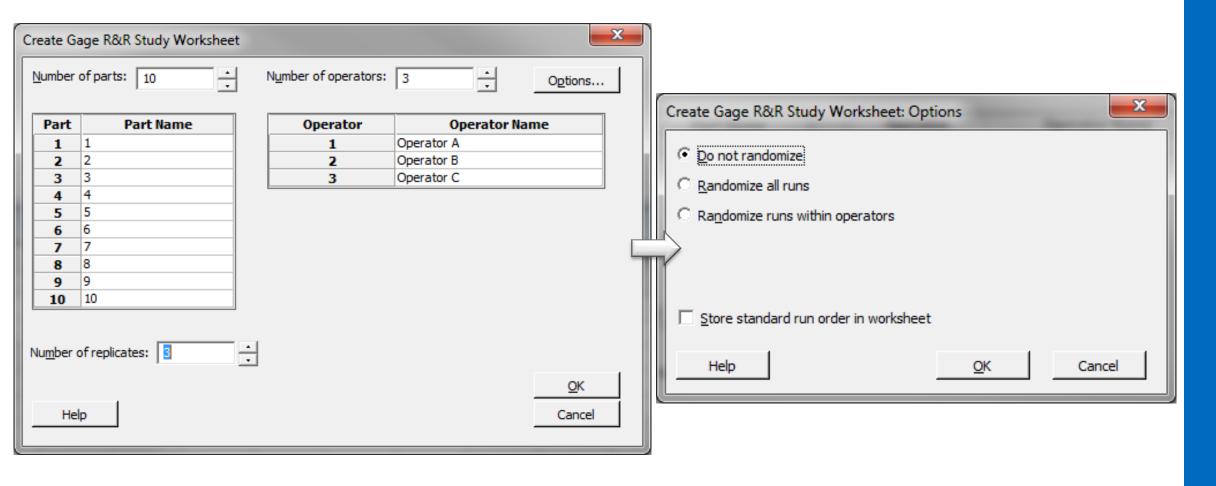


Use Minitab to Implement a Variable MSA

- Data File: "Variable MSA" tab in "Sample Data.xlsx"
- Step 1: Initiate the MSA study
 - 1) Click on Stat → Quality Tools → Gage Study → Create Gage R&R Study Worksheet.
 - 2) A new window named "Create Gage R&R Study Worksheet" pops up.
 - 3) Select 10 as the "Number of Parts."
 - 4) Select 3 as the "Number of Operators."
 - 5) Select 3 as the "Number of Replicates."
 - 6) Enter the part name (e.g., Part 01, Part 02, Part 03, ...).
 - 7) Enter the operator name (e.g., Operator A, Operator B, Operator C).
 - Click on the "Options" button and another window named "Create Gage R&R Study Worksheet – Options" pops up.
 - 9) Select the radio button "Do not randomize."
 - 10) Click "OK" in the window "Create Gage R&R Study Worksheet Options."
 - 11) Click "OK' in the window "Create Gage R&R Study Worksheet."
 - 12) A new data table is generated.



Use Minitab to Implement a Variable MSA





Use Minitab to Implement a Variable MSA

- Step 2: Data collection
 - In the newly-generated data table, Minitab has provided the data layout where we can add our actual measurement data when it's collected.
 - When you actually conduct a variable MSA in your work environment it would be necessary to set up your study just as we have in the previous steps and then you could collect your measurement data properly.
 - However, for our purposes today, we have provided you with an MSA that is setup and data collected. We will use our "Variable MSA" tab in "Sample Data.xlsx," for the next steps.

+	C1	C2-T	C3-T	C4	C5
	RunOrder	Parts	Operators	Measurement	
4	4	2	Operator A		
5	5	2	Operator B		
6	6	2	Operator C		
7	7	3	Operator A		
8	8	3	Operator B		
9	9	3	Operator C		
10	10	4	Operator A		
11	11	4	Operator B		
12	12	4	Operator C		
13	13	5	Operator A		
14	14	5	Operator B		

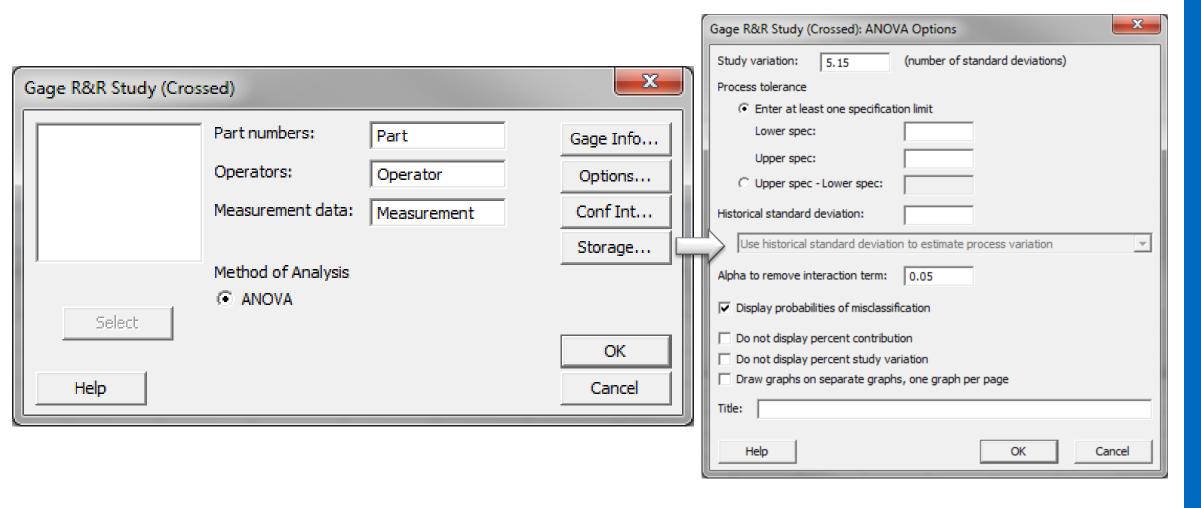
Your data collection plan is now setup for your variable MSA. In real life you would collect your measurement data adding it to this table.

For our purposes today, we have provided you with that data so that we can continue learning how to perform a variable MSA.



- Step 4: Implement Gauge R&R (using our "Variable MSA" data in "Sample Data.xlsx)
 - 1) Click Stat → Quality Tools → Gage Study → Gage R&R Study (Crossed).
 - 2) A new window named "Gage R&R Study (Crossed)" appears.
 - 3) Select "Part" as "Part numbers."
 - 4) Select "Operator" as "Operators."
 - 5) Select "Measurement" as "Measurement data."
 - 6) Click on the "Options" button and another new window named "Gage R&R Study (Crossed) ANOVA Options" pops up.
 - 7) Enter 5.15 as the "Study variation (number of standard deviations)".
 - 8) Click "OK" in the window "Gage R&R Study (Crossed) ANOVA Options."
 - 9) Click "OK" in the window "Gage R&R Study (Crossed)."
 - 10) The MSA analysis results appear in the new window and the session window.







• 5.15 is the recommended standard deviation multiplier by the Automotive Industry Action Group (AIAG). It corresponds to 99% of data in the normal distribution. If we use 6 as the standard deviation multiplier, it corresponds to 99.73% of the data in the normal distribution.

Confidence Level	Sigma Multiplier
90%	3.29
95%	3.92
99%	5.15
99.73%	6



- Step 5: Analyze the MSA results
 - The percentage of variation that R&R contributes to the total variation is 27.86% and the precision level of this measurement system is at best "marginal" bordering on unacceptable.
 - Actions should be required to calibrate the measurement system.

Gage R&R

Variance Components

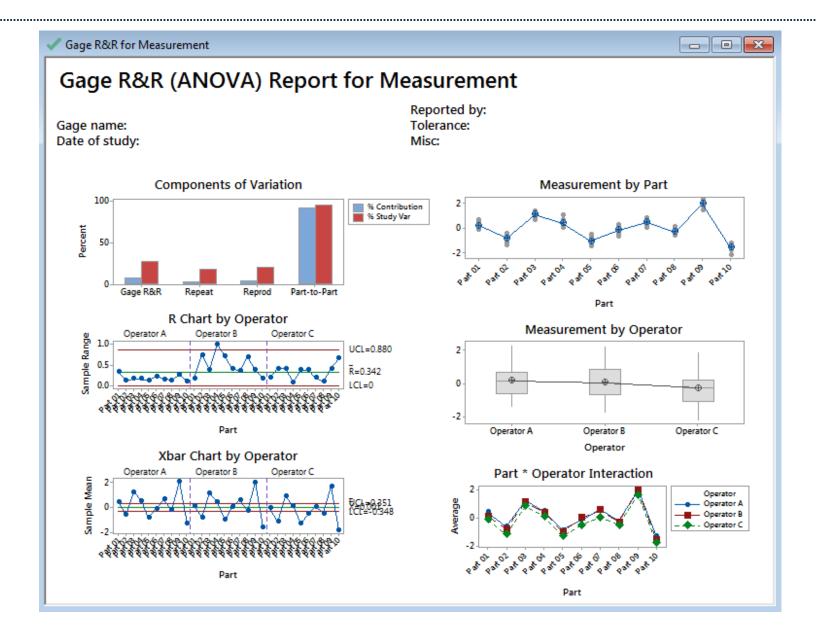
		%Contribution
Source	VarComp	(of VarComp)
Total Gage R&R	0.09143	7.76
Repeatability	0.03997	3.39
Reproducibility	0.05146	4.37
Operator	0.05146	4.37
Part-To-Part	1.08645	92.24
Total Variation	1.17788	100.00

Gage Evaluation

		Study Var	%Study Var
Source	StdDev (SD)	$(5.15 \times SD)$	(%SV)
Total Gage R&R	0.30237	1.55721	27.86
Repeatability	0.19993	1.02966	18.42
Reproducibility	0.22684	1.16821	20.90
Operator	0.22684	1.16821	20.90
Part-To-Part	1.04233	5.36799	96.04
Total Variation	1.08530	5.58929	100.00

Number of Distinct Categories = 4





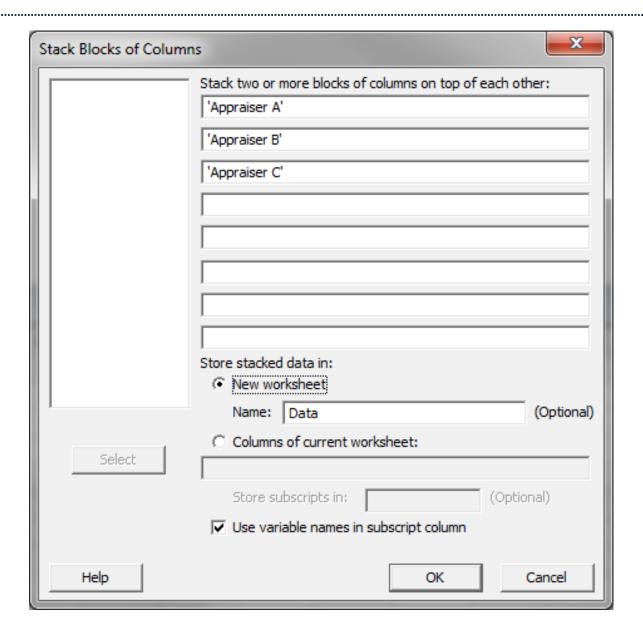


- Data File: "Attribute MSA" tab in "Sample Data.xlsx" (an example in the AIAG MSA Reference Manual, 3rd Edition).
- Steps in Minitab to run an attribute MSA



- Step 1: Reorganize the original data into four new columns (i.e., Appraiser, Assessed Result, Part, and Reference).
 - 1) Click Data → Stack → Blocks of Columns.
 - 2) A new window named "Stack Blocks of Columns" pops up.
 - 3) Select "Appraiser A," "Part," and "Reference" as block one.
 - 4) Select "Appraiser B," "Part," and "Reference" as block two.
 - 5) Select "Appraiser C," "Part," and "Reference" as block three.
 - 6) Select the radio button of "New worksheet" and name the sheet "Data."
 - 7) Check the box "Use variable names in subscript column."
 - 8) Click "OK."
 - 9) The stacked columns are created in the new worksheet named "Data."
 - 10) Name the four columns from left to right in worksheet "Data": Appraiser, Assessed Result, Part, and Reference.







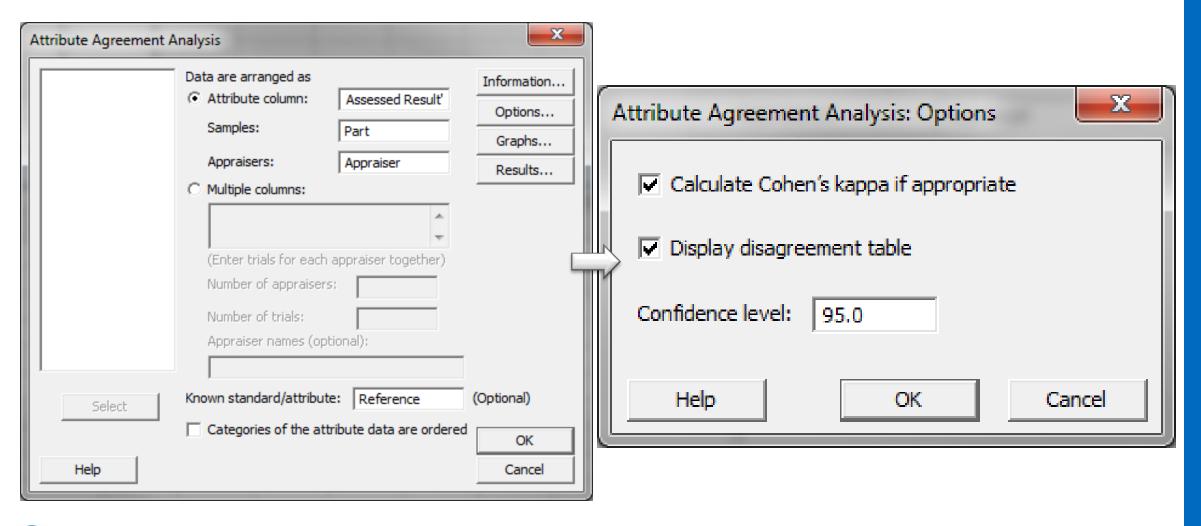
Data ***						
+	C1-T	C2	C3	C4		
	Subscripts					
1	Appraiser A	1	1	1		
2	Appraiser A	1	1	1		
3	Appraiser A	1	1	1		
4	Appraiser A	1	2	1		
5	Appraiser A	1	2	1		
6	Appraiser A	1	2	1		
7	Appraiser A	0	3	0		
8	Appraiser A	0	3	0		
9	Appraiser A	0	3	0		

■ D	■ Data ***							
+	C1-T	C2	C3	C4				
	Appraiser	Assessed Result	Part	Reference				
1	Appraiser A	1	1	1				
2	Appraiser A	1	1	1				
3	Appraiser A	1	1	1				
4	Appraiser A	1	2	1				
5	Appraiser A	1	2	1				
6	Appraiser A	1	2	1				
7	Appraiser A	0	3	0				
8	Appraiser A	0	3	0				
9	Appraiser A	0	3	0				



- Step 2: Run MSA using Minitab
 - Click Stat → Quality Tools → Attribute Agreement Analysis.
 - A new window named "Attribute Agreement Analysis" pops up.
 - 3) Click in the blank box next to "Attribute column" and the variables appear in the list box on the left.
 - 4) Select "Assessed Result" as "Attribute column."
 - 5) Select "Part" as "Samples."
 - 6) Select "Appraiser" as "Appraisers."
 - 7) Select "Reference" as "Known standard/attribute."
 - 8) Click the "Options" button and another window named "Attribute Agreement Analysis Options" pops up.
 - 9) Check the boxes of both "Calculate Cohen's kappa if appropriate" and "Display disagreement table."
 - 10) Click "OK" in the window "Attribute Agreement Analysis Options."
 - 11) Click "OK" in the window "Attribute Agreement Analysis."
 - 12) The MSA results appear in the newly-generated window and the session window.







Attribute Agreement Analysis for Assessed Result

Within Appraisers

Assessment Agreement

Appraiser	# Inspected	# Matched	Percent	95% CI
Appraiser A	50	42	84.00	(70.89, 92.83)
Appraiser B	50	45	90.00	(78.19, 96.67)
Appraiser C	50	40	80.00	(66.28, 89.97)

[#] Matched: Appraiser agrees with him/herself across trials.

Fleiss' Kappa Statistics

Appraiser	Response	Kappa	SE Kappa	Z	P(vs > 0)
Appraiser A	0	0.760000	0.0816497	9.3081	0.0000
	1	0.760000	0.0816497	9.3081	0.0000
Appraiser B	0	0.845073	0.0816497	10.3500	0.0000
	1	0.845073	0.0816497	10.3500	0.0000
Appraiser C	0	0.702911	0.0816497	8.6089	0.0000
	1	0.702911	0.0816497	8.6089	0.0000

Within Appraiser Agreement Percent: the agreement percentage within each individual appraiser.



Each Appraiser vs Standard

Assessment Agreement

Appraiser	# Inspected	# Matched	Percent	95% CI
Appraiser A	50	42	84.00	(70.89, 92.83)
Appraiser B	50	45	90.00	(78.19, 96.67)
Appraiser C	50	40	80.00	(66.28, 89.97)

[#] Matched: Appraiser's assessment across trials agrees with the known standard.

Assessment Disagreement

Appraiser	#1/0	Percent	#0/1	Percent	# Mixed	Percent
Appraiser A	0	0.00	0	0.00	8	16.00
Appraiser B	0	0.00	0	0.00	5	10.00
Appraiser C	0	0.00	0	0.00	10	20.00

^{# 1 / 0:} Assessments across trials = 1 / standard = 0.

Fleiss' Kappa Statistics

Appraiser	Response	Kappa	SE Kappa	Z	P(vs > 0)
Appraiser A	0	0.880236	0.0816497	10.7806	0.0000
	1	0.880236	0.0816497	10.7806	0.0000
Appraiser B	0	0.922612	0.0816497	11.2996	0.0000
	1	0.922612	0.0816497	11.2996	0.0000
Appraiser C	0	0.774703	0.0816497	9.4881	0.0000
	1	0.774703	0.0816497	9.4881	0.0000

Each Appraiser vs. Standard Agreement Percent: the agreement percentage between each appraiser and the standard. It reflects the accuracy of the measurement system.



^{# 0 / 1:} Assessments across trials = 0 / standard = 1.

[#] Mixed: Assessments across trials are not identical.

Between Appraisers

Assessment Agreement

Inspected # Matched Percent 95% Cl
50 39 78.00 (64.04, 88.47)

Matched: All appraisers' assessments agree with each other.

Fleiss' Kappa Statistics

Response	Kappa	SE Kappa	Z	P(vs > 0)
0	0.793606	0.0235702	33.6698	0.0000
1	0.793606	0.0235702	33.6698	0.0000

All Appraisers vs. Standard Agreement Percent: overall agreement percentage of both within and between appraisers. It reflects how precise the measurement system performs. Between Appraiser Agreement Percent: the agreement percentage between different appraisers.

All Appraisers vs Standard

Assessment Agreement

Inspected # Matched Percent 95% CI 50 39 78.00 (64.04, 88.47)

Matched: All appraisers' assessments agree with the known standard.

Fleiss' Kappa Statistics

Response	Kappa	SE Kappa	Z	P(vs > 0)
0	0.859184	0.0471405	18.2260	0.0000
1	0.859184	0.0471405	18.2260	0.0000

Cohen's Kappa Statistics

Response	Kappa	SE Kappa	Z	P(vs > 0)
0	0.859313	0.0470879	18.2491	0.0000
1	0.859313	0.0470879	18.2491	0.0000



- Kappa statistic is a coefficient indicating the agreement percentage above the expected agreement by chance.
- Kappa ranges from -1 (perfect disagreement) to 1 (perfect agreement).
- When the observed agreement is less than the chance agreement, Kappa is negative.
- When the observed agreement is greater than the chance agreement, Kappa is positive.
- Rule of thumb: If Kappa is greater than 0.7, the measurement system is acceptable. If Kappa is greater than 0.9, the measurement system is excellent.



Attribute Agreement Analysis for Assessed Result

Within Appraisers

Assessment Agreement

Appraiser	# Inspected	# Matched	Percent	95% CI
Appraiser A	50	42	84.00	(70.89, 92.83)
Appraiser B	50	45	90.00	(78.19, 96.67)
Appraiser C	50	40	80.00	(66.28, 89.97)

[#] Matched: Appraiser agrees with him/herself across trials

Fleiss' Kappa Statistics

Appraiser	Response	Карра	SE Kappa	Z	P(vs > 0)
Appraiser A	0	0.760000	0.0816497	9.3081	0.0000
	1	0.760000	0.0816497	9.3081	0.0000
Appraiser B	0	0.845073	0.0816497	10.3500	0.0000
	1	0.845073	0.0816497	10.3500	0.0000
Appraiser C	0	0.702911	0.0816497	8.6089	0.0000
	1	0.702911	0.0816497	8.6089	0.0000

Kappa statistic of the agreement between individual appraiser and the standard

Kappa statistic of the agreement within each appraiser

Each Appraiser vs Standard

Fleiss' Kappa Statistics

Appraiser	Response	Kappa	SE Kappa	Z	P(vs > 0)
Appraiser A	0	0.880236	0.0816497	10.7806	0.0000
	1	0.880236	0.0816497	10.7806	0.0000
Appraiser B	0	0.922612	0.0816497	11.2996	0.0000
	1	0.922612	0.0816497	11.2996	0.0000
Appraiser C	0	0.774703	0.0816497	9.4881	0.0000
	1	0.774703	0.0816497	9.4881	0.0000

Cohen's Kappa Statistics

Appraiser	Response	Kappa	SE Kappa	Z	P(vs > 0)
Appraiser A	0	0.880395	0.0815419	10.7968	0.0000
	1	0.880395	0.0815419	10.7968	0.0000
Appraiser B	0	0.922634	0.0816199	11.3040	0.0000
	1	0.922634	0.0816199	11.3040	0.0000
Appraiser C	0	0.774910	0.0815140	9.5065	0.0000
	1	0.774910	0.0815140	9.5065	0.0000



Between Appraisers

Assessment Agreement

# Inspected	# Matched	Percent	95% CI
50	39	78.00	(64.04, 88.47)

Matched: All appraisers' assessments agree with each other.

Fleiss' Kappa Statistics

Res	sponse	Карра	SE Kappa	Z	P(vs > 0)
0		0.793606	0.0235702	33.6698	0.0000
1		0.793606	0.0235702	33.6698	0.0000

Kappa statistic of the overall agreement between appraisers and the standard. Again, greater than 0.7 suggests an acceptable measurement system.

Kappa statistic of the agreement between appraisers. They are greater than 0.7 and acceptable.

All Appraisers vs Standard

Assessment Agreement

# Inspected	# Matched	Percent	95% CI
50	39	78.00	(64.04, 88.47)

Matched: All appraisers' assessments agree with the known standard.

Fleiss' Kappa Statistics

- 00	Response	Kappa	SE Kappa	Z	P(vs > 0)
	0	0.859184	0.0471405	18.2260	0.0000
	1	0.859184	0.0471405	18.2260	0.0000

Cohen's Kappa Statistics

Response	Kappa	SE Kappa	Z	P(vs > 0)
0	0.859313	0.0470879	18.2491	0.0000
1	0.859313	0.0470879	18.2491	0.0000



2.4 Process Capability



Black Belt Training: Measure Phase

2.1 Process Definition

- 2.1.1 Cause and Effect Diagrams
- 2.1.2 Cause and Effects Matrix
- 2.1.3 Process Mapping
- 2.1.4 FMEA: Failure Modes & Effects Analysis
- 2.1.5 Theory of Constraints

2.2 Six Sigma Statistics

- 2.2.1 Basic Statistics
- 2.2.2 Descriptive Statistics
- 2.2.3 Distributions and Normality
- 2.2.4 Graphical Analysis

2.3 Measurement System Analysis

- 2.3.1 Precision and Accuracy
- 2.3.2 Bias, Linearity, and Stability
- 2.3.3 Gage R&R
- 2.3.4 Variable and Attribute MSA

2.4 Process Capability

- 2.4.1 Capability Analysis
- 2.4.2 Concept of Stability
- 2.4.3 Attribute and Discrete Capability
- 2.4.4 Monitoring Techniques



2.4.1 Capability Analysis



What is Process Capability?

- The process capability measures how well the process performs to meet given specified outcome.
- It indicates the conformance of a process to meet given requirements or specifications.
- Capability analysis helps to better understand the performance of the process with respect to meeting customer's specifications and identify the process improvement opportunities.



Process Capability Analysis Steps

- Step 1: Determine the metric or parameter to measure and analyze.
- Step 2: Collect the historical data for the parameter of interest.
- Step 3: Prove the process is statistically stable (i.e., in control).
- Step 4: Calculate the process capability indices.
- Step 5: Monitor the process and ensure it remains in control over time.
 Update the process capability indices if needed.



Process Capability Indices

- Process capability can be presented using various indices depending on the nature of the process and the goal of the analysis.
- Popular process capability indices:
 - C_p
 - Pp
 - C_{pk}
 - P_{pk}
 - C_{pn}



C_p

C_p stands for capability of the process.

$$C_p = \frac{USL - LSL}{6 \times \sigma_{within}}$$

where

$$\sigma_{within} = \frac{s_p}{c_4(d+1)} \qquad \qquad s_p = \sqrt{\frac{\sum_i \sum_j (x_{ij} - \overline{x_i})}{\sum_i (n_i - 1)}}$$

$$d = \sum_{i} (n_i - 1)$$

$$c_4 = \frac{4(n-1)}{(4n-3)}$$

USL and **LSL** are the upper and lower specification limits. *n* is the sample size.



C_p

- C_p measures the process' potential capability to meet the two-sided specifications.
- It does not take the process average into consideration.
- High C_p indicates the small spread of the process with respect to the spread of the customer specifications.
- C_p is recommended when the process is centered between the specification limits.
- C_p works when there are both upper and lower specification limits.



• P_p stands for **performance of the process**.

$$P_p = \frac{USL - LSL}{6 \times \sigma_{overall}}$$

where

$$\sigma_{overall} = \frac{S}{c_4(n)}$$

$$s = \sqrt{\sum_{i} \sum_{j} \frac{(x_{ij} - \overline{x})^{2}}{n - 1}}$$

$$c_4 = \frac{4(n-1)}{(4n-3)}$$

USL and **LSL** are the upper and lower specification limits. *n* is the sample size.



- Similar to C_p, P_p measures the capability of the process to meet the two-sided specifications.
- It only focuses on the spread and does not take the process centralization into consideration.
- It is recommended when the process is centered between the specification limits.
- C_p considers the within-subgroup standard deviation and P_p considers the total standard deviation from the sample data.

P_D works when there are both upper and lower specification limits.

$C_{\sf pk}$

C_{pk} stands for the capability of the process with a k factor adjustment.

$$C_{pk} = (1 - k) \times C_p$$

where

$$k = \frac{|m - \mu|}{USL - LSL} \qquad m = \frac{USL + LSL}{2}$$

 μ is the process mean; n is the sample size.

USL and LSL are the upper and lower specification limits.



• The formulas to calculate $C_{\rm pk}$ can also be expressed as follows:

$$C_{pk} = \min \left(\frac{USL - \mu}{3 \times \sigma_{within}}, \frac{\mu - LSL}{3 \times \sigma_{within}} \right)$$

where

$$\sigma_{within} = \frac{S_p}{c_4(d+1)}$$

$$S_p = \sqrt{\frac{\sum_{i} \sum_{j} (x_{ij} - x_i)}{\sum_{i} (n_i - 1)}}$$

$$d = \sum_{i} (n_i - 1)$$

$$c_4 = \frac{4(n-1)}{(4n-3)}$$

USL and **LSL** are the upper and lower specification limits.



- C_{pk} measures the process' actual capability by taking both the variation and average of the process into consideration.
- The process does not need to be centered between the specification limits to make the index meaningful.
- C_{pk} is recommended when the process is not in the center between the specification limits.
- When there is only a one-sided limit, C_{pk} is calculated using C_{pu} or C_{pl}.



• C_{pk} for upper specification limit:

$$C_{pu} = \frac{USL - \mu}{3 \times \sigma_{within}}$$

• C_{pk} for lower specification limit:

$$C_{pl} = \frac{\mu - LSL}{3 \times \sigma_{within}}$$

USL and **LSL** are the upper and lower specification limits. μ is the process mean.



• P_{pk} stands for the performance of the process with a k factor adjustment.

$$P_{pk} = (1 - k) \times P_p$$

where

$$k = \frac{\left| m - \mu \right|}{USL - LSL}$$

$$m = \frac{USL + LSL}{2}$$

 $\it USL$ and $\it LSL$ are the upper and lower specification limits. $\it \mu$ is the process mean.



• The formulas to calculate P_{pk} can also be expressed as follows:

$$P_{pk} = \min \left(\frac{USL - \mu}{3 \times \sigma_{overall}}, \frac{\mu - LSL}{3 \times \sigma_{overall}} \right)$$

$$\sigma_{overall} = \frac{s}{c_4(n)} \qquad s = \sqrt{\sum_i \sum_j \frac{(x_{ij} - \overline{x})^2}{n - 1}}$$

$$c_4 = \frac{4(n-1)}{(4n-3)}$$

USL and **LSL** are the upper and lower specification limits. μ is the process mean. n is the sample size.



- Similar to C_{pk} , P_{pk} measures the process capability by taking both the variation and the average of the process into consideration.
- P_{pk} solves the decentralization problem P_p cannot overcome.
- C_{pk} considers the within-subgroup standard deviation, while P_{pk} considers the total standard deviation from the sample data.
- When there is only a one-sided specification limit, P_{pk} is calculated using P_{pu} or P_{pl} .



• P_{pk} for upper specification limit:

$$P_{pu} = \frac{USL - \mu}{3 \times \sigma_{overall}}$$

P_{pk} for lower specification limit:

$$P_{pl} = \frac{\mu - LSL}{3 \times \sigma_{overall}}$$

USL and **LSL** are the upper and lower specification limits.



$C_{\sf pm}$

- C_p , P_p , C_{pk} , and P_{pk} all consider the variation of the process. C_{pk} and P_{pk} take both the variation and the average of the process into consideration when measuring the process capability.
- It is possible that the process average fails to meet the target customers require while the process still remains between the specification limits. C_{pm} (Taguchi's capability index) helps to capture the variation from the specified target.



$C_{\sf pm}$

Formula to calculate C_{pm}

$$C_{pm} = \frac{\min(T - LSL, USL - T)}{3 \times \sqrt{s^2 + (\mu - T)^2}}$$

USL and **LSL** are the upper and lower specification limits. T is the specified target. μ is the process mean.

Note: Cpm can work only if there is a target value specified.



Interpreting Capability

- Interpreting the results of a capability analysis is very dependent upon the nature of the business, product and/or customer. The Automotive Industry Action Group (AIAG) suggests that:
 - P_p and P_{pk} should be > 1.67
 - C_p and C_{pk} should be > 1.33.
- The reality is that anything greater than 1.0 is a fairly capable process but your business needs to assess the costs vs. benefits of achieving capability greater than 1.67 or even higher. Below is a simple table for quick interpretation

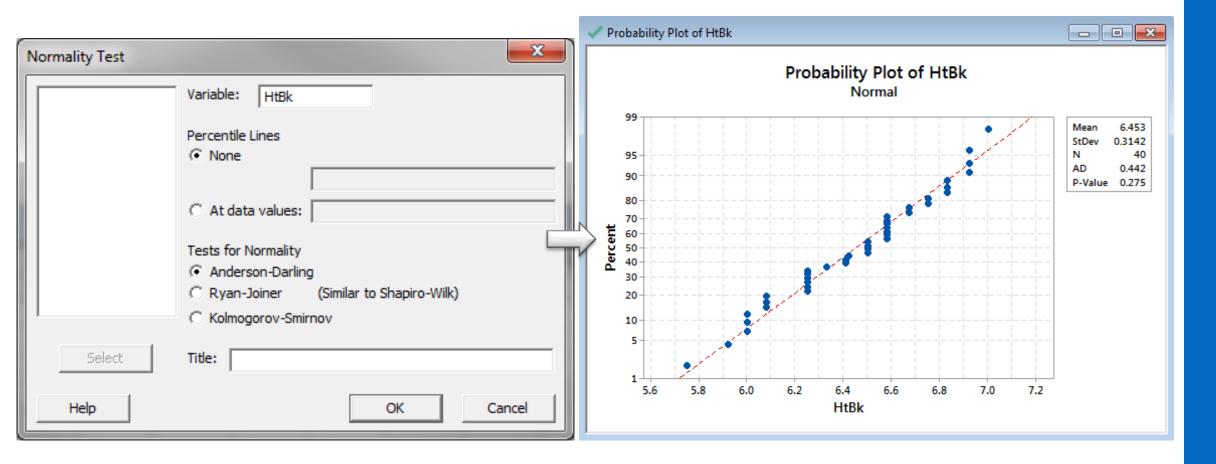
Index	Value	Interpretation	Sigma Level
C_{pk}	<1.0	Not Very Capable	<3
C _{pk}	1.0-1.99	Capable	3-6
C _{pk}	>2.0	Very Capable	>6

• If you're dealing with customer safety or life and death influences then obviously the product necessitates a capability greater than 2.0.

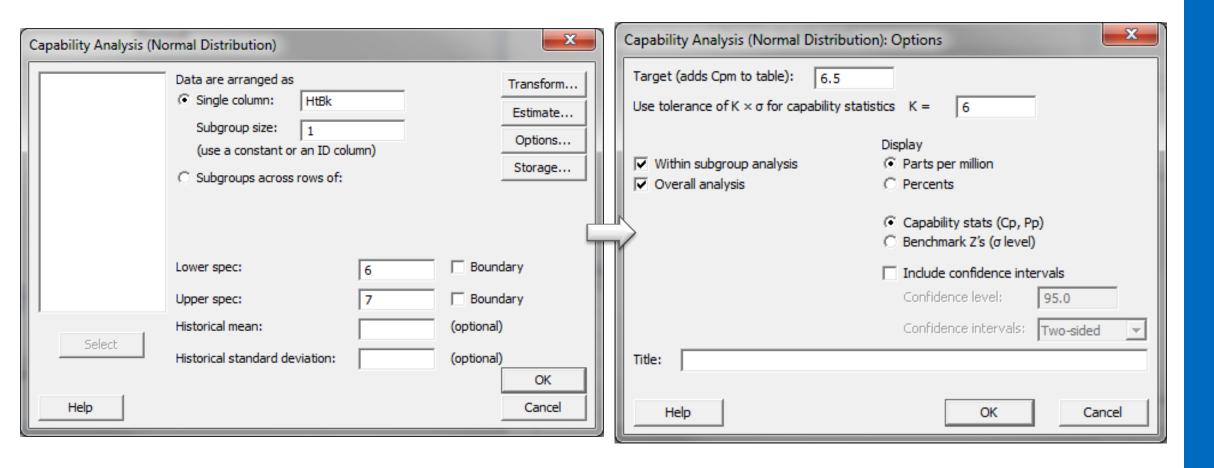


- Data File: "Capability Analysis" tab in "Sample Data.xlsx"
- Steps in Minitab to run a process capability analysis:
 - Click Stat → Basic Statistics → Normality Test.
 - A new window named "Normality Test" pops up.
 - Select "HtBk" as the variable and Click "OK."
 - The histogram and the normality test results appear in the new window.
 - In this example, the p-value is 0.275, greater than the alpha level (0.05). We fail to reject the hypothesis and conclude that the data are normally distributed.
 - Click Stat \rightarrow Quality Tools \rightarrow Capability Analysis \rightarrow Normal.
 - A new window named "Capability Analysis (Normal Distribution)" pops up.
 - Select "HtBk" as the single column and enter "1" as the subgroup size.
 - Enter "6" as the "Lower spec" and "7" as the "Upper spec"
 - 10) Click "Options" button and another new window named "Capability Analysis (Normal Distribution) Options" pops up.
 - 11) Enter "6.5" as the target and click "OK."
 - Click "OK" in the "Capability Analysis (Normal Distribution)" window. The capability analysis results are displayed.

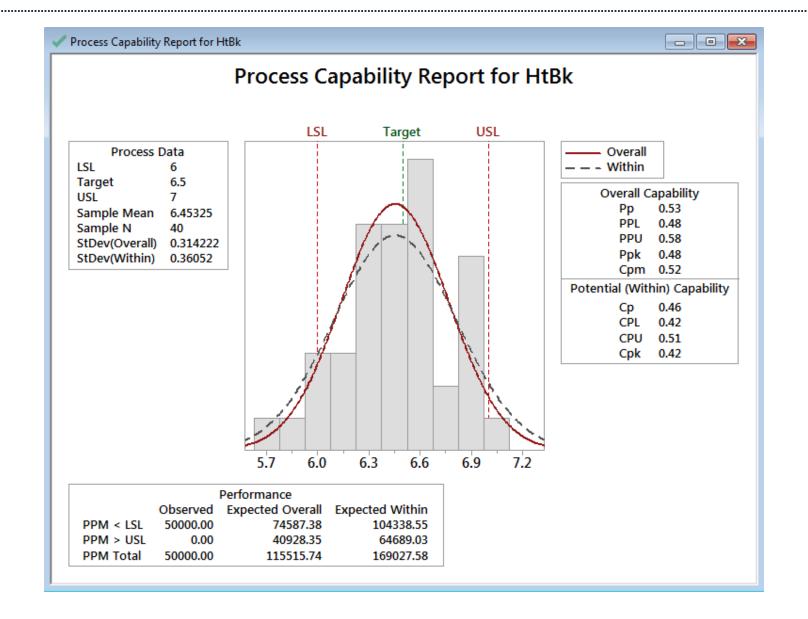








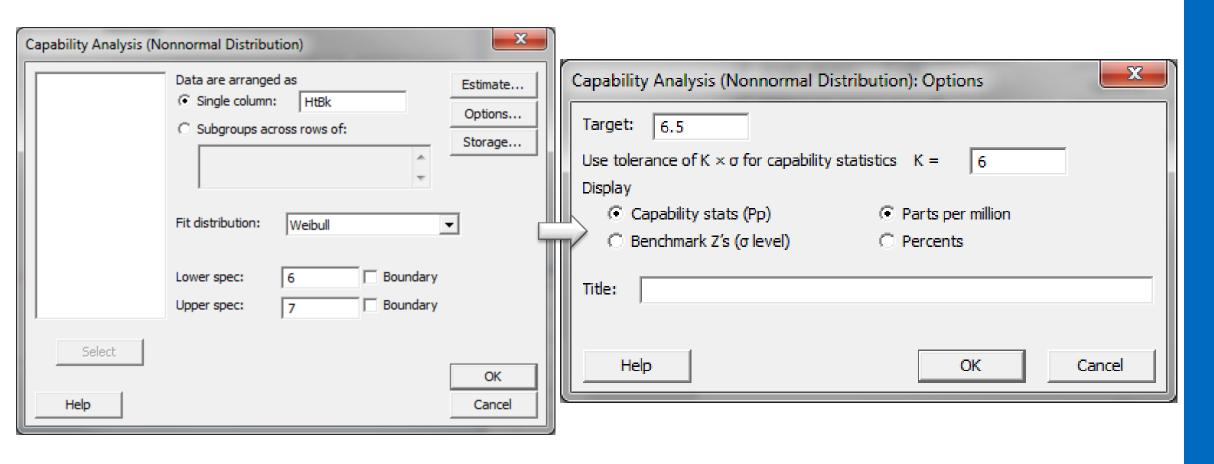




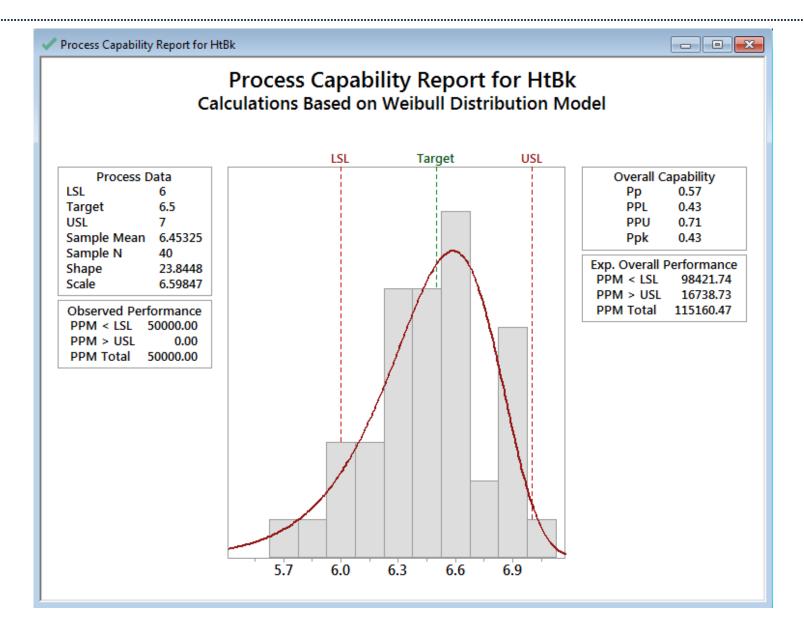


- If the p-value of the normality test is smaller than the alpha level (0.05), we reject the null hypothesis and conclude that the data are not normally distributed. In this case we run a non-normal analysis.
 - 1) Click Stat → Quality Tools → Capability Analysis → Nonnormal
 - 2) A new window named "Capability Analysis (Nonnormal Distribution)" pops up.
 - 3) Select "HtBk" as the single column.
 - 4) Enter "6" as the "Lower spec" and "7" as the "Upper spec."
 - 5) Click "Options" button and another new window named "Capability Analysis (Nonnormal Distribution) Options" pops up.
 - 6) Enter "6.5" as the target and click "OK."
 - 7) Click "OK" in the "Capability Analysis (Nonnormal Distribution)" window.
 - 8) The capability analysis results appear in the new window.











2.4.2 Concept of Stability



What is Process Stability?

- A process is said to be stable when:
 - the process is in control
 - the future behavior of the process is predictable at least between some limits
 - there is only random variation involved in the process.
 - the causes of variation in the process are only due to chance or common causes
 - there are not any trends, patterns, or outliers in the control chart of the process.



Root Causes of Variation in the Process

Common Cause:

- Chance
- Random and anticipated
- Natural noise
- Inherent in the process
- Unable to be eliminated from the process.

Special Cause:

- Assignable cause
- Unanticipated
- Unnatural pattern
- Signal of changes in the process
- Able to be eliminated from the process.



Control Charts

- Control charts are the graphical tools to analyze the stability of a process.
- A control chart is used to identify the presence of potential special causes in the process and to determine whether the process is statistically in control.
- If the samples or calculations of samples are all in control, the process is stable and the data from the process can be used to predict the future performance of the process.



Popular Control Charts

- I-MR Chart
- Xbar-R Chart
- Xbar-S Chart
- C Chart
- U Chart
- P Chart
- NP Chart
- EWMA Chart
- CUSUM Chart

Note: More details of the control charts will be introduced in the Control module.



Process Stability vs. Process Capability

- Process stability indicates how stable a process performed in the past.
- When the process is stable, we can use the data from the process to predict its future behavior.
- Process capability indicates how well a process performs with respect to meeting the customer's specifications.
- The process capability analysis is valid only if the process is statistically stable (i.e., in control, predictable).
- Being stable does *not* guarantee that the process is also capable. However, being stable is the prerequisite to determine whether a process is capable.



2.4.3 Attribute & Discrete Capability



Process Capability Analysis for Binomial Data

• If we are measuring the count of defectives in each sample set to assess the process performance of meeting the customer specifications, we use "%Defective" (percentage of items in the samples that are defective) as the process capability index.

%Defective =
$$\frac{N_{defectives}}{N_{overall}}$$

where $N_{\text{defectives}}$ is the total count of defectives in the samples and N_{overall} is the sum of all the sample sizes.



Process Capability Analysis for Poisson Data

• If we are measuring the count of defects in each sample set to assess the process performance of meeting the customer specifications, we use Mean DPU (defects per unit of measurement) as the process capability index.

$$DPU = rac{N_{defects}}{N_{overall}}$$

where $N_{defects}$ is the total count of defects in the samples and $N_{overall}$ is the sum of all the units in the samples.



2.4.4 Monitoring Techniques



Capability and Monitoring

- In the Measure phase of the project, process stability analysis and process capability analysis are used to baseline the performance of current process.
- A stable process means we can expect the same process capability in the future.
- In the Control phase of the project, process stability analysis and process capability analysis are combined to monitor whether the improved process is maintained consistently as expected.



3.0 Analyze Phase



Black Belt Training: Analyze Phase

3.1 Inferential Statistics

- 3.1.1 Understanding Inference
- 3.1.2 Sampling Techniques and Uses
- 3.1.3 Sample Size
- 3.1.4 Central Limit Theorem

3.2 Hypothesis Testing

- 3.2.1 Goals of Hypothesis Testing
- 3.2.2 Statistical Significance
- 3.2.3 Risk; Alpha and Beta
- 3.2.4 Types of Hypothesis Tests

3.3 Hypothesis Testing: Normal Data

- 3.3.1 One and Two Sample T-Tests
- 3.3.2 One sample variance
- 3.3.3 One Way ANOVA

3.4 Hypothesis Testing: Non-Normal Data

- 3.4.1 Mann-Whitney
- 3.4.2 Kruskal-Wallis
- 3.4.3 Moods Median
- 3.4.4 Friedman
- 3.4.5 One Sample Sign
- 3.4.6 One Sample Wilcoxon
- 3.4.7 One and Two Sample Proportion
- 3.4.8 Chi-Squared (Contingency Tables)
- 3.4.9 Test of Equal Variances



3.1 Inferential Statistics



Black Belt Training: Analyze Phase

3.1 Inferential Statistics

- 3.1.1 Understanding Inference
- 3.1.2 Sampling Techniques and Uses
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- 3.1.4 Central Limit Theorem

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- 3.2.3 Risk; Alpha and Beta
- 3.2.4 Types of Hypothesis Tests

3.3 Hypothesis Testing: Normal Data

- 3.3.1 One and Two Sample T-Tests
- 3.3.2 One sample variance
- 3.3.3 One Way ANOVA

3.4 Hypothesis Testing: Non-Normal Data

- 3.4.1 Mann-Whitney
- 3.4.2 Kruskal-Wallis
- 3.4.3 Moods Median
- 3.4.4 Friedman
- 3.4.5 One Sample Sign
- 3.4.6 One Sample Wilcoxon
- 3.4.7 One and Two Sample Proportion
- 3.4.8 Chi-Squared (Contingency Tables)
- 3.4.9 Test of Equal Variances



3.1.1 Understanding Inference



What is Statistical Inference?

- Statistical inference is the process of making inferences regarding the characteristics of an unobservable population based on the characteristics of an observed sample.
- We rely on sample data to draw conclusions about the population from which the sample is drawn.
- Statistical inference is widely used since it is difficult or sometimes impossible to collect the entire population data.

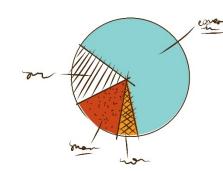


Outcome of Statistical Inference

- The outcome or conclusion of statistical inference is a statistical proposition about the population.
- Examples of statistical propositions:
 - Estimating a population parameter
 - Identifying an interval or a region where the true population parameter would fall with some certainty
 - Deciding whether to reject a hypothesis made on characteristics of the population of interest
 - Making predictions
 - Clustering or partitioning data into different groups.



Population and Sample



- A **statistical population** is an entire set of objects or observations about which statistical inferences are to be drawn based on its sample.
- It is usually impractical or impossible to obtain the data for the entire population. For example, if we are interested in analyzing the population of all the trees, it is extremely difficult to collect the data for all the trees that existed in the past, exist now, and will exist in the future.
- A **sample** is a subset of the population (like a piece of the pie above). It is necessary for samples to be *representative* of the population.
- The process of selecting a subset of observations within a population is referred to as sampling.



Population and Sample

- Population Parameters (Greek letters)
 - Mean: µ
 - Standard deviation: σ
 - Variance: σ²
 - Median: η

- Sample Statistics (Roman letters)
 - Sample Mean: \bar{X}
 - Standard deviation: S
 - Variance: S²
 - Median: X

- The population parameter is the numerical summary of a population.
- The sample statistic is the numerical measurement calculated based on a sample of that population. It is used to estimate the true population parameter.



Descriptive Statistics vs. Statistical Inference

Descriptive Statistics

- Descriptive statistics summarize the characteristics of a collection of data.
- Descriptive statistics are descriptive only and they do not make any generalizations beyond the data at hand.
- Data used for descriptive statistics are for the purpose of representing or reporting.

Statistical Inference

- Statistical inference makes generalizations from a sample at hand of a population.
- Data used for statistical inference are for the purpose of making inferences on the entire population of interest.
- A complete statistical analysis includes both descriptive statistics and statistical inference.



Error Sources of Statistical Inference

- Statistical inference uses sample data to best approximate the true features of the population.
- A valid sample must be unbiased and representative of the population.
- Two sources of error in statistical inference:
 - Random sampling error
 - Selection bias.



Error Sources of Statistical Inference

- Random Sampling Error
 - Random variation due to observations being selected randomly
 - It is inherent to the sampling process and beyond one's control.
- Selection Bias
 - Non-random variation due to inadequate design of sampling
 - It can be improved by adjusting the sampling size and sampling strategy.



3.1.2 Sampling Techniques



What is Sampling?

- Sampling is the process of selecting objects or observations from a population of interest about which we wish to make a statistical inference.
- It is extensively used to collect information about a population, which can represent physical or intangible objects.





Advantages of Sampling

- It is usually impractical or impossible to collect the data of an entire population:
 - High cost
 - Time consuming
 - Unavailability of historical records
 - Dynamic nature of the population.
- Advantages of sampling a representative subset of the population:
 - Lower cost
 - Faster data collection
 - Easier to manipulate.



Uses of Sample

- With valid samples collected, we can draw statistical conclusions about the population of our interest.
- There are two major uses of samples in making statistical inference:
 - Estimation: estimating the population parameters using the sample statistics.
 - *Hypothesis testing*: testing a statement about the population characteristics using sample data.
- This module covers sample size calculation for estimation purposes only. Sample size calculations for hypothesis testing purposes will be covered in the Hypothesis Testing module.



Basic Sampling Steps

- 1) Determine the population of interest
- 2) Determine the sampling frame
- 3) Determine the sampling strategy
- 4) Calculate the sample size
- 5) Conduct sampling



Population

- **Population** in statistics is the entire set of objects or observations about which we are interested to draw conclusions or make generalizations based on some representative sample data.
- A population can be either physical or intangible.
 - Physical: trees, customers, monitors etc.
 - Intangible: credit score, pass/fail decisions etc.
- A population can be static or dynamic
 - Characteristics of individuals are relatively static over time
 - Items making up the population continue to change or be generated over time
- The population covers all the items with characteristics we are interested to analyze.



Sampling Frame

- In the ideal situation, the scope of a population could be defined.
 - Example: Mary is interested to know how the soup she is cooking tastes. The population is simply the pot of soup.
- However, in some other situations, the population cannot be identified or defined precisely.
 - Example: to collect the information for an opinion poll, we do not have a list of all the people in the world at hand.
- A **sampling frame** is a list of items of the population (preferably the entire population; if not, approximate population).
 - Example: The telephone directory would be a sampling frame for opinion poll data collection.



Basic Sampling Strategies

- Simple Random Sampling
- Matched Random Sampling
- Stratified Sampling
- Systematic Sampling
- Proportional to Size Sampling
- Cluster Sampling



Simple Random Sampling

- Simple random samples are selected in such a way that each item in the population has an equal chance of being selected.
- There is no bias involved in the sample selection. Such selection minimizes the variation between the characteristics of samples and the population.
- It is the basic sampling strategy.



Matched Random Sampling

 Matched random samples are samples randomly selected in pairs, each of which has the same attribute.

- Example:
 - Researchers are interested in understanding the weight of twins.
 - Researchers are interested in understanding the patients' blood pressure before and after taking some medicine.



Stratified Sampling

- A population can be grouped or "stratified" into distinct and independent categories. An individual category can be considered as a sub-population. **Stratified samples** are randomly selected in each category of the population.
- The categories can be gender, region, income level etc.
- Stratified sampling requires advanced knowledge of the population characteristics.
- Example: A fruit store wants to measure the quality of all their oranges. They decide to use stratified sampling by region to collect sample data. Since about 40% of their oranges are from California, 40% of the sample is selected from the California oranges sub-population.



Systematic Sampling

- Systematic samples are selected at regular intervals based on an ordered list where items in the population are arranged according to a certain criterion.
- Systematic sampling has a random start and then every ith item is selected going forward.
- For example, we are sampling the every 5th unit produced on the production line.



Cluster Sampling

- Cluster sampling is a sampling method in which samples are only selected from certain clusters or groups of the population.
- It reduces the cost and time spent on the sampling but bears the risk that the selected clusters are biased.
- For example, selecting samples from the region where researchers are located so that the cost and time spent on travelling is reduced.



Sampling Strategy Decision Factors

- When determining the sampling strategy, we need to consider the following factors:
 - Cost and time constraints
 - Nature of the population of interest
 - Availability of advanced knowledge of the population
 - Accuracy requirement.



Sample Size

- The sample size is a critical element that can influence the results of statistical inference.
- The smaller the sample size, the higher the risk that the sample statistic will not reflect the true population parameter.
- The greater the sample size, the more time and money we will spend on collecting the samples.



Sample Size Factors

- Is the variable of interest continuous or discrete?
- How large is the population size?
- How much risk do you want to take regarding missing the true population parameters?
- What is the acceptable margin of error you want to detect?
- How much is the variation in the population?



Sample Size Calculation for Continuous Data

Sample size equation for continuous data

$$n_0 = \left(\frac{Z_{\alpha/2} \times s}{d}\right)^2$$

where

 n_0 is the number of samples.

 $Z_{\alpha/2}$ is the Z score when risk level is $\alpha/2$.

- When α is 0.05, it is 1.96.
- When α is 0.10, it is 1.65.

s is the estimation of standard deviation in the population d is the acceptable margin of error.



Sample Size Calculation for Continuous Data

When the sample size calculated using the formula

$$n_0 = \left(\frac{Z_{\alpha/2} \times s}{d}\right)^2$$

exceeds 5% of the population size, we use a correction formula to calculate the final sample size.

$$n = \frac{n_0}{\left(1 + \frac{n_0}{N}\right)}$$

where

 n_0 is the sample size calculated using equation $n_0 = \left(\frac{Z_{\alpha/2} \times s}{d}\right)^2$ *N* is the population size.

$$n_0 = \left(\frac{Z_{\alpha/2} \times s}{d}\right)^2$$



Sample Size Calculation for Discrete Data

Sample size equation for discrete data

$$n_0 = \left(\frac{Z_{\alpha/2}}{d}\right)^2 \times p \times (1-p)$$

where

 n_0 is the number of samples.

 $Z_{\alpha/2}$ is the Z score when risk level is $\alpha/2$.

- When α is 0.05, it is 1.96.
- When *α* is 0.10, it is 1.65.

s is the estimation of standard deviation in the population.

d is the acceptable margin of error.

p is the proportion of one type of event occurring (e.g., proportion of passes).

 $p \times (1 - p)$ is the estimate of variance.



Sample Size Calculation for Discrete Data

When the sample size calculated using the formula

$$n_0 = \left(\frac{Z_{\alpha/2}}{d}\right)^2 \times p \times (1-p)$$

exceeds 5% of the population size, we use a correction formula to calculate the final sample size.

$$n = \frac{n_0}{\left(1 + \frac{n_0}{N}\right)}$$

where

where n_0 is the sample size calculated using equation $n_0 = \left(\frac{Z_{\alpha/2}}{d}\right)^2 \times p \times (1-p)$ *N* is the population size.

$$n_0 = \left(\frac{Z_{\alpha/2}}{d}\right)^2 \times p \times (1-p)$$



Sampling Errors

- Random Sampling Error
 - Random variation due to observations being selected randomly
 - It is inherent in the sampling process and beyond one's control.
- Selection Bias
 - Non-random variation due to inadequate design of sampling
 - It can be improved by adjusting the sampling size and sampling strategy.



3.1.3 Sample Size



Sample Size

- The sample size is a critical element that can influence the results of hypothesis testing.
- The smaller the sample size, the higher the risk that the statistical conclusions will not reflect the population relationship.
- The greater the sample size, the more time and money we will spend on collecting the samples.



Sample Size Calculation

General sample size formula for continuous data

$$n = \left(\frac{\left(Z_{\alpha/2} + Z_{\beta}\right) \times s}{d}\right)^{2}$$

General sample size formula for discrete data

$$n = \left(\frac{Z_{\alpha/2} + Z_{\beta}}{d}\right)^{2} \times p \times (1-p)$$



Sample Size Calculation

- *n* is the number of observations in the sample.
- α is the risk of committing a false positive error.
- β is the risk of committing a false negative error.
- s is the estimation of standard deviation in the population
- *d* is the size of effect you want to be able to detect.
- *p* is the proportion of one type of event occurring (e.g. proportion of passes).



- Case study:
 - We are interested in comparing the average retail price of a product between two states.
 - We will run a hypothesis test on the two sample means to determine whether there is a statistically significant difference between the retail price in the two states.
 - The average retail price of the product is \$23 based on our estimation and the standard deviation is 3. We want to detect at least a \$2 difference with a 90% chance when it is true and we can tolerate the alpha risk at 5%
 - What should the sample size be?



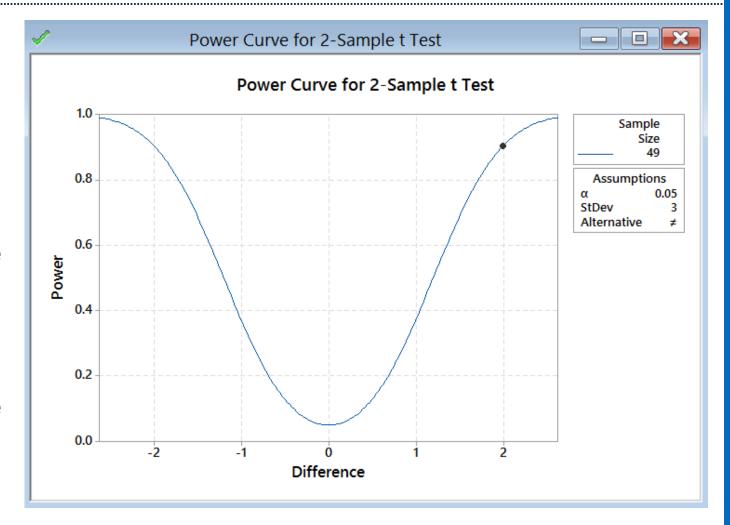
- Steps to calculate the sample size in Minitab
 - 1) Click Stat → Power and Sample Size → 2-Sample t.
 - 2) A window named "Power and Sample Size for 2-Sample t" appears.
 - 3) Enter "2" as "Differences."
 - 4) Enter "0.9" as "Power values."
 - 5) Enter "3" as "Standard deviation."
 - 6) Click "OK."
 - 7) The sample size calculation results appear in a new window.



Power and Sample Size for 2-Sample t	
Specify values for any two of the following: Sample sizes:	
Differences: 2	
Power values: 0.	9
Standard deviation: 3	
	Options Graph
Help	OK Cancel

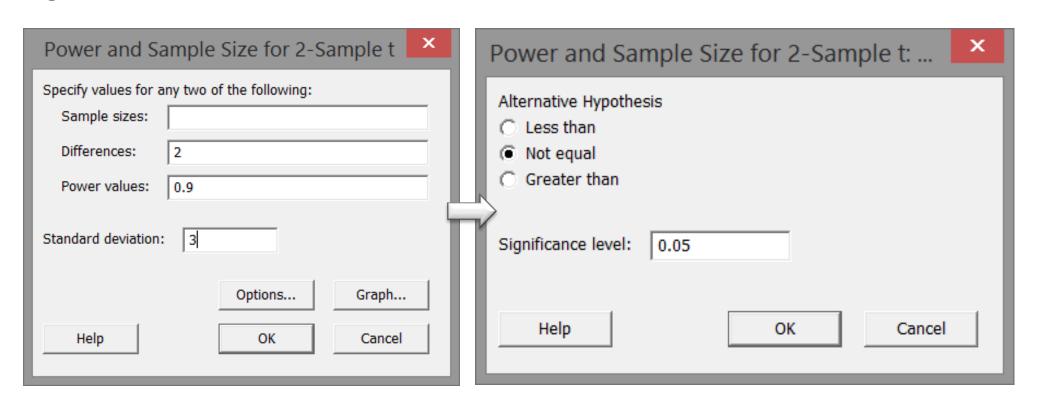


- The sample size for each group is 49 based on the sample size calculator.
- When the absolute value of the difference to detect decreases, the power decreases accordingly if the sample size is kept the same.





- 0.05 is the significance level by default.
- To customize the significance level, click on the "Options" button in the window "Power and Sample Size for 2-Sample t," specify the significance value in the box, and click "OK."





3.1.4 Central Limit Theorem



What is Central Limit Theorem?

- The Central Limit Theorem is one of the fundamental theorems of probability theory.
- It states a condition under which the mean of a large number of independent and identically-distributed random variables, each of which has a finite mean and variance, would be approximately normally distributed.



What is Central Limit Theorem?

- Let us assume $Y_1, Y_2, ..., Y_n$ is a sequence of n i.i.d. random variables, each of which has finite mean μ and variance σ^2 , where $\sigma^2 > 0$.
- When n increases, the sample average of the n random variables is approximately normally distributed, with the mean equal to μ and variance equal to σ^2/n , regardless of the common distribution Y_i follows where i = 1, 2, ..., n.



Independent and Identically Distributed

- A sequence of random variables is **independent and identically distributed** (i.i.d.) if each random variable is independent of others and has the same probability distribution as others.
- It is one of the basic assumptions in Central Limit Theorem.



Central Limit Theorem Example



- Let us assume we have 10 fair die at hand.
- Each time we roll all 10 die together we record the average of the 10 die.
- We repeat rolling the die 50 times until we will have 50 data points.
- Upon doing so, we will discover that the probability distribution of the sample average approximates the normal distribution even though a single roll of a fair die follows a discrete uniform distribution.



Central Limit Theorem Explained in Formulas

• Let us assume Y₁, Y₂, ..., Y_n are i.i.d. random variables with

$$E(Y_i) = \mu_Y$$
 where $-\infty < \mu_Y < \infty$ $var(Y_i) = \sigma_Y^2$ where $0 < \sigma_Y^2 < \infty$

As $n \to \infty$, the distribution of \overline{Y} becomes approximately normally distributed with

$$E(\overline{Y}) = \mu_{Y}$$

$$\operatorname{var}(\overline{Y}) = \sigma_{\overline{Y}}^{2} = \frac{\sigma_{Y}^{2}}{n}$$



Central Limit Theorem Application

Use the sample mean to estimate the population mean.

If the assumptions of Central Limit Theorem are met,

$$E(Y_i) = E(\overline{Y})$$
 where $i = 1, 2, ..., n$



Central Limit Theorem Application

 Use standard error of the mean to measure the standard deviation of the sample mean estimate of a population mean.

$$SE_{\overline{Y}} = \frac{S}{\sqrt{n}}$$

where *s* is the standard deviation of the sample and *n* is the sample size.

Standard deviation of the population mean

$$SD_{\overline{Y}} = \frac{\sigma}{\sqrt{n}}$$

where σ is the standard deviation of the population and n is the sample size.



Central Limit Theorem Application

- Use a larger sample size, if economically feasible, to decrease the variance of the sampling distribution.
- The larger the sample size, the more precise the estimation of the population parameter.
- Use a confidence interval to describe the region which the population parameter would fall in.
- The sample distribution approximates the normal distribution in which 95% of the data stays within two standard deviations from the center.
- Population mean would fall in the interval of two standard errors of the mean away from the sample mean, 95% of the time.



Confidence Interval

- The **confidence interval** is an interval where the true population parameter would fall within a certain confidence level.
- A 95% confidence interval indicates that the population parameter would fall in that region 95% of the time or we are 95% confident that the population parameter would fall in that region.
- 95% is the most commonly used confidence level.
- Confidence interval is used to describe the reliability of a statistical estimate of a population parameter.



Confidence Interval

- The width of a confidence interval depends on:
 - Confidence level
 - Sample size
 - Variability in the data.
- The higher the confidence level, the wider the confidence interval.
- The smaller the sample size, the wider the confidence interval.
- The more variability, the wider the confidence interval.



Confidence Interval of the Mean

• Confidence interval of the population mean μ_Y of a continuous variable Y is

$$\left[\overline{Y} - Z_{\alpha/2} \times \left(\frac{\sigma}{\sqrt{n}}\right), \overline{Y} + Z_{\alpha/2} \times \left(\frac{\sigma}{\sqrt{n}}\right)\right]$$

where

 \overline{Y} is the sample mean

 σ is the standard deviation of the population

n is the sample size

 $Z_{\alpha/2}$ is the Z score when risk level is $\alpha/2$.

- When α is 0.05, it is 1.96.
- When α is 0.10, it is 1.65.

 α is (1 – confidence level).

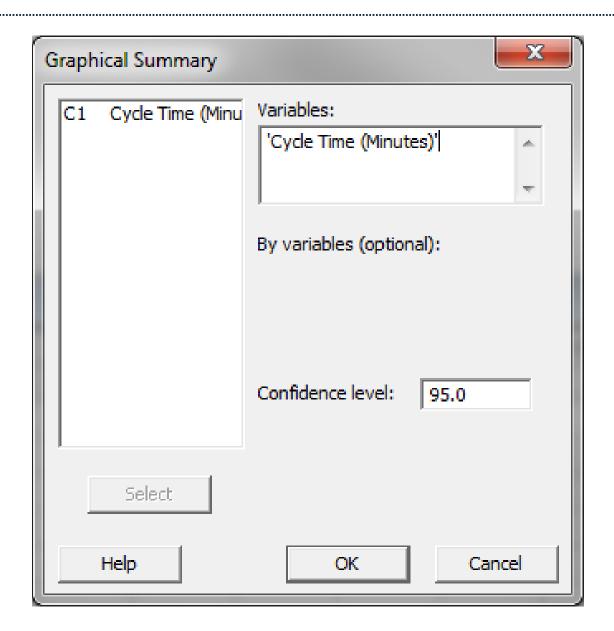
- When confidence level is 95%, α is 5%.
- When the confidence level is 90%, α is 10%.

Special Note: Since 95% is the most commonly used confidence level, 0.05 is the most commonly used α (also called alpha level).



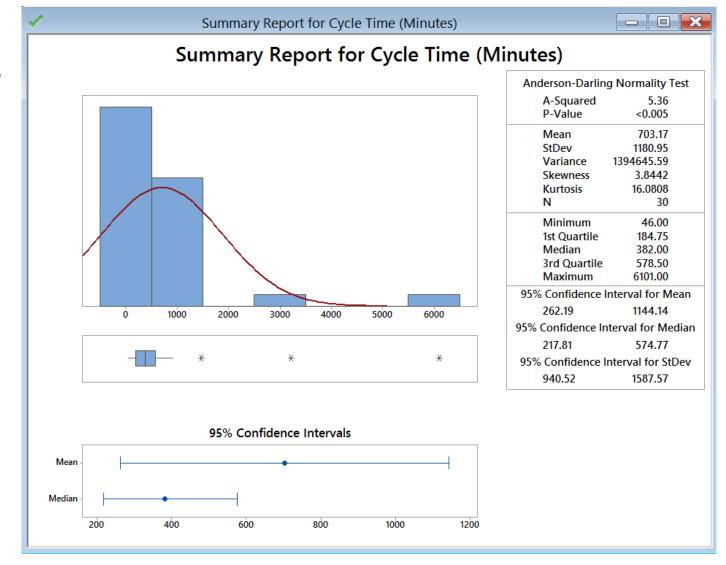
- Data File: "Central Limit Theorem" tab in "Sample Data.xlsx"
 - 1) Click Stat → Basic Statistics → Graphical Summary.
 - 2) A new window named "Graphical Summary" pops up.
 - 3) Select "Cycle Time (Minutes)" as the variable.
 - 4) The confidence level is 0.95 by default.
 - 5) Click "OK."
 - 6) A new window named "Summary for Cycle Time (Minutes)" pops up.







The 95% confidence interval of the mean is shown in the newly-generated "Summary for Cycle Time (Minutes)."



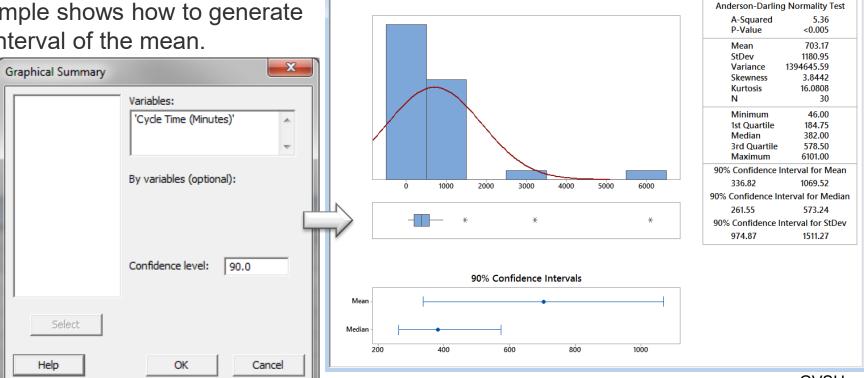


The confidence level is 95% by default.

 In order to see the confidence interval of "Cycle Time (Minutes)" at other confidence levels, we need to enter the confidence level of our interest in the window "Graphical Summary"

and click "OK."

 The following example shows how to generate 90% confidence interval of the mean.



Summary Report for Cycle Time (Minutes)

Summary Report for Cycle Time (Minutes)



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- - X

3.2 Hypothesis Testing



Black Belt Training: Analyze Phase

3.1 Inferential Statistics

- 3.1.1 Understanding Inference
- 3.1.2 Sampling Techniques and Uses
- 3.1.3 Sample Size
- 3.1.4 Central Limit Theorem

3.2 Hypothesis Testing

- 3.2.1 Goals of Hypothesis Testing
- 3.2.2 Statistical Significance
- 3.2.3 Risk; Alpha and Beta
- 3.2.4 Types of Hypothesis Tests

3.3 Hypothesis Testing: Normal Data

- 3.3.1 One and Two Sample T-Tests
- 3.3.2 One sample variance
- 3.3.3 One Way ANOVA

3.4 Hypothesis Testing: Non-Normal Data

- 3.4.1 Mann-Whitney
- 3.4.2 Kruskal-Wallis
- 3.4.3 Moods Median
- 3.4.4 Friedman
- 3.4.5 One Sample Sign
- 3.4.6 One Sample Wilcoxon
- 3.4.7 One and Two Sample Proportion
- 3.4.8 Chi-Squared (Contingency Tables)
- 3.4.9 Test of Equal Variances



3.2.1 Goals of Hypothesis Testing



What is Hypothesis Testing?

- A **hypothesis test** is a statistical method in which a specific hypothesis is formulated about a population, and the decision of whether to reject the hypothesis is made based on sample data.
- Hypothesis tests help to determine whether a hypothesis about a population or multiple populations is true with certain confidence level based on sample data.



Hypothesis Testing Examples

- Hypothesis testing tries to answer whether there is a difference between different groups or there is some change occurring.
 - Are the average SAT scores of graduates from high school A and B the same?
 - Is the error rate of one group of operators higher than that of another group?
 - Are there any non-random causes influencing the height of kids in one state?



What is Statistical Hypothesis?

- A statistical hypothesis is an assumption about one or multiple population.
- It is a statement about whether there is any difference between different groups.
- It can be a conjecture about the population parameters or the nature of the population distributions.
- A statistical hypothesis is formulated in pairs:
 - Null Hypothesis
 - Alternative Hypothesis.



Null and Alternative Hypotheses

- *Null Hypothesis* (H₀) states that:
 - there is no difference in the measurement of different groups
 - no changes occurred
 - sample observations result from random chance.
- Alternative Hypothesis (H₁ or H_a) states that:
 - there is a difference in the measurement of different groups
 - some changes occurred
 - sample observations are affected by non-random causes.



Null and Alternative Hypotheses

 A statistical hypothesis can be expressed in mathematical language by using population parameters (Greek letters) and mathematical symbols.

Population Parameters (Greek letters)

- Mean: μ
- Standard deviation: σ
- Variance: σ²
- Median: η

Mathematical Symbols

- Equal: =
- Not equal: ≠
- Greater than: >
- Smaller than: <



Null and Alternative Hypotheses

 Examples of null and alternative hypotheses written in mathematical language.

$$\begin{cases} H_0: \mu_1 = \mu_2 \\ H_1: \mu_1 \neq \mu_2 \end{cases}$$

$$\begin{cases} H_0: \sigma_1 = 0 \\ H_1: \sigma_1 \neq 0 \end{cases}$$

$$\begin{cases} H_0: \eta_1 = 10 \\ H_1: \eta_1 > 10 \end{cases}$$

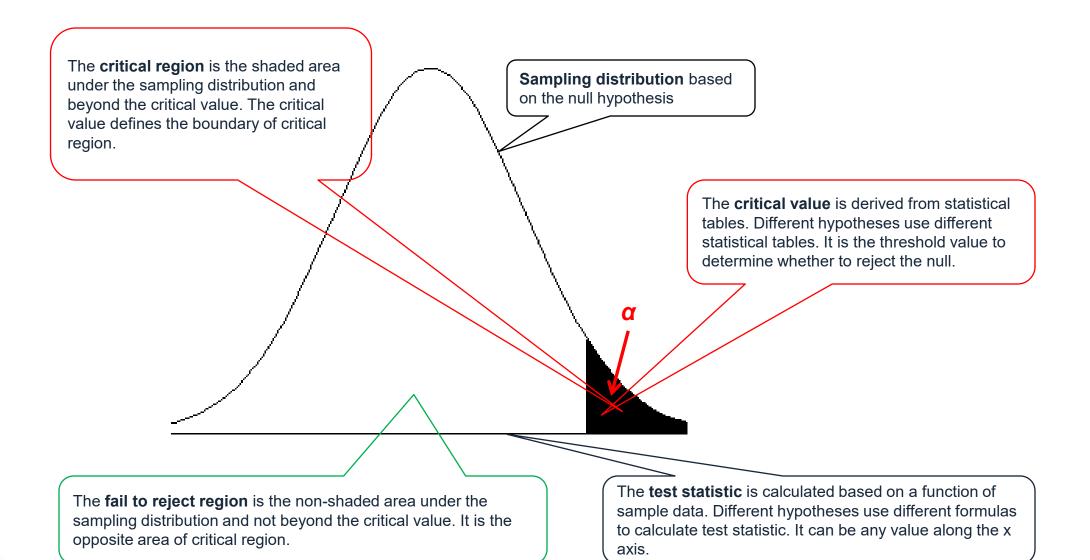
$$\begin{cases} H_0: \mu_1 = 10 \\ H_1: \mu_1 < 10 \end{cases}$$



Hypothesis Testing Conclusion

- There are two possible conclusions of hypothesis testing:
 - Reject the null
 - Fail to reject the null.
- When there is enough evidence based on the sample information to prove the alternative hypothesis, we reject the null.
- When there is *not* enough evidence or the sample information is *not* sufficiently persuasive, we fail to reject the null.







- The test statistic in hypothesis testing is a value calculated using a function of the sample.
- Test statistics are considered the sample data's numerical summary that can be used in hypothesis testing.
- Different hypothesis tests have different formulas to calculate the test statistic.
- The critical value in hypothesis testing is a threshold value to which the test statistic is compared in order to determine whether the null hypothesis is rejected.
- The critical value is obtained from statistical tables.
- Different hypothesis tests need different statistical tables for critical values.



- When the test statistic falls into the fail to reject region, we fail to reject the null and claim that there is no statistically significant difference between the groups.
- When the test statistic falls into the critical region, we reject the null and claim that there is a statistically significant difference between the groups.



- The proportion of the area under the sampling distribution and beyond the critical value indicates α risk (also called *α level*). The most commonly selected α level is 5%.
- The proportion of the area under the sampling distribution and beyond the test statistic is the *p-value*. It is the probability of getting a test statistic at least as extreme as the observed one, given the null is true.



- When the p-value is smaller than the α level, we reject the null and claim that there is a statistically significant difference between different groups.
- When the p-value is higher than the α level, we fail to reject the null and claim that there is no statistically significant difference between different groups.



Steps in Hypothesis Testing

- Step 1: State the null and alternative hypothesis.
- Step 2: Determine α level.
- Step 3: Collect sample data.
- Step 4: Select a proper hypothesis test.
- Step 5: Run the hypothesis test.
- Step 6: Determine whether to reject the null.



3.2.2 Statistical Significance



Statistical Significance

- In statistics, an observed difference is *statistically significant* if it is unlikely that the difference occurred by pure chance, given a predetermined probability threshold.
- Statistical significance indicates that there are some non-random factors causing the result to take place.
- The statistical significance level in hypothesis testing indicates the amount of evidence which is sufficiently persuasive to prove that a difference between groups exists not due to random chance alone.



Practical Significance

- An observed difference is *practically significant* when it is large enough to make a practical difference.
- A difference between groups that is *statistically significant* might not be large enough to be practically significant.
- In some business situations, statistical differences can have little to no meaning because the difference is not large enough to be practical for a business to act upon.



Example

- You started to use the premium gas recently, which was supposed to make your car run better.
- After running a controlled experiment to measure the performance of the car before and after using the premium gas, you performed a statistical hypothesis test and found that the difference before and after was statistically significant.
- Using premium gas did improve the performance.



• However, due to the high cost of the premium gas, you decided that the difference was not large enough to make you pay extra money for it. In other words, the difference is not practically significant.



3.2.3 Risk; Alpha & Beta



Errors in Hypothesis Testing

- In statistical hypothesis testing, there are two types of errors:
 - Type I Error
 - a null hypothesis is rejected when it is true in fact.
 - Type II Error
 - a null hypothesis is not rejected when it is not true in fact.

	Null hypothesis is true	Alternative hypothesis is true
Fail to reject null hypothesis	Correct	Incorrect (Type II Error)
Reject null hypothesis	Incorrect (Type I Error)	Correct



Type I Error

- Type I error is also called false positive, false alarm, or alpha (α) error.
- Type I error is associated with the risk of accepting false positives.
- It occurs when we think there is a difference between groups but in fact there is none.
- Example: telling a patient he is sick and in fact he is not.



Alpha (α)

- α indicates the probability of making a type I error. It ranges from 0 to 1.
- α risk is the risk of making a type I error.
- 5% is the most commonly used α .
- $(100\% \alpha)$ is the confidence level which is used to calculate the confidence intervals.
- When making a decision on whether to reject the null, we compare the p-value against α:
 - If p-value is smaller than α , we reject the null
 - If p-value is greater than α , we fail to reject the null.
- To reduce the α risk, we decrease the α value to which the p-value is compared.



Type II Error

- Type II error is also called false negative, oversight, or beta (β) error.
- Type II error is associated with the risk of accepting false negatives.
- It occurs when we think there is not any difference between groups but in fact there is.
- Example: telling a patient he is not sick and in fact he is.



Beta (β)

- β indicates the probability of making a type II error. It ranges from 0 to 1.
- β risk is the risk of making a type II error.
- 10% is the most commonly used β.
- $(100\% \beta)$ is called *power*, which denotes the probability of detecting a difference between groups when in fact the difference truly exists.
- To reduce the β risk, we increase the sample size.
- When holding other factors constant, β is inversely related to α .



3.2.4 Types of Hypothesis Tests



Two Types of Hypothesis Tests

Two-tailed hypothesis test

- It is also called two-sided hypothesis test
- It is a statistical hypothesis test in which the critical region is split into two equal areas, each of which stays on one side of the sampling distribution.

One-tailed hypothesis test

- It is also called one-sided hypothesis test
- It is a statistical hypothesis test in which the critical region is only on one side of the sampling distribution.



Two-Tailed Hypothesis Test

- A two-tailed hypothesis test is used when we care about whether there
 is a difference between groups and we do not care about the direction
 of the difference.
- Examples of a two-tailed hypothesis test:

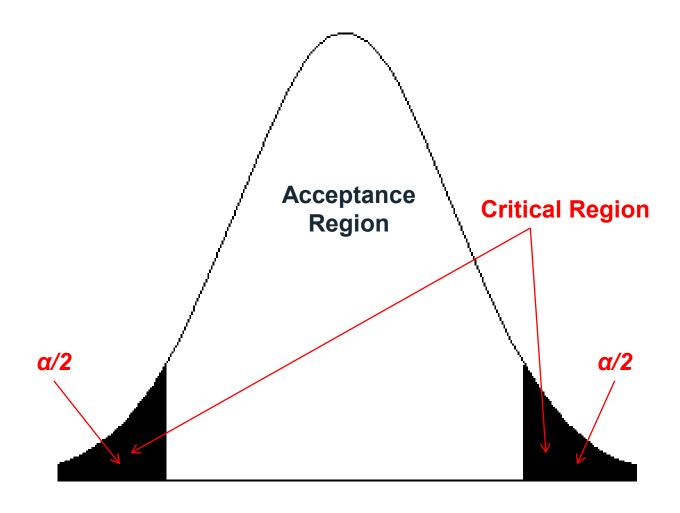
$$H_0$$
: $\mu_1 = 10$
 H_a : $\mu_1 \neq 10$

The null hypothesis (H₀) is rejected when:

- the test statistic falls into either half of the critical region
- the test statistic is either sufficiently small or sufficiently large
- the absolute value of the test statistic is greater than the absolute value of the critical value.
- A two-tailed hypothesis test is the most commonly used hypothesis test. In the next modules we will cover more details about it.



Two-Tailed Hypothesis Test



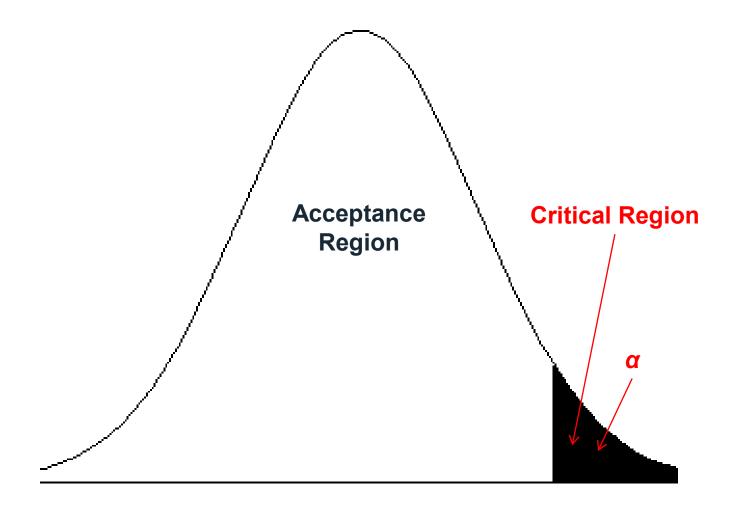


- A one-tailed hypothesis test is used when we care about one direction of the difference between groups.
- Examples of one-tailed hypothesis tests:

$$H_0$$
: $\mu_1 = 10$
 H_a : $\mu_1 > 10$

- The null hypothesis is rejected when the test statistic:
 - falls in the critical region which exists on the right side of the sampling distribution
 - is sufficiently large
 - is greater than the critical value.



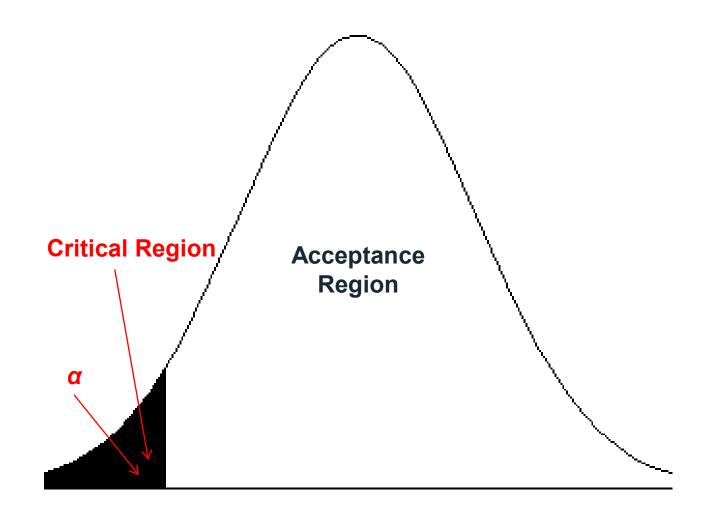




Examples of one-tailed hypothesis test

$$H_0$$
: $\mu_1 = 10$
 H_a : $\mu_1 < 10$

- The null hypothesis is rejected when the test statistic:
 - falls into the critical region which only exists on the left side of the sampling distribution
 - is sufficiently small
 - is smaller than the critical value.





3.3 Hypothesis Tests: Normal Data



Black Belt Training: Analyze Phase

3.1 Inferential Statistics

- 3.1.1 Understanding Inference
- 3.1.2 Sampling Techniques and Uses
- 3.1.3 Sample Size
- 3.1.4 Central Limit Theorem

3.2 Hypothesis Testing

- 3.2.1 Goals of Hypothesis Testing
- 3.2.2 Statistical Significance
- 3.2.3 Risk; Alpha and Beta
- 3.2.4 Types of Hypothesis Tests

3.3 Hypothesis Testing: Normal Data

- 3.3.1 One and Two Sample T-Tests
- 3.3.2 One sample variance
- 3.3.3 One Way ANOVA

3.4 Hypothesis Testing: Non-Normal Data

- 3.4.1 Mann-Whitney
- 3.4.2 Kruskal-Wallis
- 3.4.3 Moods Median
- 3.4.4 Friedman
- 3.4.5 One Sample Sign
- 3.4.6 One Sample Wilcoxon
- 3.4.7 One and Two Sample Proportion
- 3.4.8 Chi-Squared (Contingency Tables)
- 3.4.9 Test of Equal Variances



3.3.1 One & Two Sample T-Tests



What is a T-Test?

- In statistics, a **t-test** is a hypothesis test in which the test statistic follows a *Student t* distribution if the null hypothesis is true.
- We apply a t-test when the population variance (σ) is unknown and we use the sample standard deviation (s) instead.



What is One Sample T-Test?

- One sample t-test is a hypothesis test to study whether there is a statistically significant difference between a population mean and a specified value.
 - Null Hypothesis (H_0): $\mu = \mu_0$
 - Alternative Hypothesis (H_a): $\mu \neq \mu_0$

where μ is the mean of a population of our interest and μ_0 is the specific value we want to compare against.



Assumptions of One Sample T-Test

- The sample data drawn from the population of interest are unbiased and representative.
- The data of the population are continuous.
- The data of the population are normally distributed.
- The variance of the population of our interest is unknown.
- One sample t-test is more robust than the z-test when the sample size is small (< 30).



Normality Test

- To check whether the population of our interest is normally distributed, we need to run normality test.
 - Null Hypothesis (H₀): The data are normally distributed.
 - Alternative Hypothesis (H_a): The data are not normally distributed.
- There are a lot of normality tests available:
 - Anderson-Darling
 - Sharpiro-Wilk
 - Jarque-Bera etc.



Test Statistic and Critical Value of One Sample T-Test

Test Statistic

$$t_{calc} = \frac{\overline{Y}}{\sqrt{N}}$$
 , where

 \overline{Y} is the sample mean, n is the sample size, and s is the sample standard deviation

$$s = \sqrt{\frac{\sum_{i=1}^{n} (Y_i - \overline{Y})^2}{n-1}}$$

- Critical Value
 - t_{crit} is the t-value in a Student t distribution with the predetermined significance level α and degrees of freedom (n –1).
 - t_{crit} values for a two-sided and a one-sided hypothesis test with the same significance level α and degrees of freedom (n-1) are different.



Decision Rules of One Sample T-Test

- Based on the sample data, we calculated the test statistic t_{calc}, which is compared against t_{crit} to make a decision of whether to reject the null.
 - Null Hypothesis (H_0): $\mu = \mu_0$
 - Alternative Hypothesis (H_a): $\mu \neq \mu_0$
- If $|t_{calc}| > t_{crit}$, we reject the null and claim there is a statistically significant difference between the population mean μ and the specified value μ_0 .
- If $|t_{calc}| < t_{crit}$, we fail to reject the null and claim there is not any statistically significant difference between the population mean μ and the specified value μ_0 .

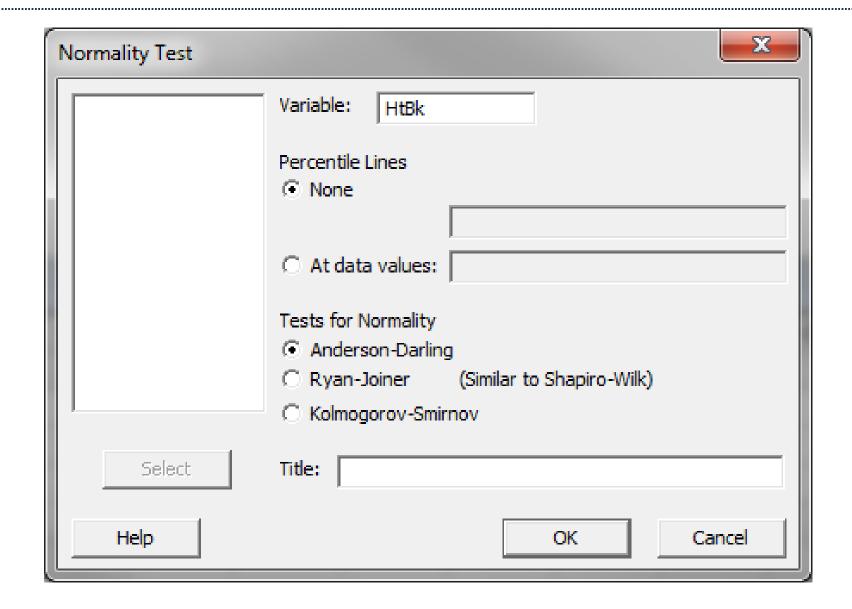


- Case study: We want to compare the average height of basketball players against 7 feet.
 - Data File: "One Sample T-Test" tab in "Sample Data.xlsx"
 - Null Hypothesis (H_0): $\mu = 7$
 - Alternative Hypothesis (H_a): µ ≠ 7



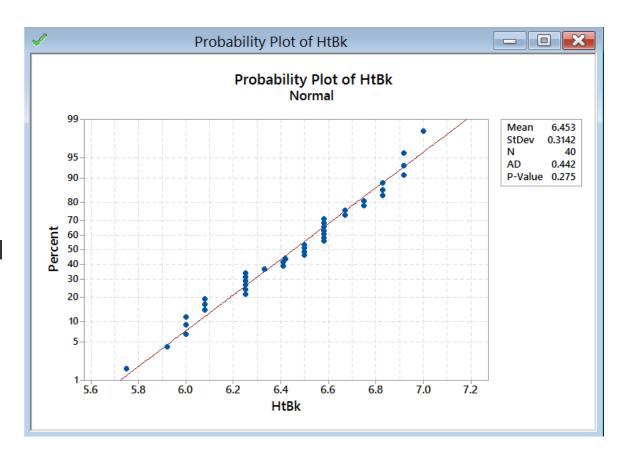
- Step 1: Test whether the data are normally distributed
 - Click Stat → Basic Statistics → Normality Test.
 - 2) A new window named "Normality Test" pops up.
 - 3) Select "HtBk" as the variable.
 - 4) Click "OK."
 - 5) A new window named "Probability Plot of HtBk" appears, which covers the results of the normality test.







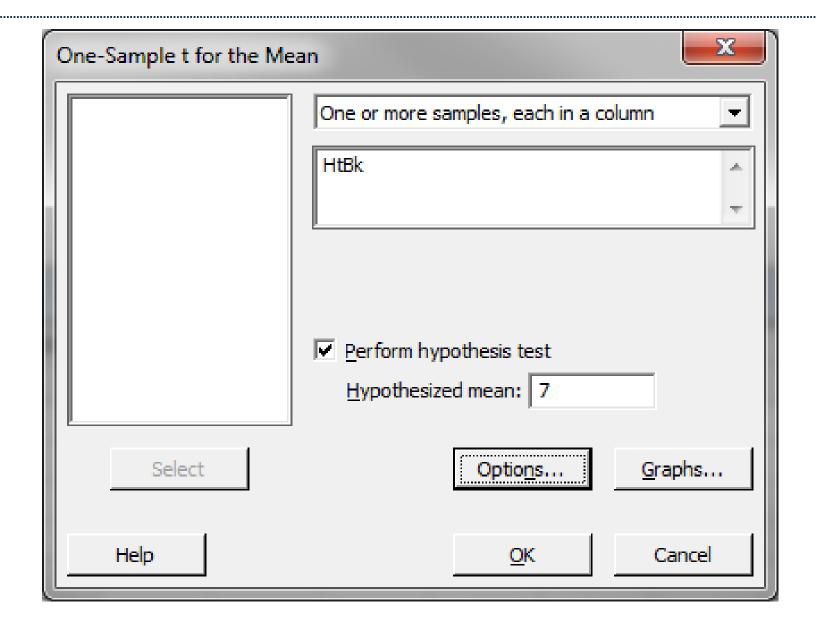
- Null Hypothesis (H₀): The data are normally distributed.
- Alternative Hypothesis (H_a): The data are not normally distributed.
- Since the p-value of the normality is 0.275, which is greater than alpha level (0.05), we fail to reject the null and claim that the data are normally distributed.
- If the data are not normally distributed, you need to use hypothesis tests other than the one sample t-test.





- Step 2: Run the one-sample t-test
 - 1) Click Stat → Basic Statistics → 1 Sample t.
 - 2) A new window named "1 Sample t (Test and Confidence Interval)" pops up.
 - 3) Click in the blank box under "Samples in columns" and "HtBk" appears in the list box on the left.
 - 4) Select "HtBk" as the "Samples in columns."
 - 5) Check the box of "Perform hypothesis test."
 - 6) Enter the hypothesized value "7" into the box next to "Perform hypothesis test."
 - 7) Click "OK."
 - 8) The one-sample t-test result appears automatically in the session window.







- Null Hypothesis (H_0): $\mu = 7$
- Alternative Hypothesis (H_a): µ ≠ 7
- Since the p-value is smaller than alpha level (0.05), we reject the null hypothesis and claim that average of basketball players is statistically different from 7 feet.

One-Sample T: HtBk

Descriptive Statistics

N	Mean	StDev	SE Mean	95% CI for μ
40	6.4532	0.3142	0.0497	(6.3528, 6.5537)

μ: mean of HtBk

Test

Null hypothesis H_0 : $\mu = 7$ Alternative hypothesis H_1 : $\mu \neq 7$

T-Value P-Value -11.00 0.000



What is Two Sample T-Test?

- **Two sample t-test** is a hypothesis test to study whether there is a statistically significant difference between the means of two populations.
 - Null Hypothesis (H_0): $\mu_1 = \mu_2$
 - Alternative Hypothesis (H_a): μ₁ ≠ μ₂

where μ_1 is the mean of one population and μ_2 is the mean of the other population of our interest.



Assumptions of Two Sample T-Tests

- The sample data drawn from both populations are unbiased and representative.
- The data of both populations are continuous.
- The data of both populations are normally distributed.
- The variances of both populations are unknown.
- Two sample t-test is more robust than a z-test when the sample size is small (< 30).



Three Types of Two Sample T-Tests

- 1. Two sample t-test when the variances of two populations are unknown but equal
 - Two sample t-test (when $\sigma_1^2 = \sigma_2^2$)
- 2. Two sample t-test when the variances of the two population are unknown and unequal
 - Two sample t-test (when $\sigma_1^2 \neq \sigma_2^2$)
- 3. Paired t-test when the two populations are dependent of each other



• To check whether the variances of two populations of interest are statistically significant different, we use the test of equal variance.

- Null Hypothesis $(H_0)\sigma_1^2 = \sigma_2^2$
- Alternative Hypothesis $(H_1)\sigma_1^2 \neq \sigma_2^2$
- An F-test is used to test the equality of variances between two normally distributed populations.



- An **F-test** is a statistic hypothesis test in which the test statistic follows an F-distribution when the null hypothesis is true.
- The most known F-test is the test of equal variance for two normally distributed populations.
- The F-test is very sensitive to non-normality. When any one of the two
 populations is not normal, we use the Brown-Forsythe test for checking
 the equality of variances.



Test Statistic

$$F_{calc} = \frac{S_1^2}{S_2^2}$$

where

s₁ and s₂ are the sample standard deviations.

Critical Value

- F_{crit} is the F value in a F distribution with the predetermined significance level α and degrees of freedom $(n_1 1)$ and $(n_2 1)$.
- F_{crit} values for a two-sided and a one-sided F-test with the same significance level α and degrees of freedom $(n_1 1)$ and $(n_2 1)$ are different.



- Based on the sample data, we calculated the test statistic F_{calc} , which is compared against F_{crit} to make a decision of whether to reject the null.
 - Null Hypothesis (H₀): $\sigma_1^2 = \sigma_2^2$
 - Alternative Hypothesis (H_a): $\sigma_1^2 \neq \sigma_2^2$
- If F_{calc} > F_{crit}, we reject the null and claim there is a statistically significant difference between the variances of the two populations.
- If F_{calc} < F_{crit}, we fail to reject the null and claim there is not any statistically significant difference between the variances of the two populations.



Test Statistic & Critical Value of a Two Sample T-Test when $\sigma_1^2 = \sigma_2^2$

Test Statistic

$$t_{calc} = \frac{\overline{Y}_1 - \overline{Y}_2}{S_{pooled} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} \qquad s_{pooled} = \sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{(n_1 + n_2 - 2)}}$$

where

 \overline{Y}_1 and \overline{Y}_2 are the sample means of the two populations of our interest. n_1 and n_2 are the sample sizes. n_1 is not necessarily equal to n_2 . s_{pooled} is a pooled estimate of variance. s_1 and s_2 are the sample standard deviations.

Critical Value

 t_{crit} is the t value in a Student t distribution with the predetermined significance level α and degrees of freedom $(n_1 + n_2 - 2)$.

 t_{crit} values for a two-sided and a one-sided t-test with the same significance level α and degrees of freedom $(n_1 + n_2 - 2)$ are different.



Test Statistic & Critical Value of a Two Sample T-Test when $\sigma_1^2 \neq \sigma_2^2$

Test Statistic

$$t_{calc} = \frac{\overline{Y}_{1} - \overline{Y}_{2}}{\frac{S_{1}^{2}}{n_{1}} + \frac{S_{2}^{2}}{n_{2}}}$$

$$df = \frac{\left(\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}\right)^2}{\frac{\left(\frac{S_1}{n_1}\right)^2}{n_1 - 1} + \frac{\left(\frac{S_2}{n_2}\right)^2}{n_2 - 1}}$$

where

 \overline{Y}_1 and \overline{Y}_2 are the sample means of the two populations of our interest.

 n_1 and n_2 are the sample sizes. n_1 is not necessarily equal to n_2 .

 s_1 and s_2 are the sample standard deviations.

Critical Value

 t_{crit} is the t value in a Student t distribution with the predetermined significance level α and degrees of freedom df calculated using the formula above.

 t_{crit} values for a two-sided and a one-sided t-test with the same significance level α and degrees of freedom *df are different*.



Test Statistic & Critical Value of a Paired T-Test

Test Statistic

$$t_{calc} = \frac{\overline{d}}{S_d / \sqrt{n}}$$

where

d is the difference between each pair of data.

d is the average of d.

n is the sample size of either population of interest.

 s_d is standard deviation of d.

Critical Value

 t_{crit} is the t value in a Student t distribution with the predetermined significance level α and degrees of freedom (n-1).

 t_{crit} values for a two-sided and a one-sided t-test with the same significance level α and degrees of freedom (n-1) are different.

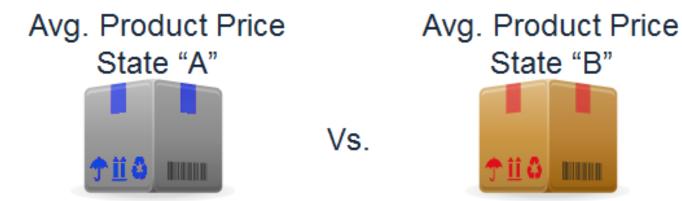


Decision Rules of a Two Sample T-Test

- Based on the sample data, we calculated the test statistic t_{calc}, which is compared against t_{crit} to make a decision of whether to reject the null.
 - Null Hypothesis (H_0): $\mu_1 = \mu_2$
 - Alternative Hypothesis (H_a): μ₁ ≠ μ₂
- If $|t_{calc}| > t_{crit}$, we reject the null and claim there is a statistically significant difference between the means of the two populations.
- If |t_{calc}| < t_{crit}, we fail to reject the null and claim there is not any statistically significant difference between the means of the two populations.



- Case study: We are trying to compare the average retail price of a product in state A and state B.
 - Data File: "Two-Sample T-Test" tab in "Sample Data.xlsx"

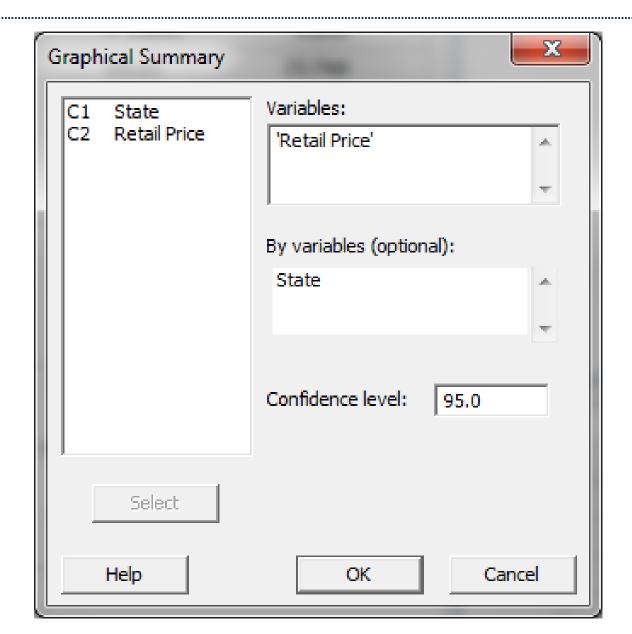


- Null Hypothesis (H_0): $\mu_1 = \mu_2$
- Alternative Hypothesis (H_a): $\mu_1 \neq \mu_2$



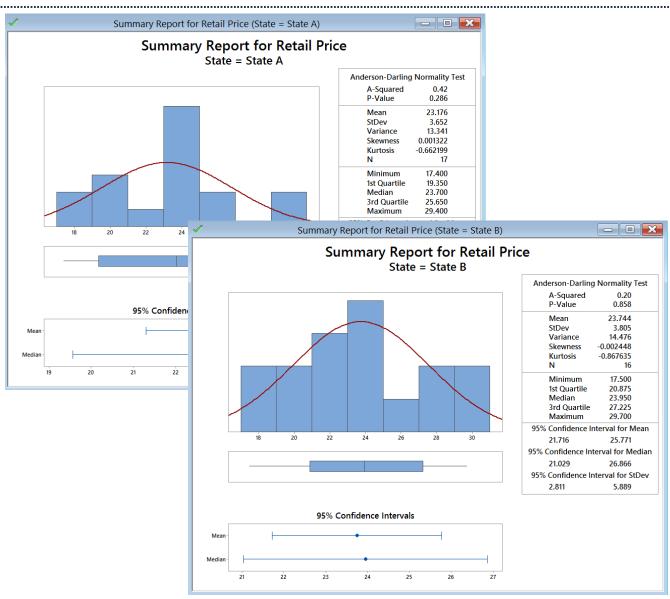
- Step 1: Test the normality of the retail price for both state A and B.
 - 1) Click Stat → Basic Statistics → Graphical Summary.
 - 2) A window named "Graphical Summary" pops up.
 - 3) Select "Retail Price" as the "Variables."
 - 4) Click in the blank box right below "By variables (optional)" and "State" appears in the list box on the left.
 - 5) Select the "State" as the "By variables (optional)."
 - 6) Click "OK."
 - 7) The normality test results would appear in the new windows.







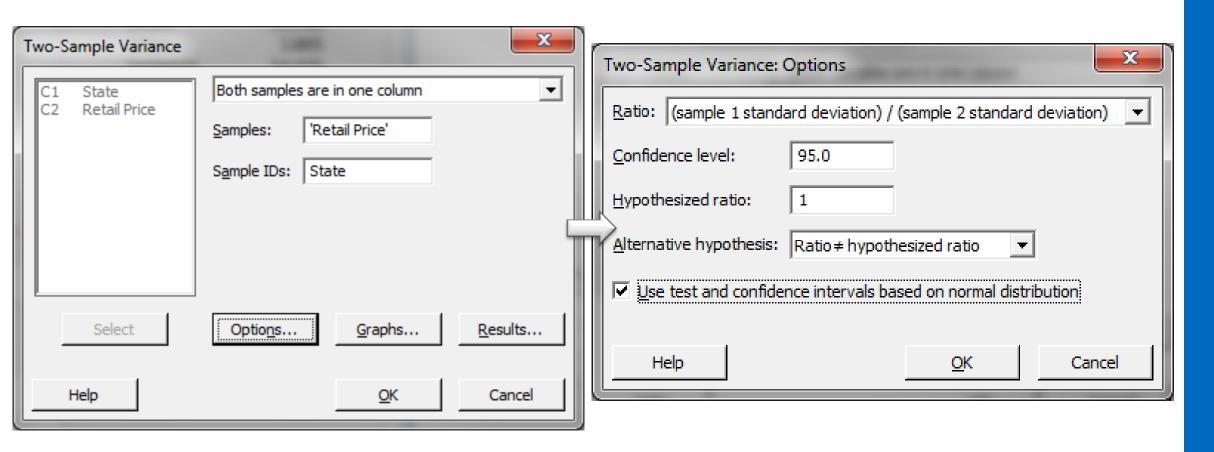
- Null Hypothesis (H0): The data are normally distributed.
- Alternative Hypothesis (Ha): The data are not normally distributed.
- Both retail price data of state A and B are normally distributed since the p-values are both greater than alpha level (0.05).
- If any of the data series is not normally distributed, we need to use other hypothesis testing methods other than the two sample t-test.





- Step 2: Test whether the variances of the two data sets are equal.
 - Null Hypothesis (H₀): $\sigma_1^2 = \sigma_2^2$
 - Alternative Hypothesis (H_a): $\sigma_1^2 \neq \sigma_2^2$
 - 1) Click Basic Statistics → 2 Variances.
 - 2) A new window named "2 Sample Variance" pops up.
 - 3) Click in the blank box of "Samples" and "Retail Price" appears in the list box on the left.
 - 4) Select "Retail Price" as the "Samples."
 - 5) Click in the blank box of "Sample ID" and "State" appears in the list box on the left.
 - 6) Select "State" as the "Sample ID."
 - 7) Click "Options" and select "Use test and confidence intervals based on normal distribution"
 - 8) Click "OK." to the "Options" dialogue box
 - 9) Click "OK." to the 2 Variances (Test Confidence Intervals) window.
 - 10) The results show up in the session window.







- Because the retail prices at state A and state B are both normally distributed, an F test is used to test their variance equality.
- The p-value of F test is 0.870, greater than the alpha level (0.05), so we fail to reject the null hypothesis and we claim that the variances of the two data sets are equal. We will use the two sample t-test (when $\sigma_1^2 = \sigma_2^2$) to compare the means of the two groups.
- If $\sigma_1^2 \neq \sigma_2^2$, we will use the two sample t-test (when $\sigma_1^2 \neq \sigma_2^2$) to compare the means of the two groups.

Test and CI for Two Variances: Retail Price vs State

Method

 σ_1 : standard deviation of Retail Price when State = State A σ_2 : standard deviation of Retail Price when State = State B Ratio: σ_1/σ_2

F method was used. This method is accurate for normal data only.

Descriptive Statistics

State	Ν	StDev	Variance	95% CI for σ
State A	17	3.652	13.341	(2.720, 5.559)
State B	16	3.805	14.476	(2.811, 5.889)

Ratio of Standard Deviations

95% CI for
Estimated Ratio using
Ratio F

0.959986 (0.570, 1.603)

Test

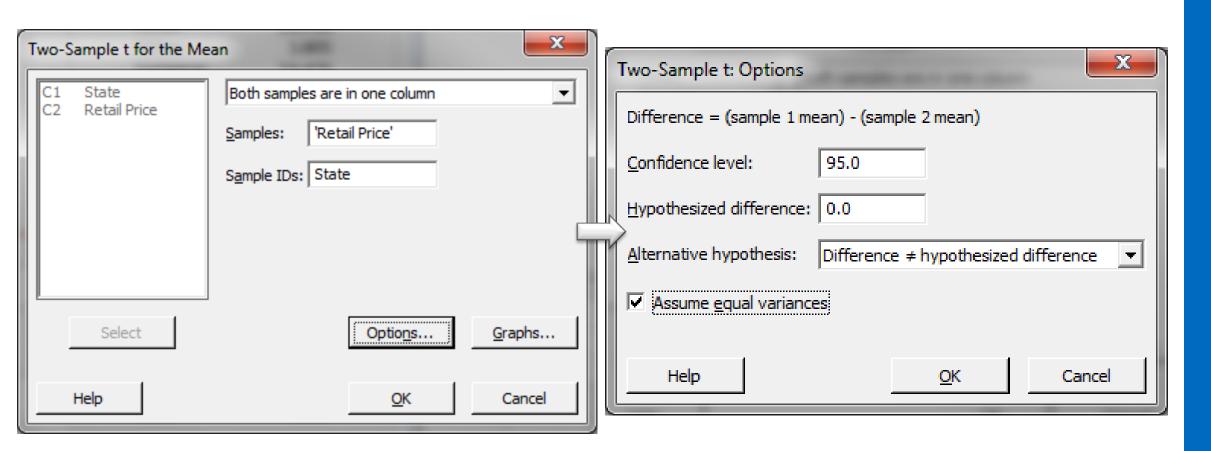
Null hypothesis H_0 : $\sigma_1 / \sigma_2 = 1$ Alternative hypothesis H_1 : $\sigma_1 / \sigma_2 \neq 1$ Significance level $\alpha = 0.05$

	Test			
Method	Statistic	DF1	DF2	P-Value
F	0.92	16	15	0.870



- Step 3: Run two-sample t-test to compare the means of two groups.
 - 1) Click Stat → Basic Statistics → 2-Sample t.
 - 2) A new window named "2-Sample t for the Mean" pops up.
 - 3) Click in the blank box right next to "Samples" and the "Retail Price" appears in the list box on the left.
 - 4) Select "Retail Price" as the "Samples."
 - 5) Click in the blank box right next to "Sample ID" and the "State" appears in the list box on the left.
 - 6) Select "State" as the "Sample ID."
 - 7) Click "Options"
 - 8) Check the box "Assume equal variances."
 - 9) Click "OK." in the "Options" window.
 - 10)Click "OK" in the 2-Sample t window
 - 11)The results of the two-sample t-test (when $\sigma_1 = \sigma_2$) appear automatically in the session window.







Since the p-value of the t-test (assuming equal variance) is 0.665, greater than the alpha level (0.05), we fail to reject the null hypothesis and we claim that the means of the two data sets are equal.

Descriptive Statistics: Retail Price

St	ate	Ν	Mean	StDev	SE Mean
St	ate A	17	23.18	3.65	0.89
St	ate B	16	23.74	3.80	0.95

Estimation for Difference

	Pooled	95% CI for
Difference	StDev	Difference
-0.57	3.73	(-3.21, 2.08)

Test

Null hypothesis H_0 : $\mu_1 - \mu_2 = 0$ Alternative hypothesis H_1 : $\mu_1 - \mu_2 \neq 0$

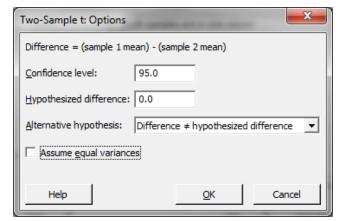
T-Value	DF	P-Value
-0.44	31	0.665

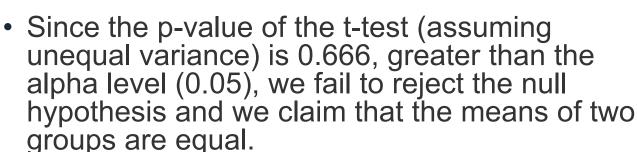


• If the variances of the two groups do not equal, we will need to use the two-sample t-test (when $\sigma_1 \neq \sigma_2$) to compare the means of the two groups.

• In the window of "2-Sample (Test and Confidence Interval)," uncheck the box next to "Assume equal variances" and run the 2-sample t-test

again.





Descriptive Statistics: Retail Price

State	Ν	Mean	StDev	SE Mean
State A	17	23.18	3.65	0.89
State B	16	23.74	3.80	0.95

Estimation for Difference

	95% CI for
Difference	Difference
-0.57	(-3.22, 2.09)

Test

Null hypo	othesi	H_0 : $\mu_1 - \mu_2 = 0$	
Alternativ	/e hy	$H_1: \mu_1 - \mu_2 \neq 0$	
T-Value	DF	P-Value	_
-0.44	30	0.666	



Case study:

- We are interested to know whether the average salaries (\$1000/yr.) of male and female professors at the same university are the same.
 - Data File: "Paired T-Test" tab in "Sample Data.xlsx"
 - The data were randomly collected from 22 universities. For each university, the salaries of a male and female professors were randomly selected.
- Null Hypothesis (H_0): $\mu_{male} \mu_{female} = 0$
- Alternative Hypothesis (H_a): μ_{male} − μ_{female} ≠ 0

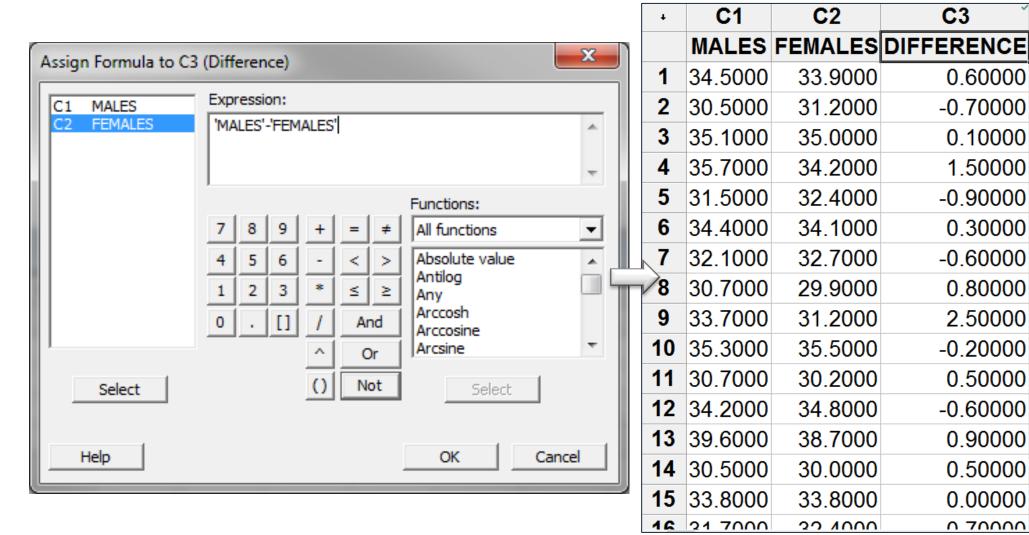


 Step 1: Create a new column of the difference between the two data sets of interest and name it "Difference."

(Note: if your data already has a column for "Difference" with data in it, this step will not be necessary.)

- 1) Right click on the newly-generated column, "Difference."
- 2) Select "Formulas" and then "Assign Formula to Column."
- 3) A new window named "Assign Formula to C3 (Difference)" pops up.
- 4) Enter "'MALES'-' FEMALES' " into the box right below "Expression."
- 5) Click "OK."
- 6) The values would appear in the column "Difference."







C3

0.60000

-0.70000

0.10000

1.50000

-0.90000

0.30000

-0.60000

0.80000

2.50000

-0.20000

0.50000

-0.60000

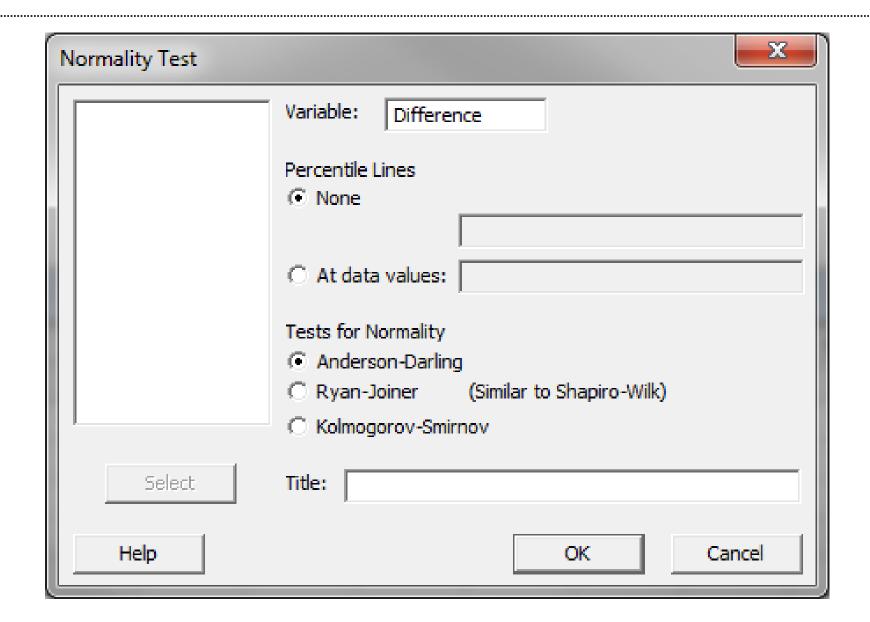
0.90000

0.50000

0.00000 0.70000

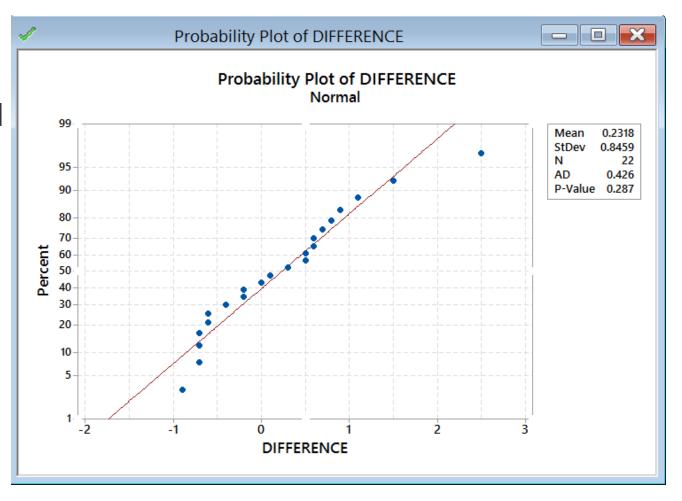
- Step 2: Test whether the difference is normally distributed.
 - Null Hypothesis (H₀): The difference between two data sets is normally distributed.
 - Alternative Hypothesis (H_a): The difference between two data sets is not normally distributed.
 - 1) Click Stat → Basic Statistics → Normality Test.
 - 2) A new window named "Normality Test" pops up.
 - Select "Difference" as the "Variable."
 - 4) Click "OK."
 - 5) The normality test results would appear in the new window.







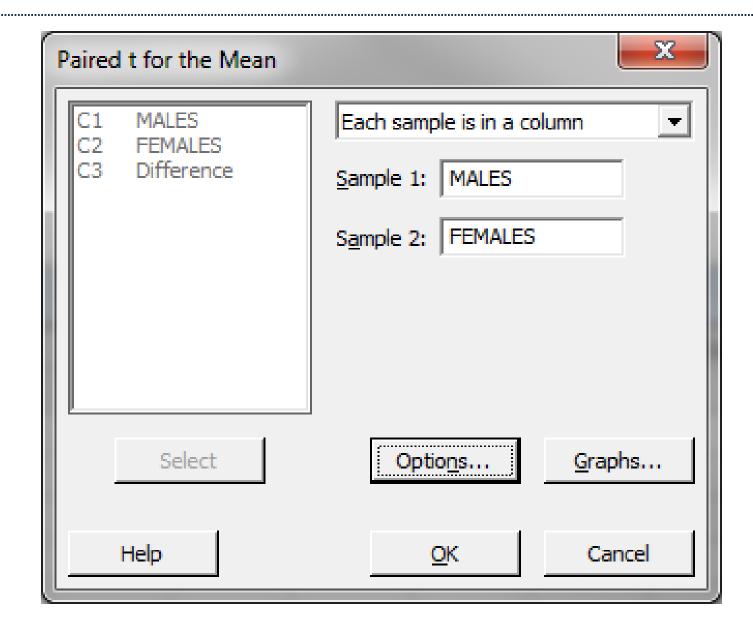
- The p-value of the normality test is 0.287, greater than the alpha level (0.05), so we fail to reject the null hypothesis and we claim that the difference is normally distributed.
- When the difference is not normally distributed, we need other hypothesis testing methods.





- Step 3: Run the paired t-test to compare the means of two dependent data sets.
 - 1) Click Stat → Basic Statistics → Paired t.
 - 2) A new window named "Paired t (Test and Confidence Interval)" pops up.
 - 3) Click in the blank box right next to "First Sample" and the three columns names appear in the list box on the left.
 - 4) Select "MALES" as "First Sample."
 - 5) Select "FEMALES" as "Second Sample."
 - 6) Click "OK."
 - 7) The paired t-test results appears in the session window.







 The p-value of the paired t-test is 0.213, greater than the alpha level (0.05), so we fail to reject the null hypothesis and we claim that there is no statistically significant difference between the salaries of male and those of female professors.

Descriptive Statistics

Sample	Ν	Mean	StDev	SE Mean
MALES	22	33.432	3.546	0.756
FEMALES	22	33.200	3.601	0.768

Estimation for Paired Difference

95% CI for

Mean StDev SE Mean μ_difference

0.232 0.846 0.180 (-0.143, 0.607)

 μ _difference: mean of (MALES - FEMALES)

Test

Null hypothesis H_0 : μ_d ifference = 0 Alternative hypothesis H_1 : μ_d ifference $\neq 0$

T-Value	P-Value
1.29	0.213



3.3.2 One Sample Variance



What is One Sample Variance Test?

• One sample variance test is a hypothesis testing method to study whether there is a statistically significant difference between a population variance and a specified value.

- Null Hypothesis (H₀): $\sigma^2 = \sigma_0^2$
- Alternative Hypothesis (H_a): $\sigma^2 \neq \sigma_0^2$

where σ^2 is the variance of a population of our interest and σ_0^2 is the specific value we want to compare against.



Chi-Square Test

• A **chi-square test** is a statistical hypothesis test in which the test statistic follows a chi-square distribution when the null hypothesis is true.

 A chi-square test can be used to test the equality between the variance of a normally distributed population and a specified value.



Chi-Square Test

Test Statistic

$$\chi_{calc}^2 = \frac{(n-1)s^2}{\sigma_0^2}$$

where

 s^2 is the observed variance and n is the sample size.

 σ_0^2 is the specified value we compare against.

Critical Value

- χ_{crit}^2 is the χ^2 value in a χ^2 distribution with the predetermined significance level α and degrees of freedom (n-1).
- χ_{crit}^2 values for a two-sided and a one-sided χ^2 -test with the same significance level α and degrees of freedom (n-1) are different.



Chi-Square Test

- Based on the sample data, we calculated the test statistic χ_{crit}^2 , which is compared against χ_{calc}^2 to make a decision of whether to reject the null.
 - Null Hypothesis (H₀): $\sigma^2 = \sigma_0^2$
 - Alternative Hypothesis (H_a): $\sigma^2 \neq \sigma_0^2$
- If $|\chi_{calc}|^2 > \chi_{crit}^2$, we reject the null and claim there is a statistically significant difference between the population variance and the specified value.
- If $|\chi_{calc}|^2 < \chi_{crit}^2$, we fail to reject the null and claim there is not any statistically significant difference between the population variance and specified value.

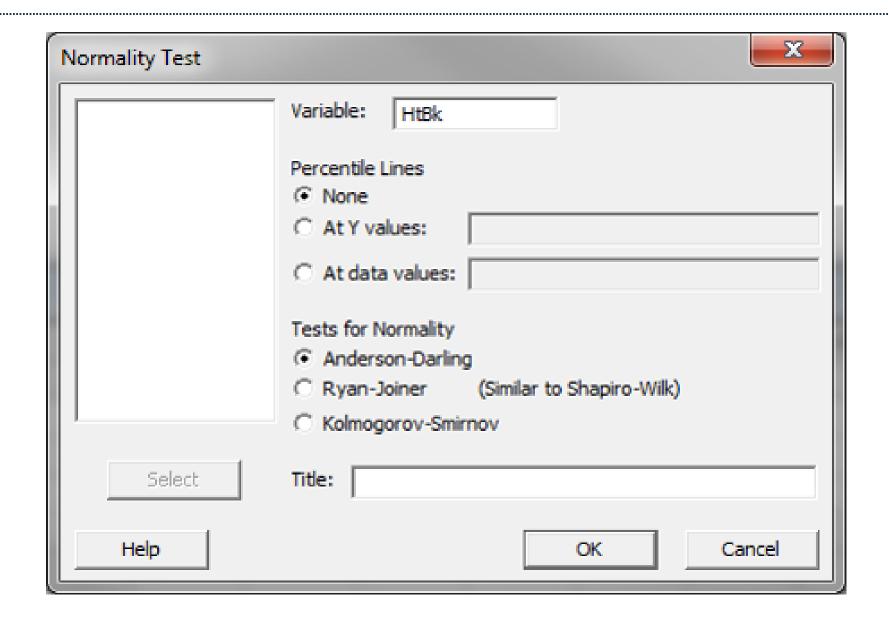


- Case study: We are interested in comparing the variance of the height of basketball players with zero.
 - Data File: "One Sample T-Test" tab in "Sample Data.xlsx"
 - Null Hypothesis (H_0): $\sigma^2 = 0$
 - Alternative Hypothesis (H_a): $\sigma^2 \neq 0$



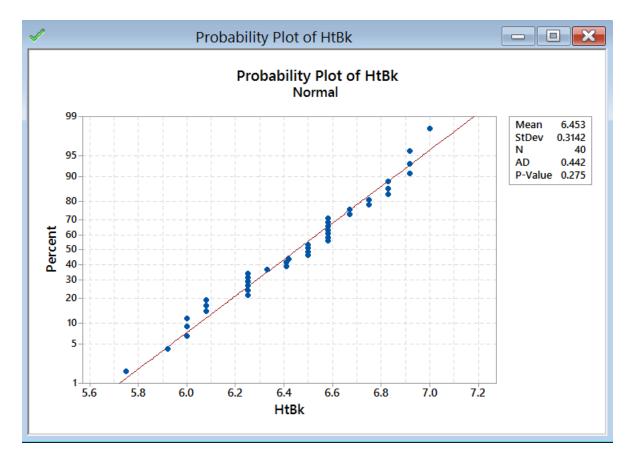
- Step 1: Test whether the data are normally distributed
 - 1. Click Stat → Basic Statistics → Normality Test.
 - 2. A new window named "Normality Test" pops up.
 - 3. Select "HtBk" as the "Variable."
 - 4. Click "OK."
 - 5. The normality test result appears in the new window.







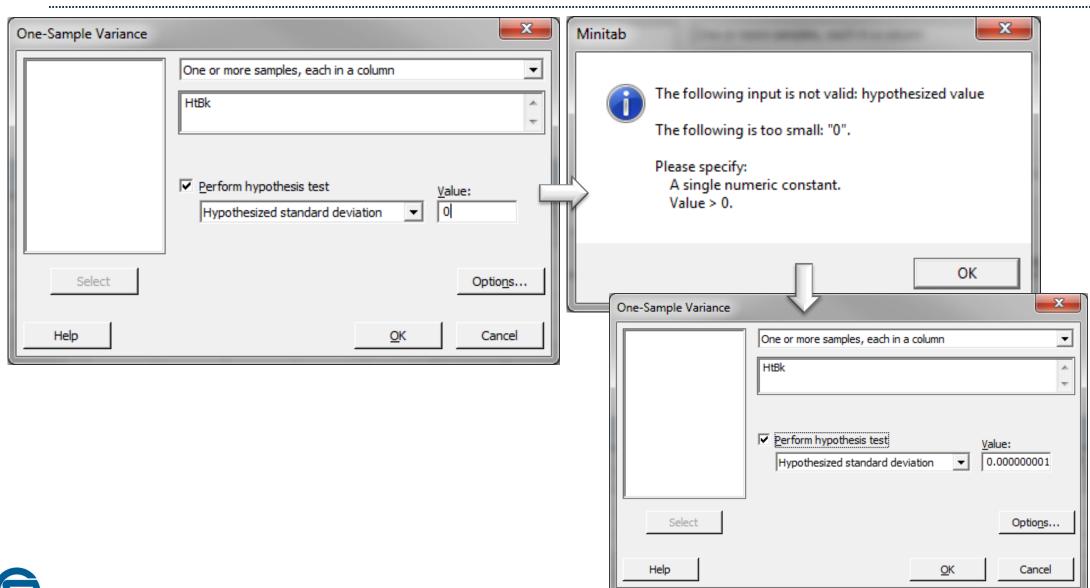
- Null Hypothesis (H₀): The data are normally distributed.
- Alternative Hypothesis (H_a): The data are not normally distributed.
- Since the p-value of the normality is 0.275, greater than alpha level (0.05), we fail to reject the null and claim that the data are normally distributed.
- If the data are not normally distributed, you need to use other hypothesis tests.





- Step 2: Check if the population variance is equal to the specified value (zero).
 - 1) Click Stat → Basic Statistics → 1 Variance.
 - 2) A new window named "1 Variance" appears.
 - 3) Click in the blank box below "Columns" and "HtBk" appears in the list box on the left.
 - 4) Check the box of "Perform hypothesis test."
 - 5) Select "Hypothesized variance" from the dropdown menu.
 - 6) Enter "0" as the "Hypothesized variance."
 - 7) Click "OK."
 - 8) An alert window pops up saying "Invalid hypothesized value. The following value is too small: "0". Please specify: A single numeric constant. Value > 0."
 - 9) Click "OK" in the alert message window.
 - 10) Enter "0.000000001" as the "Hypothesized variance" instead.
 - 11) Click "OK."
 - 12) The one sample variance test appears in new window.







Since the p-value of the one sample variance test is smaller than alpha level (0.05), we reject the null and claim that the population variance is statistically different from zero.

Another way to see whether the population variance is statistically different from zero is to check whether zero stays between the upper and lower confidence interval boundaries of the variance. If yes, we fail to reject the null hypothesis and claim that the population variance is not statistically different from zero.

Test and CI for One Variance: HtBk

Method

σ: standard deviation of HtBk

The Bonett method is valid for any continuous distribution.

The chi-square method is valid only for the normal distribution.

Descriptive Statistics

				95% CI for σ	
			95% CI for σ	using	
Ν	StDev	Variance	using Bonett	Chi-Square	
40	0.314	0.0987	(0.266, 0.390)	(0.257, 0.403)	

Test

Null hypothesis	H_0 : $\sigma^2 = 0.0000000001$
Alternative hypothesis	$H_1: \sigma^2 \neq 0.0000000001$

Method	Test Statistic	DF	P-Value
Bonett	_	_	0.000
Chi-Square	3.85068E+10	39	0.000



3.3.3 One Way ANOVA



What is One-Way ANOVA?

- One-way ANOVA (one-way analysis of variance) is a statistical method to compare means of two or more populations.
 - Null Hypothesis $(H_0)\mu_1 = \mu_2 = ... = \mu_k$
 - Alternative Hypothesis (H_a): at least one μ_i is different, where i is any value from 1 to k.
- It is a generalized form of the two sample t-test since a two sample t-test compares two population means and one-way ANOVA compares k population means where $k \ge 2$.



Assumptions of One-Way ANOVA

- The sample data drawn from *k* populations are unbiased and representative.
- The data of *k* populations are continuous.
- The data of k populations are normally distributed.
- The variances of *k* populations are equal.



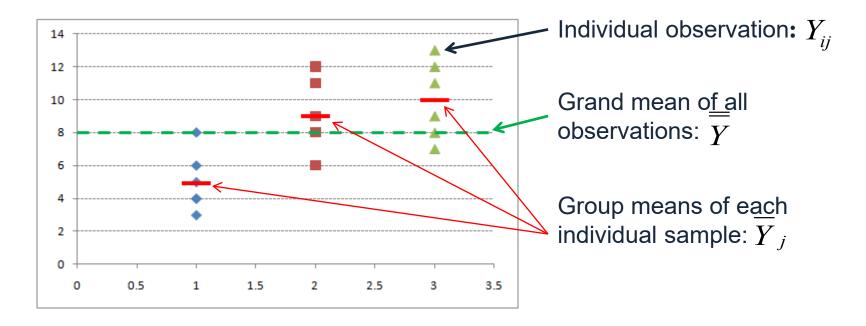
- ANOVA compares the means of different groups by analyzing the variances between and within groups.
- Let us say we are interested in comparing the means of three normally distributed populations. We randomly collected one sample for each population of our interest.
 - Null Hypothesis (H_0): $\mu_1 = \mu_2 = \mu_3$
 - Alternative Hypothesis (H_a): one of the μ is different from the others.



- Based on the sample data, the means of the three populations might look different because of two variation sources.
 - Variation between groups
 There are non-random factors leading to the variation between groups.
 - 2) Variation within groups

 There are random errors resulting in the variation within each individual group.
- What we care about the most is the variation between groups since we are interested in whether the groups are statistically different from each other.
- Variation between groups is the *signal* we want to detect and variation within groups is the *noise* which corrupts the signal.





Total Variation = SS(Total) =
$$\sum_{j=1}^{k} \sum_{i=1}^{n_j} (Y_{ij} - \overline{Y})^2$$

Between Variation = SS(Between) =
$$\sum_{j=1}^{k} n_j (\overline{Y}_j - \overline{\overline{Y}})^2$$

Within Variation = SS(Within) =
$$\sum_{j=1}^{k} \sum_{i=1}^{n_j} (Y_{ij} - \overline{Y}_j)^2$$



- Variation Components
 - Total Variation = Variation Between Groups + Variation Within Groups

$$\sum_{j=1}^{k} \sum_{i=1}^{n_j} (Y_{ij} - \overline{Y})^2 = \sum_{j=1}^{k} n_j (\overline{Y}_j - \overline{Y})^2 + \sum_{j=1}^{k} \sum_{i=1}^{n_j} (Y_{ij} - \overline{Y}_j)^2$$

- Total Variation = sums of squares of the vertical difference between the individual observation and the grand mean
- Variation Between Groups = sums of squares of the vertical difference between the group mean and the grand mean
- Variation Within Groups = sum of squares of the vertical difference between the individual observation and the group mean



- Degrees of Freedom (DF)
 - In statistics, the degrees of freedom is the number of unrestricted values in the calculation of a statistic.
- Degrees of Freedom Components
 - DF_{total}= DF_{between} + DF_{within}
 - $DF_{total} = n 1$
 - $DF_{between} = k 1$
 - $DF_{within} = n k$

where

n is the total number of observations k is the number of groups.



- Signal-to-Noise Ratio (SNR)
 - SNR denotes the ratio of a signal to the noise corrupting the signal.
 - It measures how much a signal has been corrupted by the noise.
 - When it is higher than 1, there is more signal than noise.
 - The higher the SNR, the less the signal has been corrupted by the noise.
- F-ratio is the SNR in ANOVA

$$F = \frac{MS_{between}}{MS_{within}} = \frac{SS_{between}}{SS_{within}} \frac{DF_{between}}{DF_{within}} = \frac{\sum_{j=1}^{k} (\overline{Y}_{j} - \overline{Y}_{j})^{2} / (k-1)}{\sum_{j=1}^{k} \sum_{i=1}^{n_{j}} (Y_{ij} - \overline{Y}_{j})^{2} / (n-k)}$$

- In ANOVA, we use the F-test to compare the means of different groups. The F-ratio calculated as above is the test statistic F_{calc}.
- The critical value (F_{cri}) in an F-test can be derived from the F table with predetermined significance level (α) and with (k -1) degrees of freedom in the numerator and (n - k) degrees of freedom in the denominator.



How ANOVA Works

- Null Hypothesis (H₀): $\mu_1 = \mu_2 = ... = \mu_k$
- Alternative Hypothesis (H_a): at least one μ_i is different, where i is any value from 1 to k.
- If $|F_{calc}| < F_{crit}$, we fail to reject the null and claim that the means of all the populations of our interest are the same.
- If $|F_{calc}| > F_{crit}$, we reject the null and claim that there is at least one mean different from the others.



Model Validation

 ANOVA is a modeling procedure. To make sure the conclusions made in ANOVA are reliable, we need to perform residuals analysis.

- Good residuals:
 - Have a mean of zero
 - Are normally distributed
 - Are independent of each other
 - Have equal variance.

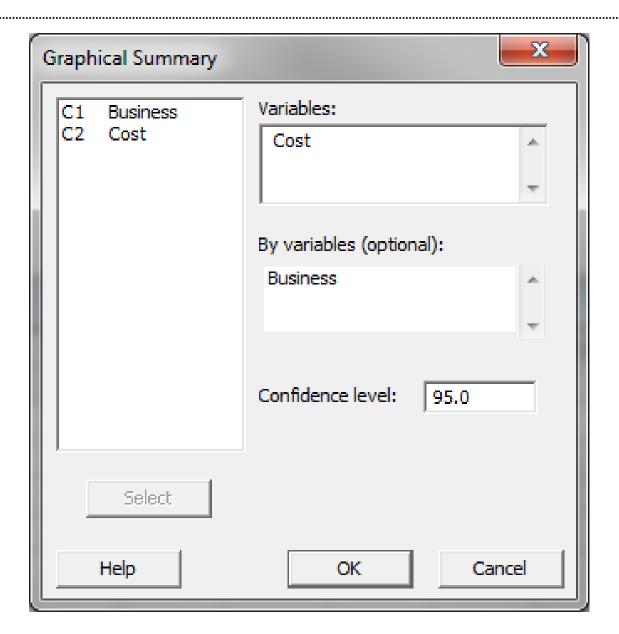


- Case study: We are interested in comparing the average startup costs
 of five kinds of business.
 - Data File: "One-Way ANOVA" tab in "Sample Data.xlsx"
- Null Hypothesis (H_0): $\mu_1 = \mu_2 = \mu_3 = \mu_4 = \mu_5$
- Alternative Hypothesis (H_a): at least one of the five means is different from others.



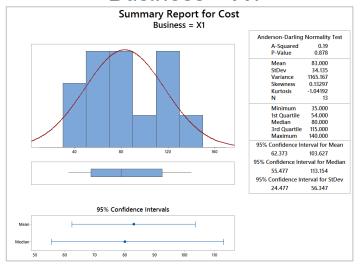
- Step 1: Test whether the data for each level are normally distributed.
 - 1) Click Stat → Basic Statistics → Graphical Summary.
 - 2) A new window named "Graphical Summary" pops up.
 - 3) Select the "Cost" as the variable.
 - 4) Click in the blank box right next to "By variables (optional)" and the "Business" appears in the list box on the left.
 - Select the "Business" as the "By variables (optional)."
 - 6) Click "OK."
 - 7) The normality results appear in the new window.



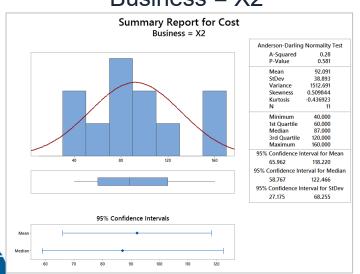




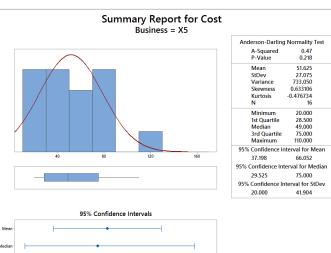
Business = X1



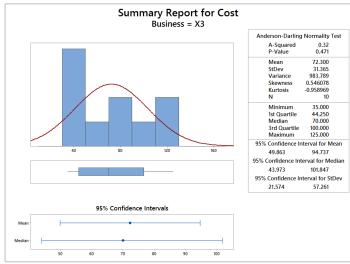
Business = X2



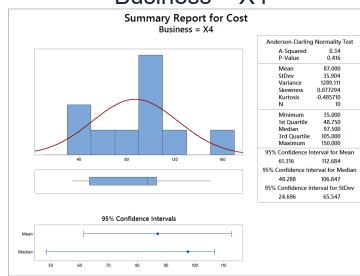
Business = X5



Business = X3



Business = X4





Null Hypothesis (H_0): The data are normally distributed. Alternative Hypothesis (H_a): The data are not normally distributed.

• Since the p-values of normality tests for the five data sets are higher than alpha level (0.05), we fail to reject the null hypothesis and claim that the startup costs for any of the five businesses are normally distributed.

 If any of the five data sets are not normally distributed, we need to use other hypothesis testing methods other than one-way ANOVA.

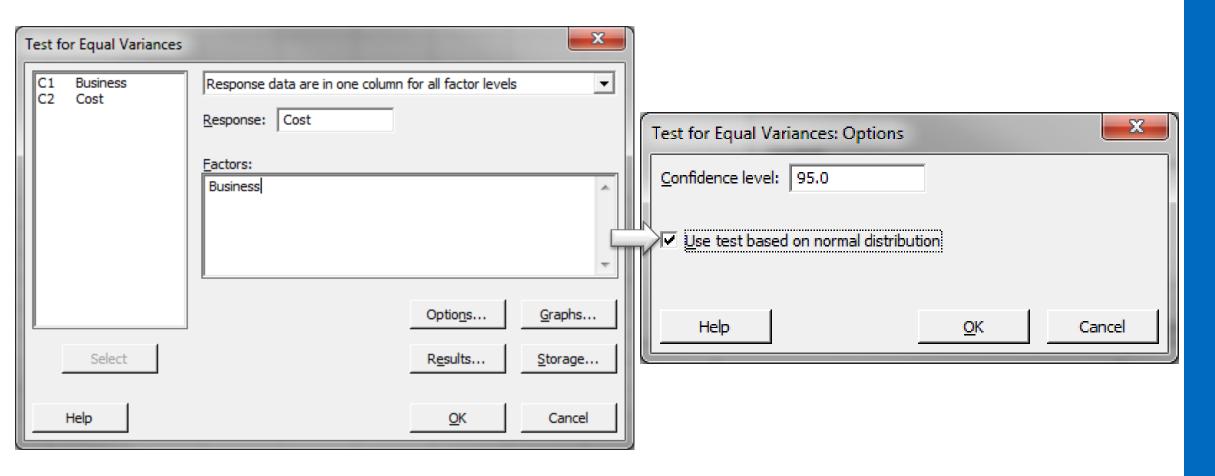


 Step 2: Test whether the variance of the data for each level is equal to the variance of other levels.

Null Hypothesis (H₀): $\sigma_1^2 = \sigma_2^2 = \sigma_3^2 = \sigma_4^2 = \sigma_5^2$ Alternative Hypothesis (H_a): at least one of the variances is different from others.

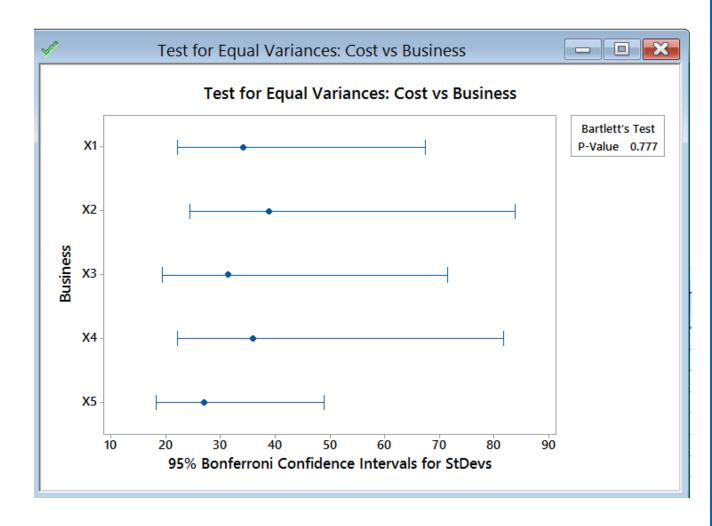
- 1) Click Stat → ANOVA → Test for equal variances.
- 2) A new window named "Test for Equal Variances" pops up.
- 3) Select the "Cost" as the "Response."
- 4) Select the "Business" as the "Factors."
- 5) Click "Options."
- 6) Select "Use test based on normal distribution"
- 7) Click "OK" to close the "Options" window
- 8) Click "OK" to run the test
- 9) The results shows up in a new window.
- 10) Use the Bartlett's test for testing the equal variances between five levels in this case since there are more than two levels in the data and the data of each level are normally distributed.







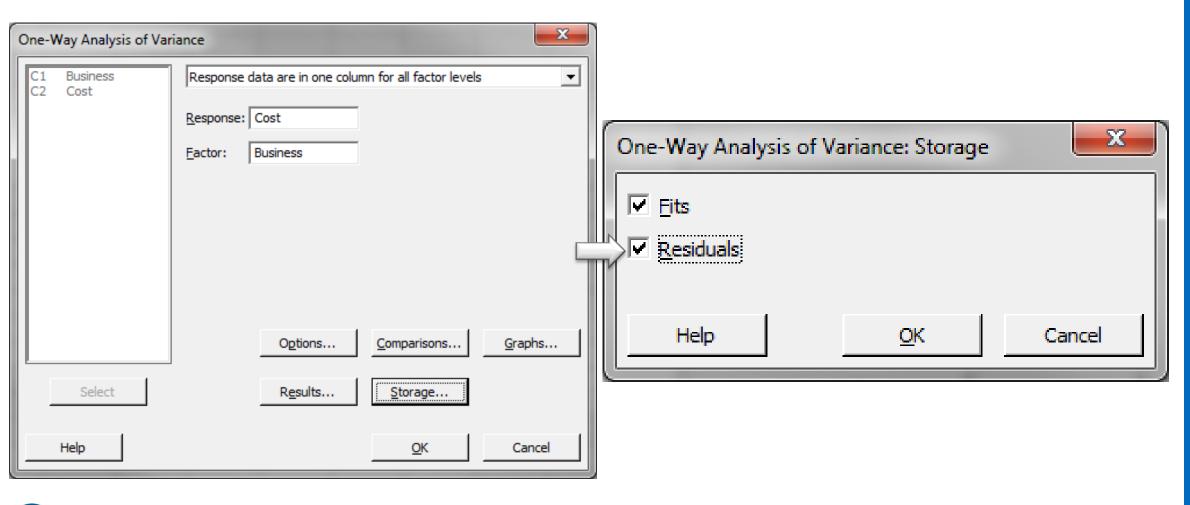
- The p-value of Barlett's test is 0.777, greater than the alpha level (0.05), so we fail to reject the null hypothesis and we claim that the variances of five groups are equal.
- If the variances are not all equal, we need to use other hypothesis testing methods other than oneway ANOVA.





- Step 3: Test whether the mean of the data for each level is equal to the means of other levels.
 - Null Hypothesis (H_0): $\mu_1 = \mu_2 = \mu_3 = \mu_4 = \mu_5$
 - Alternative Hypothesis (H_a): at lease one of the means is different from others.
 - 1) Click Stat → ANOVA → One-way.
 - 2) A new window named "One-Way Analysis of Variance" pops up.
 - 3) Select "Cost" as "Response."
 - Select "Business" as "Factor."
 - 5) Click "Storage."
 - 6) Check the boxes next to "Fits" and "Residuals"
 - 7) Click "OK" to close the "Storage" window
 - 8) Click "OK."
 - 9) The ANOVA results appear in the session window. The fitted response and the residuals are stored in the data table.







• Since the p-value of the F test is 0.018, lower than the alpha level (0.05), the null hypothesis is rejected and we conclude that the at least one of the means of the five groups is different from others.

One-way ANOVA: Cost versus Business

Method

Null hypothesis All means are equal Alternative hypothesis Not all means are equal

Significance level $\alpha = 0.05$

Equal variances were assumed for the analysis.

Factor Information

Factor	Levels	Values
Business	5	X1, X2, X3, X4, X5

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Business	4	14298	3575	3.25	0.018
Error	55	60561	1101		
Total	59	74859			

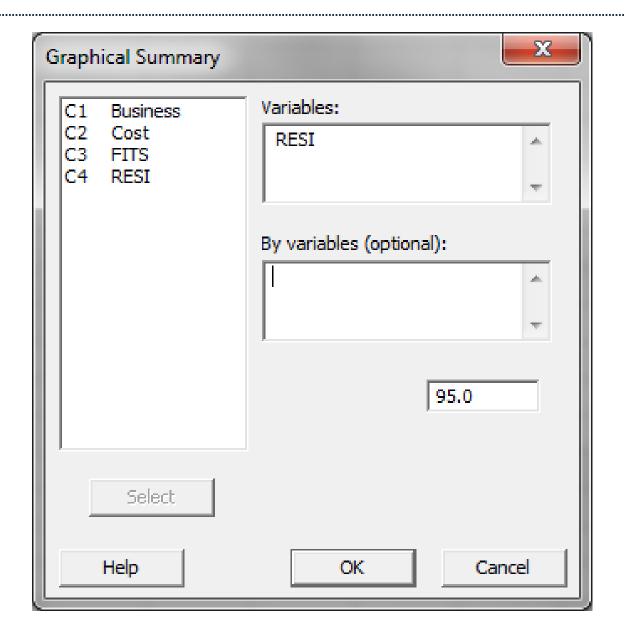
Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
33.1829	19.10%	13.22%	3.18%



- Step 4: Test whether the residuals are normally distributed with mean equal zero
 - 1) The residuals have been stored in the data table in step 4.
 - 2) Click Stat → Basic Statistics → Graphical Summary.
 - 3) A new window named "Graphical Summary" appears.
 - 4) Select "RESI" as the "Variables."
 - 5) Click "OK."
 - 6) The normality test results show up in a new window.

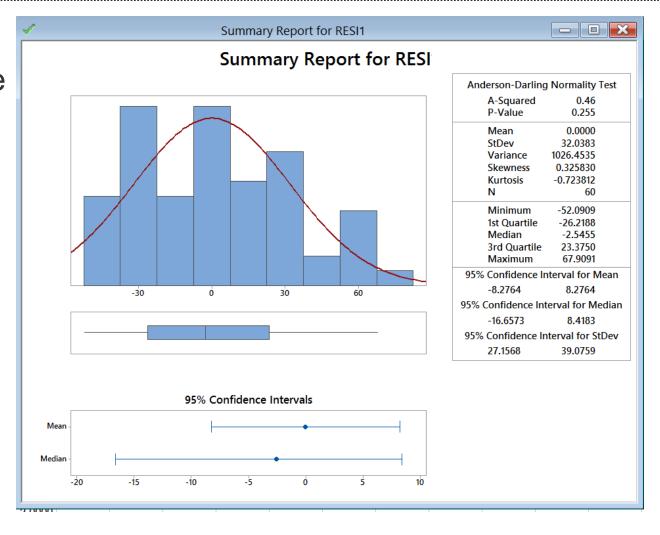






Lean Six Sigma Training - MTB

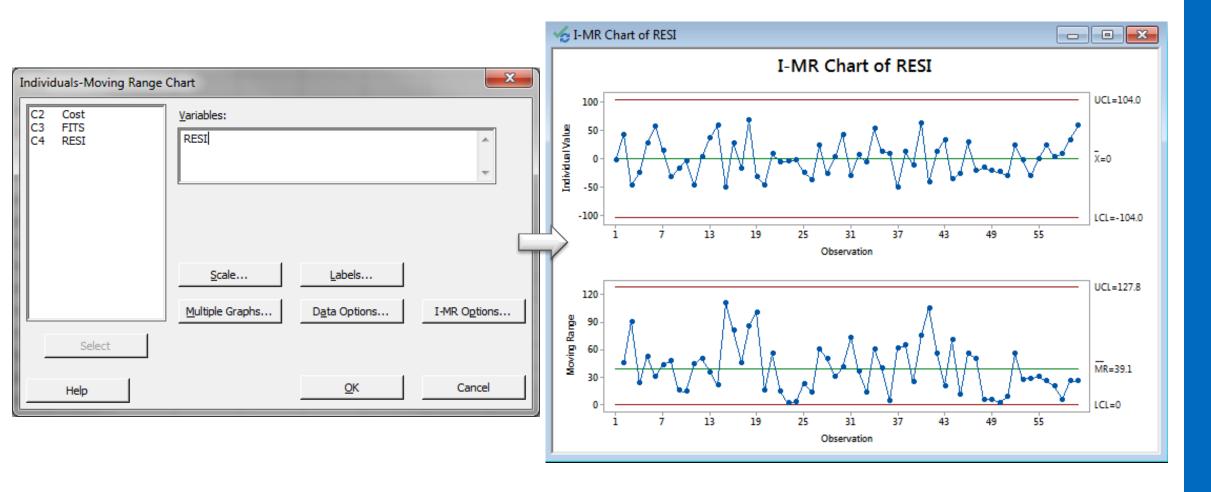
- The p-value of the normality test is 0.255, greater than the alpha level (0.05), and we conclude that the residuals are normally distributed.
- The mean of the residuals are 0.0000.





- Step 5: Check whether the residuals are independent of each other.
 - If the residuals are in time order, we can plot I-MR charts to check the independence. When no data points on the I-MR charts fail any tests, the residuals are independent of each other.
 - If the residuals are not in time order, the I-MR charts cannot deliver reliable conclusion on the independence.
 - 1) Click Stat \rightarrow Control Charts \rightarrow Variables Charts for Individuals \rightarrow I-MR.
 - 2) A new window named "Individuals Moving Range Chart" pops up.
 - 3) Select "RESI1" as the "Variables."
 - 4) Click "OK."

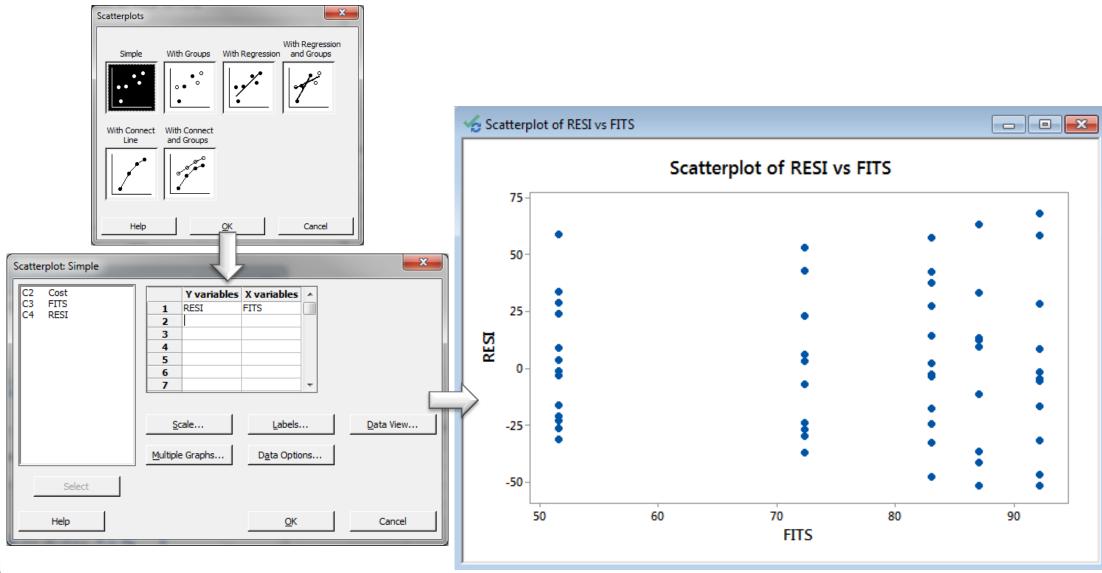






- Step 6: Plot residuals versus fitted values and check whether there is any systematic pattern.
 - 1) Click Graph → Scatter Plot.
 - 2) A new window named "Scatterplots" pops up.
 - 3) Click "OK."
 - 4) A new window named "Scatterplot Simple" pops up.
 - 5) Select the "RESI1" as the "Y variables."
 - 6) Select the "FITS1" as "X variables."
 - 7) Click "OK."
 - 8) The charts appears in a new window.
 - 9) If the data points spread out evenly at any of the five levels, we claim that the residuals have equal variances across the five levels.







3.4 Hypothesis Testing: Non-Normal Data



Black Belt Training: Analyze Phase

3.1 Inferential Statistics

- 3.1.1 Understanding Inference
- 3.1.2 Sampling Techniques and Uses
- 3.1.3 Sample Size
- 3.1.4 Central Limit Theorem

3.2 Hypothesis Testing

- 3.2.1 Goals of Hypothesis Testing
- 3.2.2 Statistical Significance
- 3.2.3 Risk; Alpha and Beta
- 3.2.4 Types of Hypothesis Tests

3.3 Hypothesis Testing: Normal Data

- 3.3.1 One and Two Sample T-Tests
- 3.3.2 One sample variance
- 3.3.3 One Way ANOVA

3.4 Hypothesis Testing: Non-Normal Data

- 3.4.1 Mann-Whitney
- 3.4.2 Kruskal-Wallis
- 3.4.3 Moods Median
- 3.4.4 Friedman
- 3.4.5 One Sample Sign
- 3.4.6 One Sample Wilcoxon
- 3.4.7 One and Two Sample Proportion
- 3.4.8 Chi-Squared (Contingency Tables)
- 3.4.9 Test of Equal Variances



3.4.1 Mann-Whitney



What is the Mann-Whitney Test?

- The Mann-Whitney test (also called Mann-Whitney U test or Wilcoxon rank-sum test) is a statistical hypothesis test to compare the medians of two populations that are not normally distributed.
 - Null Hypothesis (H_0): $\eta_1 = \eta_2$
 - Alternative Hypothesis (H_a): $\eta_1 \neq \eta_2$

where η_1 is the median of one population and η_1 is the median of the other population.



Mann-Whitney Test Assumptions

- The sample data drawn from the populations of interest are unbiased and representative.
- The data of both populations are continuous or ordinal when the spacing between adjacent values is not constant.
- The two populations are independent to each other.
- The Mann-Whitney test is robust for the non-normally distributed population.
- The Mann-Whitney test can be used when shapes of the two populations' distributions are different.



- Step 1: Group the two samples from two populations (sample 1 is from population 1 and sample 2 is from population 2) into a single data set and then sort the data in ascending order ranked from 1 to n, where n is the total number of observations.
- Step 2: Add up the ranks for all the observations from sample 1 and call it R₁. Add up the ranks for all the observations from sample 2 and call it R₂.



Step 3: Calculate the test statistics

$$U = \min(U_1, U_2)$$

where
$$U_1 = n_1 n_2 + \frac{n_1(n_1+1)}{2} - R_1$$

$$U_2 = n_1 n_2 + \frac{n_2(n_2+1)}{2} - R_2$$

 n_1 and n_2 are the sample sizes.

 R_1 and R_2 are the sum of ranks for observations from sample 1 and 2 respectively.



- Step 4: Make a decision on whether to reject the null hypothesis.
 - Null Hypothesis (H_0): $\eta_1 = \eta_2$
 - Alternative Hypothesis (H_a): $\eta_1 \neq \eta_2$
- If both of the sample sizes are smaller than 10, the distribution of U under the null hypothesis is tabulated.
 - The test statistic is U and, by using the Mann-Whitney table, we would find the p-value.
 - If the p-value is smaller than alpha level (0.05), we reject the null hypothesis.
 - If the p-value is greater than alpha level (0.05), we fail to reject the null hypothesis.



• If both sample sizes are greater than 10, the distribution of U can be approximated by a normal distribution. In other words, $\frac{U-\mu}{\sigma}$ follows a standard normal distribution.

$$Z_{calc} = \frac{U - \mu}{\sigma}$$

where

$$\mu = \frac{n_1 n_2}{2} \qquad \sigma = \sqrt{\frac{n_1 n_2 (n_1 + n_2 + 1)}{12}}$$

When $|Z_{calc}|$ is greater than Z value at $\alpha/2$ level (e.g., when $\alpha = 5\%$, the z value we compare $|Z_{calc}|$ to is 1.96), we reject the null hypothesis.



- Case study: We are interested in comparing customer satisfaction between two types of customers using a nonparametric (i.e., distribution-free) hypothesis test: Mann-Whitney test.
 - Data File: "Mann-Whitney" tab in "Sample Data.xlsx"

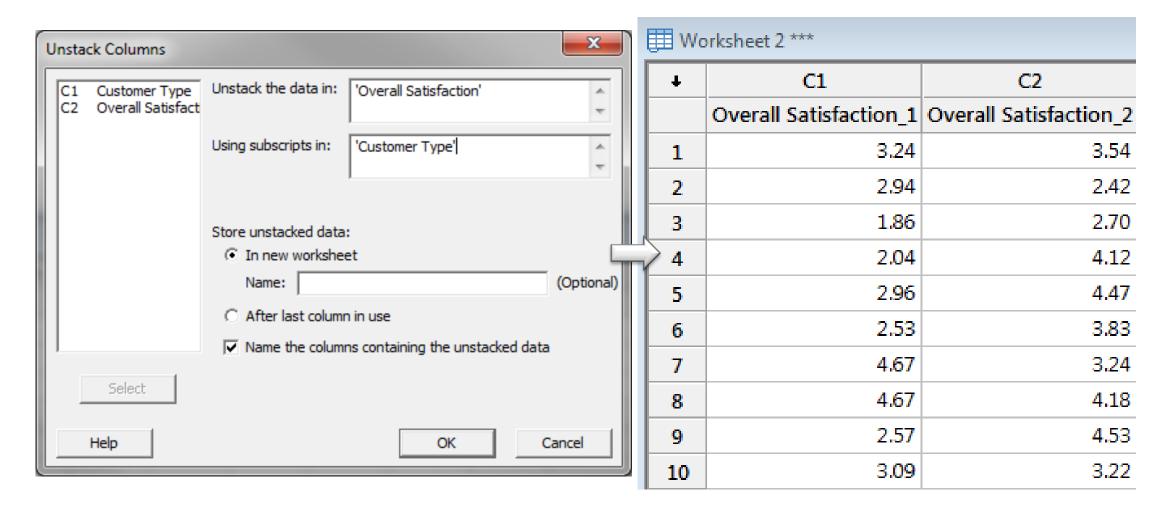


- Null Hypothesis (H_0): $\eta_1 = \eta_2$
- Alternative Hypothesis (H_a): $\eta_1 \neq \eta_2$

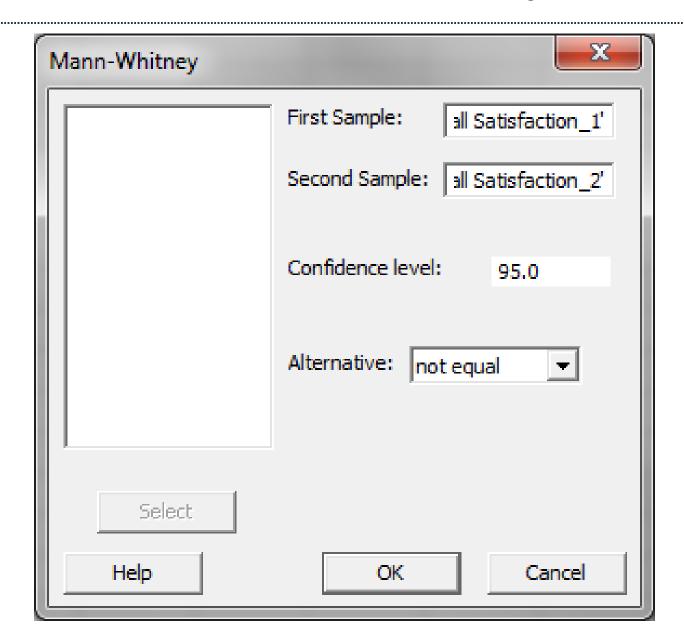


- Steps to run a Mann-Whitney Test in Minitab:
 - 1) Un-stack the data into two separate columns:
 - "Overall Satisfaction 1" for customer type = 1
 - "Overall Satisfaction 2" for customer type = 2
 - 2) Click Stat → Nonparametrics → Mann-Whitney.
 - 3) A new window named "Mann-Whitney" pops up.
 - 4) Select "Overall Satisfaction 1" as the "First Sample."
 - 5) Select "Overall Satisfaction 2" as the "Second Sample."
 - 6) Click "OK."
 - 7) The Mann-Whitney test results appear in the session window.











• The p-value of the test is lower than alpha level (0.05); so we reject the null hypothesis and conclude that there is a statistically significant difference between the overall satisfaction medians of the two customer types.

Mann-Whitney: Overall Satisfaction_1, Overall Sat

Method

 η_1 : median of Overall Satisfaction_1 η_2 : median of Overall Satisfaction_2 Difference: $\eta_1 - \eta_2$

Descriptive Statistics

Sample	Ν	Median
Overall Satisfaction_1	31	3.56
Overall Satisfaction_2	42	4.34

Estimation for Difference

	CI for	Achieved
Difference	Difference	Confidence
-0.81	(-1.17, -0.45)	95.11%

Test

Null hypothesis	H_0 : $\eta_1 - \eta_2 = 0$
Alternative hypothesis	H_1 : $\eta_1 - \eta_2 \neq 0$

Method	W-Value	P-Value
Not adjusted for ties	772.50	0.000
Adjusted for ties	772.50	0.000



3.4.2 Kruskal-Wallis



Kruskal-Wallis One-Way ANOVA

- The Kruskal-Wallis one-way analysis of variance is a statistical hypothesis test to compare the medians among more than two groups.
 - Null Hypothesis (H_0): $\eta_1 = \eta_2 = \dots = \eta_k$
 - Alternative Hypothesis (H_a): at least one of the medians is different from others.

 η_i is the median of population i, and k is the number of groups of our interest.

- It is an extension of Mann-Whitney test.
- If the distributions of k populations are not identically shaped or there are outliers in the distribution, Mood's median test is a more robust than Kruskal-Wallis.



Kruskal-Wallis One-Way ANOVA: Assumptions

- The sample data drawn from the populations of interest are unbiased and representative.
- The data of k populations are continuous or ordinal when the spacing between adjacent values is not constant.
- The k populations are independent to each other.
- The Kruskal-Wallis test is robust for the non-normally distributed population.



How Kruskal-Wallis One-Way ANOVA Works

- Step 1: Group the k samples from k populations (sample i is from population i) into one single data set and then sort the data in ascending order ranked from 1 to N, where N is the total number of observations across k groups.
- Step 2: Add up the ranks for all the observations from sample i and call it r_i, where i can be any integer between 1 and k.



How Kruskal-Wallis One-Way ANOVA Works

Step 3: Calculate the test statistic

$$T = (N-1) \frac{\sum_{i=1}^{k} n_i (\bar{r}_i - \bar{r})^2}{\sum_{i=1}^{k} \sum_{j=1}^{n_i} (r_{ij} - \bar{r})^2}$$

where

k is the number of groups.

 n_i is the sample size of sample i.

N is the total number of all the observations across k groups. r_{ij} is the rank (among all the observations) of observation j from group i.

$$-\frac{\sum_{i=1}^{n_i} r_{ij}}{n_i} \qquad -\frac{1}{2}(N+1)$$



How Kruskal-Wallis One-Way ANOVA Works

- Step 4: Make a decision of whether to reject the null hypothesis.
 - Null Hypothesis (H_0): $\eta_1 = \eta_2 = ... = \eta_k$
 - Alternative Hypothesis (H_a): at least one of the medians is different from others.
- The test statistic follows chi-square distribution when the null hypothesis is true.
 - If T is greater than χ^2_{k-1} , we reject the null and claim there is at least one median statistically different from other medians.
 - If T is smaller than χ^2_{k-1} , we fail to reject the null and claim the medians of k groups are equal.



- Case study: We are interested in comparing customer satisfaction among three types of customers using a nonparametric (i.e., distribution-free) hypothesis test: Kruskal-Wallis one-way ANOVA.
 - Data File: "Kruskal-Wallis" tab in "Sample Data.xlsx"

Customer Satisfaction Comparison

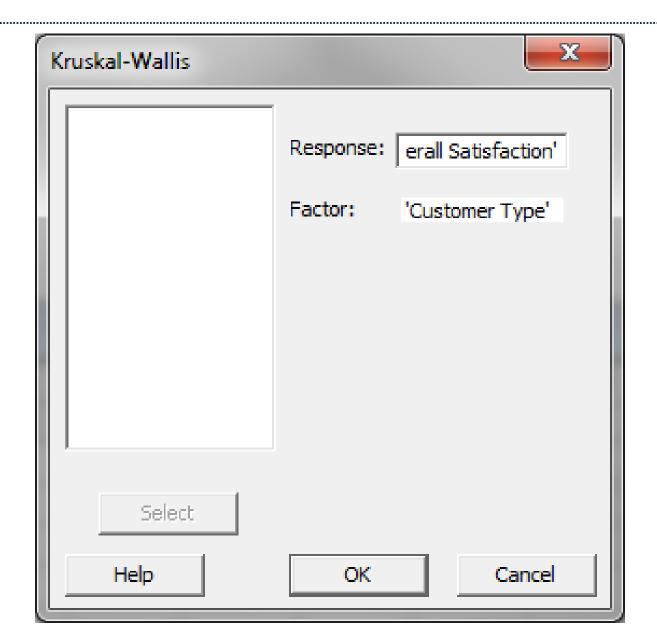


- Null Hypothesis (H_0): $\eta_1 = \eta_2 = \eta_3$
- Alternative Hypothesis (H_a): at least one of the customer types has different overall satisfaction levels from the others.



- Steps to run a Kruskal-Wallis One-Way ANOVA in Minitab
 - 1) Click Stat → Nonparametrics → Kruskal-Wallis.
 - 2) A new window named "Kruskal-Wallis" pops up.
 - 3) Select "Overall Satisfaction" as the "Response."
 - 4) Select "Customer Type" as the "Factor."
 - 5) Click "OK."
 - 6) The Kruskal-Wallis test results appear in the session window.







• The p-value of the test is lower than alpha level (0.05), and we reject the null hypothesis and conclude that at least the overall satisfaction median of one customer type is statistically different from the others.

ruskal-Wallis Test: Overall Satisfaction versus Customer Type					
Descriptive Statistics					
Customer					
Туре	Ν	Median	Mean Rank	Z-Value	2
1	31	3.56	36.0	-3.34	-
2	42	4.34	65.9	4.53	}
3	27	3.51	43.1	-1.56	; ;
Overall	100		50.5		
Test					
Null hypothesis H₀: All medians are equal					
Alternative hypothesis H ₁ : At least one median is different					
Method		DF	H-Value P	-Value	
Not adjust	ed for	ties 2	21.36	0.000	
Adjusted fo	or ties	2	21.36	0.000	



3.4.3 Mood's Median



What is Mood's Median Test?

- Mood's median test is a statistical test to compare the medians of two or more populations.
 - Null Hypothesis (H_0) : $\eta_1 = ... = \eta_k$
 - Alternative Hypothesis (H_a): at least one of the medians is different from the others.
 - k is the number of groups of our interest and is equal to or greater than two.
- Mood's median is an alternative to Kruskal-Wallis.
- It is the extension of one sample sign test.
- For the data with outliers, Mood's median test is more robust than the Kruskal-Wallis test.



Mood's Median Test Assumptions

- The sample data drawn from the populations of interest are unbiased and representative.
- The data of k populations are continuous or ordinal when the spacing between adjacent values is not constant.
- The k populations are independent to each other.
- The distributions of k populations have the same shape.
- Mood's median test is robust for non-normally distributed populations.
- Mood's median test is robust for data with outliers.



How Mood's Median Test Works

- Step 1: Group the k samples from k populations (sample i is from population i) into one single data set and get the median of this combined data set.
- Step 2: Separate the data in each sample into two groups. One consists of all the observations with values higher than the grand median and the other consists of all the observations with values lower than the grand median.



How Mood's Median Test Works

- Step 3: Run a Pearson's chi-square test to determine whether to reject the null hypothesis.
 - Null Hypothesis (H_0): $\eta_1 = ... = \eta_k$
 - Alternative Hypothesis (H_a): at least one of the medians is different from the others.
- If χ^2_{calc} is greater than χ^2_{crit} , we reject the null hypothesis and claim that at least one median is different from the others.
- If χ^2_{calc} is smaller than χ^2_{crit} , we fail to reject the null hypothesis and claim that the medians of k populations are not statistically different.



- Case study: We are interested in comparing customer satisfaction among three types of customers using a nonparametric (i.e., distribution-free) hypothesis test: Mood's median test.
 - Data File: "Median Test" tab in "Sample Data.xlsx"

Customer Satisfaction Comparison

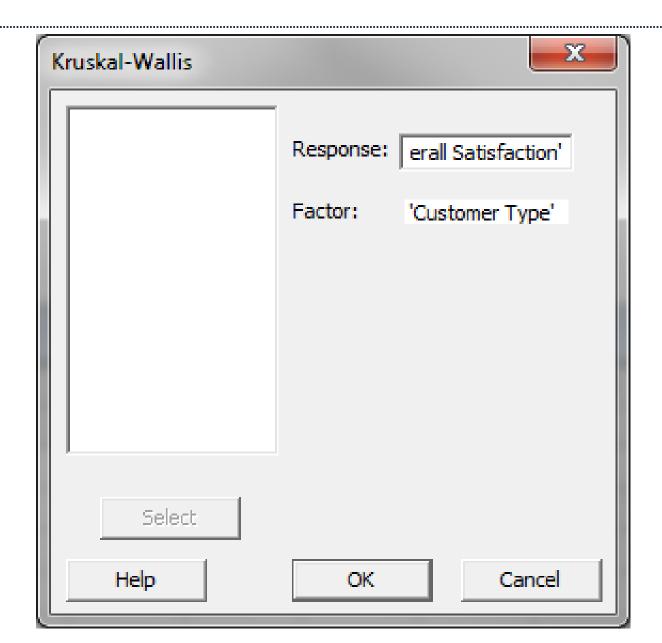


- Null Hypothesis (H_0): $\eta_1 = \eta_2 = \eta_3$
- Alternative Hypothesis (H_a): at least one customer type has different overall satisfaction from the others.



- Steps to run a Mood's median test in Minitab
 - 1) Click Stat → Nonparametrics → Mood's Median Test.
 - 2) A new window named "Mood's Median Test" pops up.
 - 3) Select "Overall Satisfaction" as the "Response."
 - 4) Select "Customer Type" as the "Factor."
 - 5) Click "OK."
 - 6) The Mood's median test results appear in the session window.







• The p-value of the test is lower than alpha level (0.05), and we reject the null hypothesis and conclude that at least the overall satisfaction median of one customer type is statistically different from the others.

Mood's Median Test: Overall Satisfaction versus Customer Type

Descriptive Statistics

Customer					
Туре	Median	N <= Overall Median	N > Overall Median	Q3 – Q1	95% Median Cl
1	3.560	21	10	1.2100	(2.95350, 3.93624)
2	4.340	12	30	0.8975	(4.09464, 4.51841)
3	3.510	17	10	0.9300	(3.28911, 4.02267)
Overall	3 945				

Test

Null hypothesis	H ₀ : The population medians are all equal
Alternative hypothesis	H₁: The population medians are not all equal

DF	Chi-Square	P-Value
2	13.43	0.001



3.4.4 Friedman



What is the Friedman Test?

- The **Friedman test** is a hypothesis test used to detect the differences in the medians of various groups across multiple attempts.
 - Null Hypothesis (H_0): The treatments have identical effects (i.e., $\eta_1 = ... = \eta_k$).
 - Alternative Hypothesis (H_a): At least one of the treatments has different effects from the others (i.e., at least one of the medians is statistically different from others).
- It is used as an alternative of the parametric repeated measures ANOVA when the assumption of normality or variance equality is not met.



Friedman Test Assumptions

- The sample data drawn from the populations of interest are unbiased and representative.
- The data are continuous or ordinal when the spacing between adjacent values is not constant.
- Results in one block are independent of the results in another.
- The Friedman test is robust for the non-normally distributed population.
- The Friedman test is robust for populations with unequal variances.



How Friedman Test Works

- Step 1: Organize the data into a tabular view with n rows indicating the blocks and k columns indicating the treatments. Each observation x_{ij} is filled into the intersection of a specific block i and a specific treatment j.
- Step 2: Calculate the ranks of each observation within each block.
- Step 3: Replace the values in the table created in step 1 with the order r_{ij} (within a block) of the corresponding observation x_{ij} .



How Friedman Test Works

Step 4: Calculate the test statistic

$$Q = \frac{n\sum_{j=1}^{k} (\bar{r}_{j} - \bar{r})^{2}}{\frac{1}{n(k-1)} \sum_{i=1}^{n} \sum_{j=1}^{k} (r_{ij} - \bar{r})^{2}}$$

where

$$r_j = \frac{1}{n} \sum_{i=1}^n r_{ij}$$

$$r = \frac{1}{nk} \sum_{i=1}^{n} \sum_{j=1}^{k} r_{ij}$$



How Friedman Test Works

- Step 5: Make a decision of whether to reject the null hypothesis.
 - When n > 15 or k > 4, the test statistic Q follows a chi-square distribution if the null hypothesis is true and the p-value is $P(\chi_{k-1}^2 \ge Q)$
 - When n < 15 and k < 4, the test statistic Q does not approximate a chisquare distribution and the p-value can be obtained from tables of Q for the Friedman test.
 - If p-value > alpha level (0.05), we fail to reject the null.



Friedman Test Examples

- A number of n water testers judge the quality of k different water samples, each of which is from a distinct water source. We will apply a Friedman Test to determine whether the water qualities of the k sources are the same.
- There are n blocks and k treatments in this experiment.
- One tester's decision would not have any influence on other testers.
- When running the experiment, each tester judges the water in a random sequence.

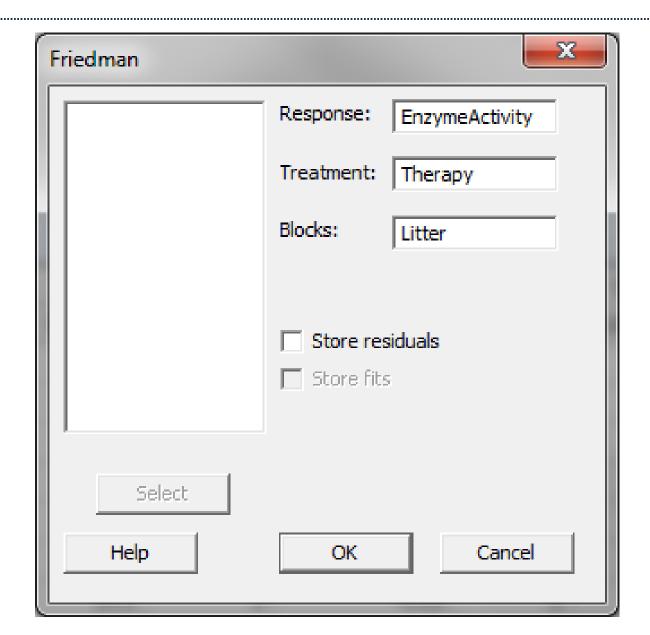


- Case study: We are interested in comparing the effect of a drug treatment on enzyme activity. Three different drug therapies were given to four animals and each animal belongs to a different litter.
 - Data File: "Friedman.MTW," (Note: this is a different data source than what you have been using to this point)
- Null Hypothesis (H_0): $\eta_1 = \eta_2 = \eta_3$
- Alternative Hypothesis (H_a): at least one treatment effect is statistically different from the others.



- Steps to run a Friedman test in Minitab
 - 1) Click Stat → Nonparametrics → Friedman.
 - 2) A new window named "Friedman" pops up.
 - 3) Select "EnzymeActivity" as the "Response."
 - 4) Select "Therapy" as the "Treatment."
 - 5) Select "Litter" as the "Blocks."
 - 6) Click "OK."
 - 7) The Friedman test results appear in the session window.







 The p-value of the Friedman test is greater than alpha level (0.05). We fail to reject the null hypothesis and conclude that there is not any statistically significant difference between the treatment effects.

Friedman Test: EnzymeActivity vs Therapy, Litter

Method

Treatment = Therapy Block = Litter

Descriptive Statistics

Therapy	Ν	Median	Sum of Ranks	
1	4	0.245000	6.5	
2	4	0.311667	7.0	
3	4	0.578333	10.5	
Overall	12	0.378333		

Test

Null hypothesis H_0 : All treatment effects are zero Alternative hypothesis H_1 : Not all treatment effects are zero

Method	DF	Chi-Square	P-Value
Not adjusted for ties	2	2.38	0.305
Adjusted for ties	2	3.80	0.150



3.4.5 One Sample Sign



What is the One Sample Sign Test?

- The **one sample sign test** is a hypothesis test to compare the median of a population with a specified value.
 - Null Hypothesis (H_0): $\eta = \eta_0$
 - Alternative Hypothesis (H_a): $\eta \neq \eta_0$
- It is an alternative test of one sample t-test when the distribution of the data is non-normal. It is robust for the data with non-symmetric distribution.



One Sample Sign Test Assumptions

- The sample data drawn from the population of interest are unbiased and representative.
- The data are continuous or ordinal when the spacing between adjacent values is not constant.
- The one sample sign test is robust for the non-normally distributed population.
- The one sample sign test does not have any assumptions on the distribution. It is a distribution-free test.



How the One Sample Sign Test Works

- Step 1: Separate the sample set of data into two groups: one with values greater than and the other with values less than the hypothesized median η_0 . Count the number of observations in each group.
- Step 2: Calculate the test statistic.
 - If the null hypothesis is true, the number of observations in each group should not be significantly different from half of the total sample size.
 - The test statistic follows a binomial distribution when the null is true. When
 n is large, we use the normal distribution to approximate the binomial
 distribution.



How the One Sample Sign Test Works

Test Statistic:

$$Z_{calc} = \frac{n_{+} - np}{\sqrt{np(1-p)}}$$

where

n₊ is the number of observations with values greater than the hypothesized median.

n is the total number of observations.

p is 0.5.

• Step 3: Make a decision on whether to reject the null hypothesis. If the $|Z_{calc}|$ is smaller than Z_{crit} , we fail to reject the null hypothesis and claim that there is no significant difference between the population median and the hypothesized median.



Use Minitab to Run a One Sample Sign Test

- Case study: We are interested in comparing the overall satisfaction of customer type 1 against a specified benchmark satisfaction (3.5) using a nonparametric (i.e., distribution-free) hypothesis test: one sample sign test.
 - Data File: "One Sample Wilcoxon" tab in "Sample Data.xlsx"

Customer Satisfaction Comparison

Customer Type 1

Vs.

3.5

- Null Hypothesis (H_0): $\eta_1 = 3.5$
- Alternative Hypothesis (H_a): $\eta_1 \neq 3.5$

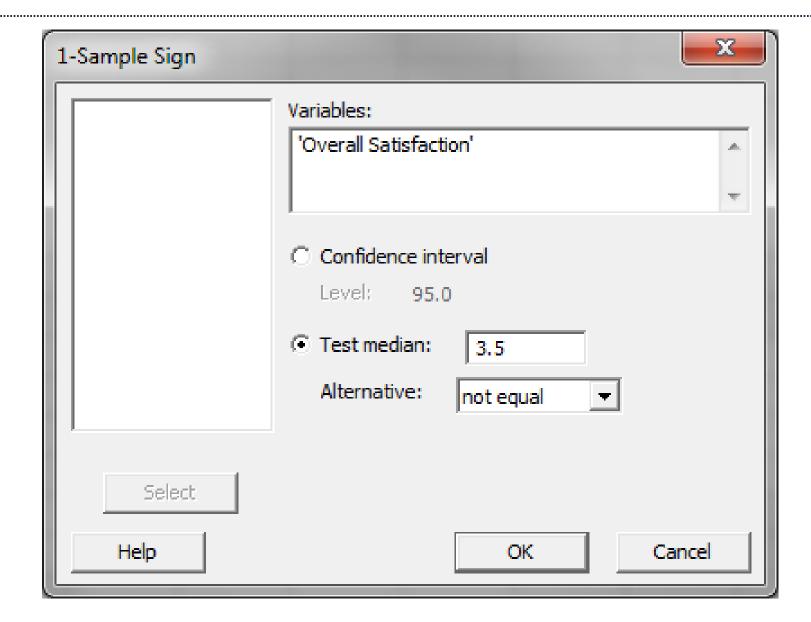


Use Minitab to Run a One Sample Sign Test

- Steps to run a one sample sign test in Minitab
 - 1) Click Stat → Nonparametrics → 1-Sample Sign.
 - 2) A new window named "1-Sample Sign" pops up.
 - 3) Select "Overall Satisfaction" as the "Variables."
 - 4) Click the box of "Test median."
 - 5) Enter "3.5" in the box next to "Test median."
 - 6) Click "OK."
 - 7) The one sample sign test results appear in session window.



Use Minitab to Run a One Sample Sign Test





 The p-value of the one sample sign test is 1.0000, higher than the alpha level (0.05), and we fail to reject the null hypothesis. There is not any statistically significant difference between the overall satisfaction of customer type 1 and the benchmark satisfaction level.

Sign Test for Median: Overall Satisfaction

Method

η: median of Overall Satisfaction

Descriptive Statistics

Sample	Ν	Median
Overall Satisfaction	31	3.56

Test

Null hypothesis	H_0 : $\eta = 3.5$
Alternative hypothesis	H₁: η ≠ 3.5

Sample	Number < 3.5	Number = 3.5	Number > 3.5	P-Value
Overall Satisfaction	15	0	16	1.000



3.4.6 One Sample Wilcoxon



What is the One Sample Wilcoxon Test?

- The **one sample Wilcoxon test** is a hypothesis test to compare the median of one population with a specified value.
 - Null Hypothesis (H_0): $\eta = \eta_0$
 - Alternative Hypothesis (H_a): $\eta \neq \eta_0$
- It is an alternative test of one sample t-test when the distribution of the data is non-normal.
- It is more powerful than one sample sign test but it assumes the distribution of the data is symmetric.



One Sample Wilcoxon Test Assumptions

- The sample data drawn from the population of interest are unbiased and representative.
- The data are continuous or ordinal when the spacing between adjacent values is not constant.
- The distribution of the data is symmetric about a median.
- The one sample Wilcoxon test is robust for the non-normally distributed population.



How the One Sample Wilcoxon Test Works

- Step 1: Create the following columns one by one:
 - Column 1: all the raw observations (X)
 - Column 2: the differences between each observation value and the hypothesized median $(X \eta_0)$
 - Column 3: the signs (+ or –) of column 2
 - Column 4: the absolute value of column 2
 - Column 5: the ranks of each item in column 4 in ascending order
 - Column 6: the product of column 3 and column 5



How the One Sample Wilcoxon Test Works

- Step 2: Calculate the test statistic W_{calc}, which is the sum of all the non-negative values in column 6.
- Step 3: Make a decision on whether to reject the null hypothesis. Use the table of critical values for the Wilcoxon test to get the W_{crit} with predetermined alpha level and number of observations.
 - If the W_{calc} is smaller than the W_{crit} , we fail to reject the null hypothesis and claim that there is no significant difference between the population median and the hypothesized median.



- Case study: We are interested in comparing the overall satisfaction of customer type 1 against a specified benchmark satisfaction (3.5) using a nonparametric (i.e., distribution-free) hypothesis test: one sample Wilcoxon test.
 - Data File: "One Sample Wilcoxon" tab in "Sample Data.xlsx"

<u>Customer Satisfaction Comparison</u>

Customer Type 1

Vs.

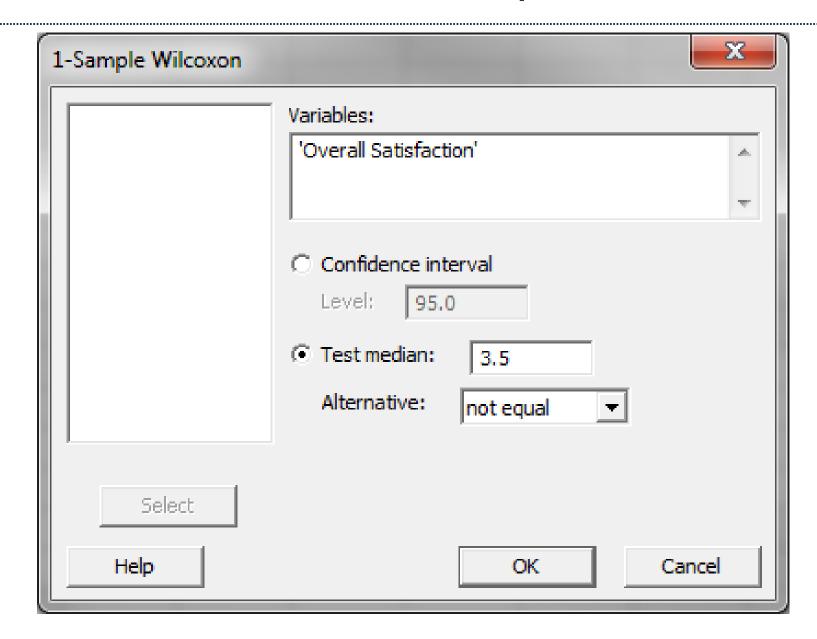
3.5

- Null Hypothesis (H_0): $\eta_1 = 3.5$
- Alternative Hypothesis (H_a): $\eta_1 \neq 3.5$



- Steps to run a one sample Wilcoxon test in Minitab
 - 1) Click Stat → Nonparametrics → 1-Sample Wilcoxon.
 - 2) A new window named "1-Sample Wilcoxon" pops up.
 - 3) Select "Overall Satisfaction" as the "Variables."
 - 4) Click the radio button "Test median."
 - 5) Enter "3.5" in the box next to "Test median."
 - 6) Click "OK."
 - 7) The one sample Wilcoxon test results appear in the session window.







 The p-value of the one sample Wilcoxon test is 0.557, higher than the alpha level (0.05), and we fail to reject the null hypothesis. There is not any statistically significant difference between the overall satisfaction of customer type 1 and the benchmark satisfaction level.

Wilcoxon Signed Rank Test: Overall Satisfaction

Method

η: median of Overall Satisfaction

Descriptive Statistics

Sample	Ν	Median
Overall Satisfaction	31	3.4075

Test

Null hypothesis	H_0 : $\eta = 3.5$
Alternative hypothesis	H₁: η ≠ 3.5

	IN for	Wilcoxon	
Sample	Test	Statistic	P-Value
Overall Satisfaction	31	217.50	0.557



3.4.7 One & Two Sample Proportion



What is the One Sample Proportion Test?

• One sample proportion test is a hypothesis test to compare the proportion of one certain outcome occurring in a population following the binomial distribution with a specified proportion.

- Null Hypothesis (H_0) : $p = p_0$
- Alternative Hypothesis (H_a): p ≠ p₀



One Sample Proportion Test Assumptions

- The sample data drawn from the population of interest are unbiased and representative.
- There are only two possible outcomes in each trial: success/failure, yes/no, and defective/non-defective etc.
- The underlying distribution of the population is binomial distribution.
- When $np \ge 5$ and $np(1 p) \ge 5$, the binomial distribution can be approximated by the normal distribution.



How the One Sample Proportion Test Works

• When $np \ge 5$ and $np(1-p) \ge 5$, we use normal distribution to approximate the underlying binomial distribution of the population.

Test Statistic:

$$Z_{calc} = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}}$$

where

 \hat{p} is the observed probability of one certain outcome occurring.

 p_0 is the hypothesized probability.

n is number of trials.

When $|Z_{calc}|$ is smaller than Z_{crit} , we fail to reject the null hypothesis and claim that there is no statistically significant difference between the population proportion and the hypothesized proportion.



- Case study: We are interested in comparing the exam pass rate of a high school this month against a specified rate (70%) using a nonparametric (i.e., distribution-free) hypothesis test: one sample proportion test.
 - Data File: "One Sample Proportion" tab in "Sample Data.xlsx"

Exam Pass Rate

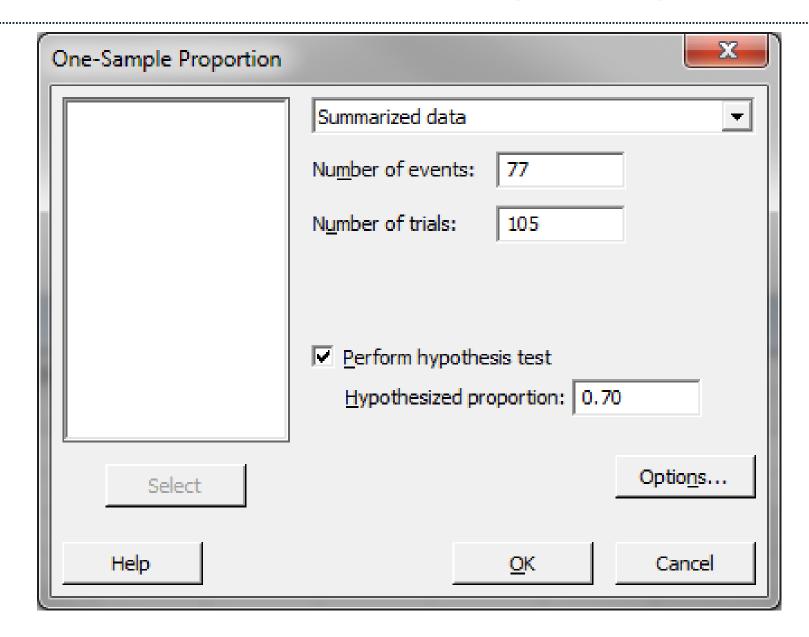


- Null Hypothesis (H_0): p = 70%
- Alternative Hypothesis (H_a): p ≠ 70%



- Steps to run a one sample proportion test in Minitab.
 - 1) Click Stat → Basic Statistics → 1 Proportion.
 - 2) A new window named "1 Proportion (Test and Confidence Interval)" appears.
 - 3) Click on the drop down menu and select "Summarized data."
 - 4) Enter "77" in the box of "Number of events."
 - 5) Enter "105" in the box of "Number of trials."
 - 6) Check the box "Perform hypothesis test."
 - 7) Enter "0.70" as the "Hypothesized proportion."
 - 8) Click "OK."
 - 9) The one sample proportion test results appear in the session window.







- The p-value of the one sample proportion test is 0.460, greater than the alpha level (0.05), and we fail to reject the null hypothesis.
- We conclude that the exam pass rate of the high school this month is not statistically different from 70%.

Test and CI for One Proportion

Method

p: event proportion Exact method is used for this analysis.

Descriptive Statistics

N	Event	Sample p	95% CI for p
105	77	0.733333	(0.638143, 0.814925)

Test

Null hypothesis H_0 : p = 0.7Alternative hypothesis H_1 : $p \neq 0.7$

P-Value 0.460



What is the Two Sample Proportion Test?

• The **two sample proportion test** is a hypothesis test to compare the proportions of one certain event occurring in two populations following the binomial distribution.

- Null Hypothesis (H_0): $p_1 = p_2$
- Alternative Hypothesis (H_a): p₁ ≠ p₂



Two Sample Proportion Test Assumptions

- The sample data drawn from the populations of interest are unbiased and representative.
- There are only two possible outcomes in each trial for both populations: success/failure, yes/no, and defective/non-defective etc.
- The underlying distributions of both populations are binomial distribution.
- When $np \ge 5$ and $np(1-p) \ge 5$, the binomial distribution can be approximated by the normal distribution.



How the Two Sample Proportion Test Works

• When $np \ge 5$ and $np(1-p) \ge 5$, we use normal distribution to approximate the underlying binomial distributions of the populations.

Test Statistic:

$$Z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}_0(1 - \hat{p}_0)\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$

where
$$\hat{p}_0 = \frac{x_1 + x_2}{n_1 + n_2}$$

 \hat{p}_1 and \hat{p}_2 are the observed proportions of events in the two samples.

 n_1 and n_2 are the number of trials in the two samples respectively.

 x_1 and x_2 are the number of events in the two samples respectively.

When $|Z_{calc}|$ is smaller than Z_{crit} , we fail to reject the null hypothesis.



- Case study: We are interested in comparing the exam pass rates of a high school in March and April using a nonparametric (i.e., distribution-free) hypothesis test: two sample proportion test.
 - Data File: "Two Sample Proportion" tab in "Sample Data.xlsx"

Exam Pass Rate Comparison

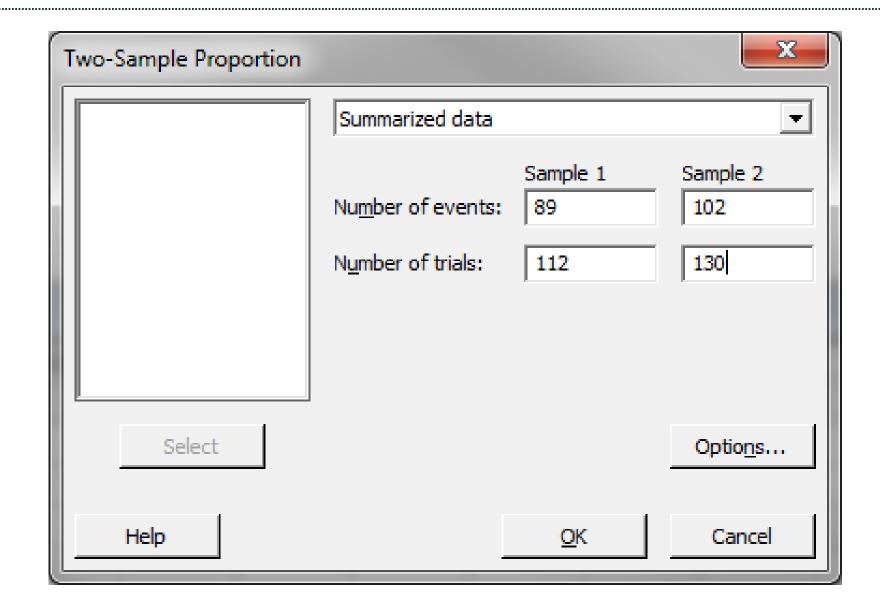


- Null Hypothesis (H_0): $p_{March} = p_{April}$
- Alternative Hypothesis (H_a): p_{March} ≠ p_{April}



- Steps to run a two sample proportion test in Minitab
 - 1) Click Stat → Basic Statistics → 2 Proportions.
 - 2) A new window named "2 Proportions (Test and Confidence Interval)" pops up.
 - 3) Choose "Summarized data" from the dropdown menu
 - 4) Enter "89" in the box intersecting "First" and "Events."
 - 5) Enter "112" in the box intersecting "First" and "Trials."
 - 6) Enter "102" in the box intersecting "Second" and "Events."
 - 7) Enter "130" in the box intersecting "Second" and "Trials."
 - 8) Click "OK."
 - 9) The two sample proportion test results appear in the session window.







- The p-value of the two sample proportion test is 0.849, greater than the alpha level (0.05), and we fail to reject the null hypothesis.
- We conclude that the exam pass rates of the high school in March and April are not statistically different.

Test and CI for Two Proportions

Method

 p_1 : proportion where Sample 1 = Event p_2 : proportion where Sample 2 = Event Difference: $p_1 - p_2$

Descriptive Statistics

Sample	Ν	Event	Sample p
Sample 1	112	89	0.794643
Sample 2	130	102	0.784615

Estimation for Difference

95% CI for
Difference Difference

0.0100275 (-0.092884, 0.112939)

CI based on normal approximation

Test

Null hypothesis

rvan riypotriesis	110. PT PZ	•
Alternative hypothesis	H ₁ : p ₁ - p ₂	<i>≠</i> 0
Method	Z-Value	P-Value
Normal approximation	0.19	0.849
Fisher's exact		0.876

 H_0 : $p_1 - p_2 = 0$



3.4.8 Chi-Squared (Contingency Tables)



What is the Chi-Square Test?

- A chi-square test is a hypothesis test in which the sampling distribution of the test statistic follows a chi-square distribution when the null hypothesis is true.
- There are multiple chi-square tests available and in this module we will cover the Pearson's chi-square test used in contingency analysis.
 - Null Hypothesis (H_0) : $p_1 = p_2 = ... = p_k$
 - Alternative Hypothesis (H_a): at least on of the proportions is different from others.
 - k is the number of populations of our interest. $k \ge 2$.



What is the Chi-Square Test?

- The chi-square test can also be used to test whether two factors are independent of each other. In other words, it can be used to test whether there is any statistically significant relationship between two discrete factors.
 - Null Hypothesis (H₀): Factor 1 is independent of factor 2.
 - Alternative Hypothesis (H_a): Factor 1 is not independent of factor 2.



Chi-Square Test Assumptions

- The sample data drawn from the populations of interest are unbiased and representative.
- There are only two possible outcomes in each trial for an individual population: success/failure, yes/no, and defective/non-defective etc.
- The underlying distribution of each population is binomial distribution.
- When $np \ge 5$ and $np(1 p) \ge 5$, the binomial distribution can be approximated by the normal distribution.



How Chi-Square Test Works

Test Statistic

$$\chi_{calc}^{2} = \sum_{i=1}^{n} \frac{(O_{i} - E_{i})^{2}}{E_{i}}$$

where

 O_i is an observed frequency.

 E_i is an expected frequency.

N is the number of cells in the contingency table.

If χ^2_{calc} is smaller than χ^2_{crit} , we fail to reject the null hypothesis.

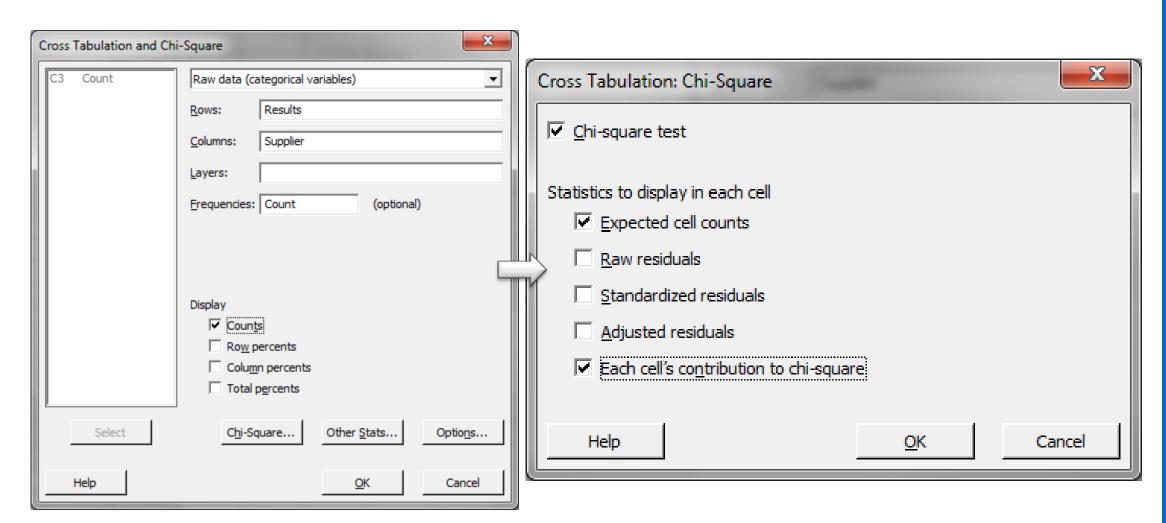


- Case study 1: We are interested in comparing the product quality exam pass rates of three suppliers A, B, and C using a nonparametric (i.e., distribution-free) hypothesis test: chi-square test.
 - Data File: "Chi-Square Test1" tab in "Sample Data.xlsx"
- Null Hypothesis (H_0) : $p_A = p_B = p_C$
- Alternative Hypothesis (H_a): at least one of the suppliers has different pass rates from the others.



- Steps to run a chi-square test in Minitab
 - 1) Click Stat → Tables → Cross Tabulation and Chi-Square.
 - 2) A new window named "Cross Tabulation and Chi-Square" pops up.
 - 3) Select "Results" as "For rows."
 - 4) Select "Supplier" as "For columns."
 - 5) Select "Count" as "Frequencies are in."
 - 6) Click the "Chi-Square" button.
 - 7) A new window named "Cross Tabulation Chi-Square" pops up.
 - 8) Check the boxes of "Chi-square test", "Expected cell counts," and "Each cell's contribution to the Chi-Square statistic."
 - 9) Click "OK" in the window named "Cross Tabulation Chi-Square."
 - 10) Click "OK" in the window named "Cross Tabulation and Chi-Square."
 - 11) The Chi-square test results appear in the session window.







- Counts are based on the sample observation.
- Expected counts are based on the assumption that the null hypothesis is true.
- Since the p-value is smaller than alpha level (0.05), we reject the null hypothesis and claim that at least one supplier has different pass rate from others.

Tabulated Statistics: Results, Supplier

Using frequencies in Count

Rows: Results Columns: Supplier

	Supplier A	Supplier B	Supplier C	All		
Fail	20	30	10	60		
	21.18	20.00	18.82			
	0.0654	5.0000	4.1360			
Pass	160	140	150	450		
	158.82	150.00	141.18			
	0.0087	0.6667	0.5515			
All	180	170	160	510		
Cell Contents						
Count						
Expected count						
Contribution to Chi-square						

Chi-Square Test

	Chi-Square	DF	P-Value	
Pearson	10.428	2	0.005	
Likelihood Ratio	10.622	2	0.005	



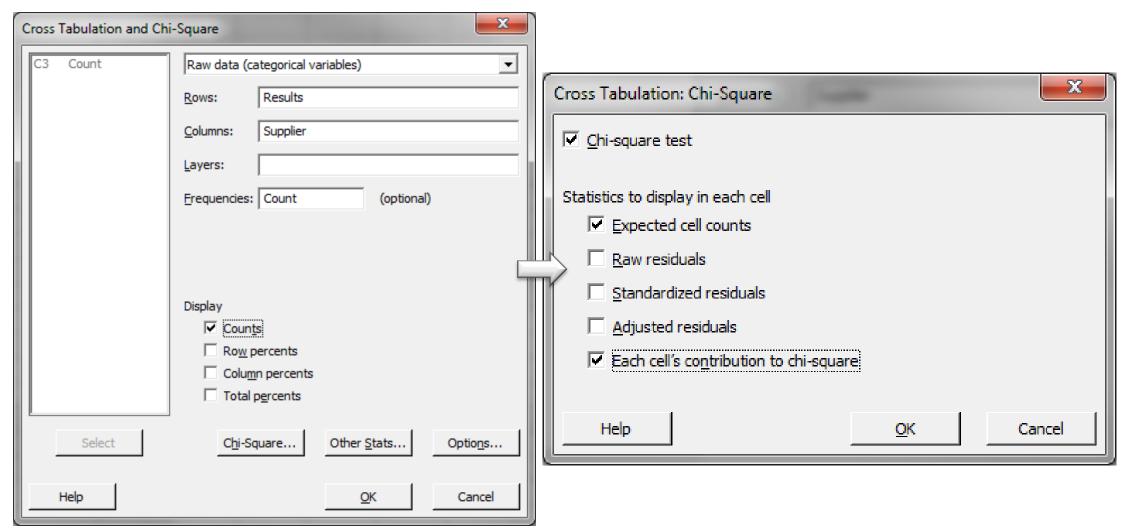
- Case study 2: We are trying to check whether there is a relationship between the suppliers and the results of the product quality exam using nonparametric (i.e., distribution-free) hypothesis test: chi-square test.
 - Data File: "Chi-Square Test2" tab in "Sample Data.xlsx"
- Null Hypothesis (H₀): product quality exam results are independent of the suppliers.
- Alternative Hypothesis (H_a): product quality exam results depend on the suppliers.



- Steps to run a chi-square test in Minitab
 - 1) Click Stat → Tables → Cross Tabulation and Chi-Square.
 - 2) A new window named "Cross Tabulation and Chi-Square" pops up.
 - 3) Select "Results" as "For rows."
 - 4) Select "Supplier" as "For columns."
 - 5) Select "Count" as "Frequencies are in."
 - 6) Click the "Chi-Square" button.
 - 7) Check the boxes of "Chi-square analysis", "Expected cell counts," and "Each cell's contribution to the Chi-Square statistic."
 - Click "OK" in the window named "Cross Tabulation Chi-Square."
 - 9) Click "OK" in the window named "Cross Tabulation and Chi-Square."
 - 10) The Chi-square test results appear in the session window.



Use Minitab to Run a Chi-Square Test





Use Minitab to Run a Chi-Square Test

• The p-value is smaller than the alpha level (0.05) and we reject the null hypothesis.

 The product quality exam results are not independent of the suppliers.

Tabulated Statistics: Results, Supplier

Using frequencies in Count

Rows: Results Columns: Supplier

	Supplier A	Supplier B	Supplier C	All
Fail	20	30	10	60
	20	20	20	
	0.000	5.000	5.000	
Marginal	20	30	40	90
	30	30	30	
	3.333	0.000	3.333	
Pass	160	140	150	450
	150	150	150	
	0.667	0.667	0.000	
All	200	200	200	600

Chi-Square Test

	Chi-Square	DF	P-Value	
Pearson	18.000	4	0.001	
Likelihood Ratio	18.595	4	0.001	



3.4.9 Tests of Equal Variance



What are Tests of Equal Variance?

• **Tests of equal variance** are a family of hypothesis tests used to check whether there is a statistically significant difference between the variances of two or more populations.

- Null Hypothesis (H₀): $\sigma_1^2 = \sigma_2^2 = ... = \sigma_k^2$
- Alternative Hypothesis (H_a): at least the variance of one population is different from others.
- k is the number of populations of interest. $k \ge 2$.
- A tests of equal variance can be used alone but most of the time it is used with other statistical methods to verify or support the assumption about the variance equality.



F-Test

- The F-test is used to compare the variances between two normally distributed populations.
- It is extremely sensitive to non-normality and serves as a preliminary step for two sample t-test.
- Test Statistic:

$$F_{calc}=rac{{S_1}^2}{{S_2}^2}$$
 , where ${f s_1}$ and ${f s_2}$ are the sample standard deviations.

The sampling distribution of the test statistic follows F distribution when the null is true.



Bartlett's Test

- Bartlett's test is used to compare the variances among two or more normally distributed populations.
- It is sensitive to non-normality and it serves as a preliminary step for ANOVA.
- Test Statistic:

$$\chi^{2} = \frac{(N-k)\ln(S_{p}^{2}) - \sum_{i=1}^{k} (n_{i}-1)\ln(S_{i}^{2})}{1 + \frac{1}{3(k-1)} \left(\sum_{i=1}^{k} \left(\frac{1}{n_{i}-1}\right) - \frac{1}{N-k}\right)} \quad \text{, where } N = \sum_{i=1}^{k} n_{i} \text{ and } S_{p}^{2} = \frac{1}{N-k} \sum_{i} (n_{i}-1)S_{i}^{2}$$

The sampling distribution of test statistic follows chi² distribution when the null is true.

Brown-Forsythe Test

- The Brown-Forsythe test is used to compare the variances between two or more populations with any distributions.
- It is not so sensitive to non-normality as Bartlett's test.
- The Test statistic is the model F statistic from the ANOVA on the transformed response $z_{ij} = \left| y_{ij} \widetilde{y}_i \right|$ where \widetilde{y}_i is the median response at ith level.



Levene's Test

- Levene's test is used to compare the variances between two or more populations with any distributions.
- It is not so sensitive to non-normality as Bartlett's test.
- The test statistic is the model F statistic from the ANOVA on the transformed response $z_{ij} = \left| y_{ij} \overline{y}_i \right|$ where \overline{y}_i is the mean response at ith level.



Brown-Forsythe Test vs. Levene's Test

$$F = \frac{(N-k)}{(k-1)} \frac{\sum_{i=1}^{k} N_i (Z_{i.} - Z_{..})^2}{\sum_{i=1}^{k} \sum_{j=1}^{N_i} (Z_{ij} - Z_{i.})^2}$$

where

N is the total number of observations.

k is the number of groups.

 N_i is the number of observations in the ith group.

 Z_i is the group mean of the ith group.

Z is the grand mean of all the observations.

In Brown-Forsythe Test, $Z_{ij} = \left| Y_{ij} - \widetilde{Y}_{ij} \right|$, where \widetilde{Y}_{ij} is the group median of the ith group. In Levene's Test, $Z_{ij} = \left| Y_{ij} - \overline{Y}_{ij} \right|$, where \overline{Y}_{ij} is the group mean of the ith group.

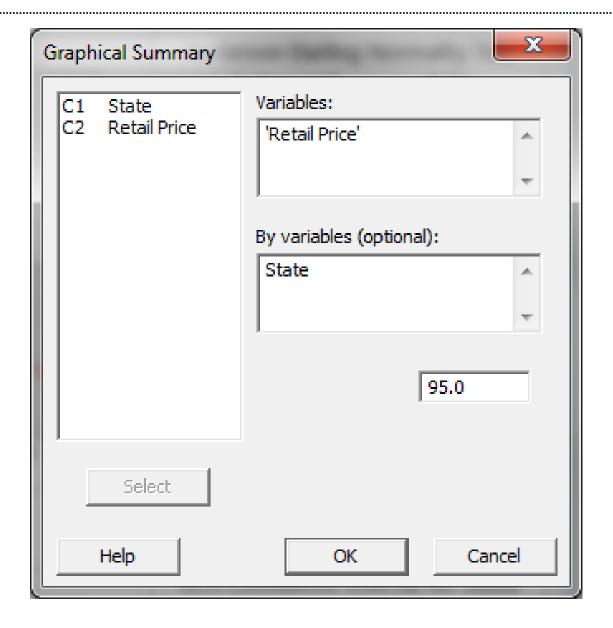


- Case study: We are interested in comparing the variances of the retail price of a product in state A and state B.
 - Data File: "Two-Sample T-Test" tab in "Sample Data.xlsx"
- Null Hypothesis (H₀): $\sigma_A^2 = \sigma_B^2$
- Alternative Hypothesis (H_a): $\sigma_A^2 \neq \sigma_B^2$



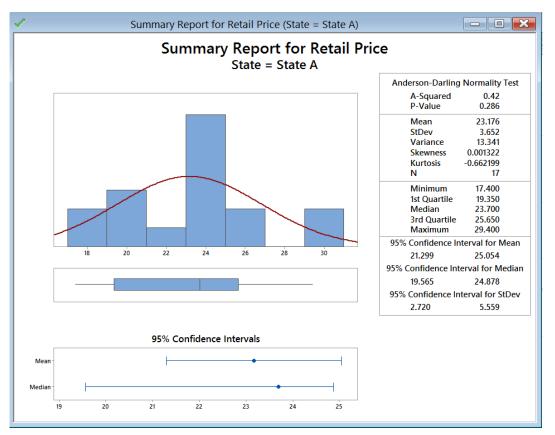
- Step 1: Run the normality test to check whether all levels of data are normally distributed
 - 1) Click Stat → Basic Statistics → Graphical Summary.
 - 2) A new window named "Graphical Summary" pops up.
 - 3) Select "Retail Price" as the "Variables."
 - 4) Click in the blank box right below "By variables" and the "State" appears in the list box on the left.
 - 5) Select "State" as "By variables."
 - 6) Click "OK."
 - 7) The normality test results appear in the new windows.

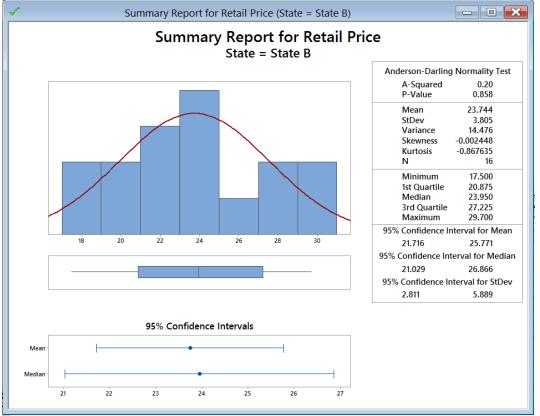






- Null Hypothesis (H0): The data are normally distributed.
- Alternative Hypothesis (Ha): The data are not normally distributed.
- Both retail price data of state A and B are normally distributed since the p-values are both greater than alpha level (0.05).

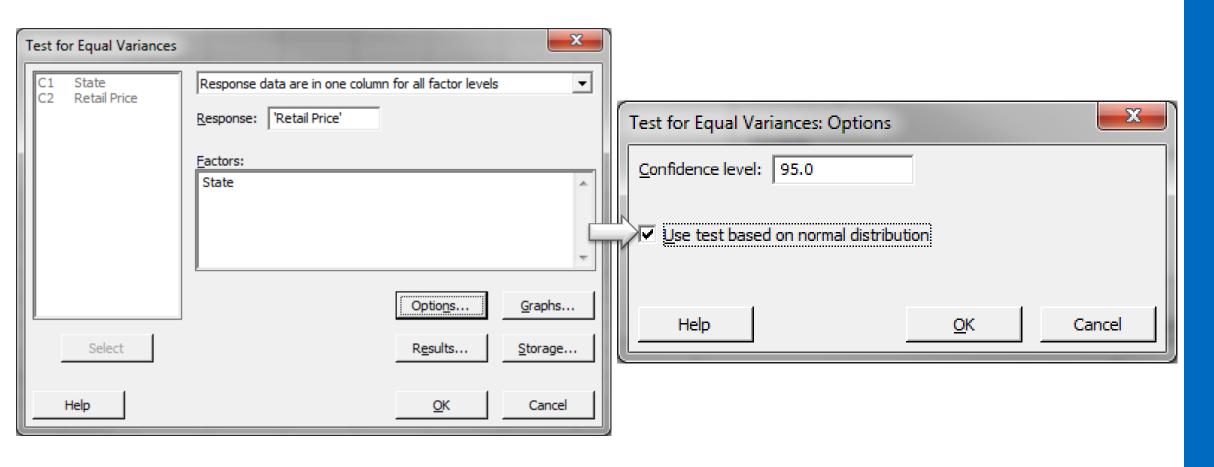






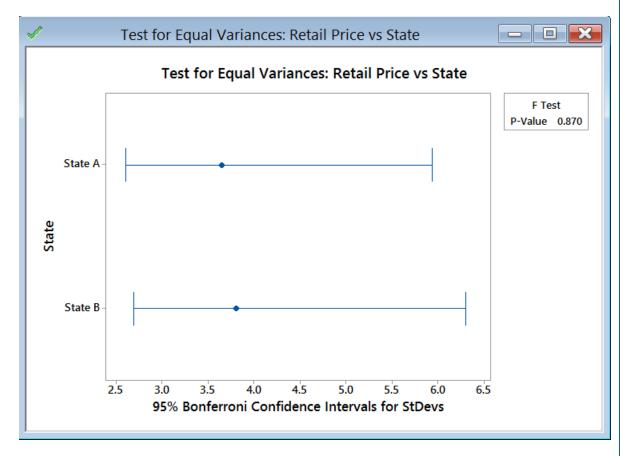
- Step 2: Run tests of equal variance in Minitab
 - 1) Click Stat → ANOVA → Test for Equal Variances.
 - 2) A new window named "Test for Equal Variances" pops up.
 - 3) Select "Retail Price" as the "Response."
 - 4) Select "State" as "Factors."
 - 5) Click "Options"
 - 6) Check "Use test based on normal distribution"
 - 7) Click "OK" to close the Options window
 - 8) Click "OK."
 - 9) The results of variances equality tests appear in the new window.







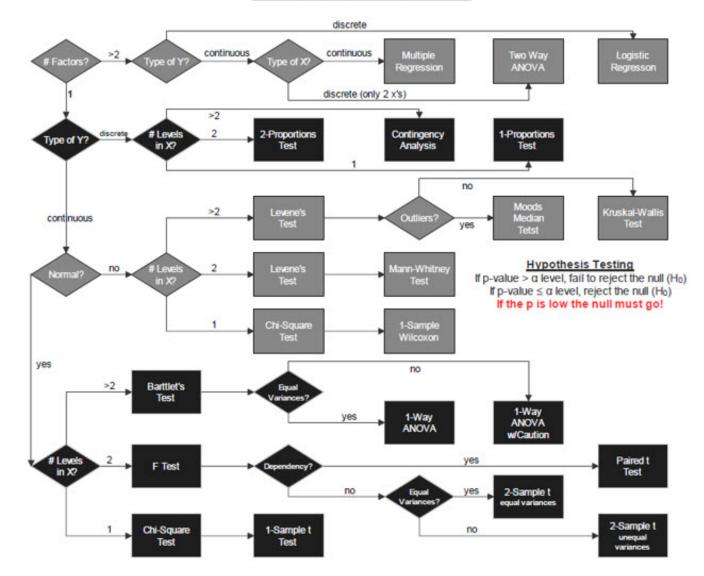
- If all the groups of data are normally distributed, use the F-test or Bartlett's test in Minitab to test the equality of the variances.
- If at least one of the groups is not normally distributed, use Levene's test in Minitab to test the equality of the variances.
- If the p-value of the variances equality test is greater than the alpha level (0.05), we fail to reject the null hypothesis and conclude that the variances of different groups are identical.





Hypothesis Testing Roadmap: Putting it all together

Hypothesis Testing Roadmap





4.0 Improve Phase



Black Belt Training: Improve Phase

4.1 Simple Linear Regression

- 4.1.1 Correlation
- 4.1.2 X-Y Diagram
- 4.1.3 Regression Equations
- 4.1.4 Residuals Analysis

4.2 Multiple Regression Analysis

- 4.2.1 Non-Linear Regression
- 4.2.2 Multiple Linear Regression
- 4.2.3 Confidence Intervals
- 4.2.4 Residuals Analysis
- 4.2.5 Data Transformation, Box Cox



4.1 Simple Linear Regression



Black Belt Training: Improve Phase

4.1 Simple Linear Regression

- 4.1.1 Correlation
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- 4.2.5 Data Transformation, Box Cox



4.1.1 Correlation



What is Correlation?

- Correlation is a statistical technique that describes whether and how strongly two or more variables are related.
- Correlation analysis helps to understand the direction and degree of association between variables, and it suggests whether one variable can be used to predict another.
- Of the different metrics to measure correlation, Pearson's correlation coefficient is the most popular. It measures the linear relationship between two variables.



Pearson's Correlation Coefficient

- Pearson's correlation coefficient is also called:
 - Pearson's r or coefficient of correlation
 - Pearson's product moment correlation coefficient (r)
- "r" is a statistic measuring the linear relationship between two variables.
- Correlation coefficients range from -1 to 1.
 - If r = 0, there is no linear relationship between the variables.
 - The *sign* of r indicates the *direction* of the relationship:
 - If r < 0, there is a negative linear correlation.
 - If r > 0, there is a positive linear correlation.
 - The *absolute value* of r describes the *strength* of the relationship:
 - If $|r| \le 0.5$, there is a weak linear correlation.
 - If |r| > 0.5, there is a strong linear correlation.
 - If |r| = 1, there is a perfect linear correlation.



Pearson's Correlation Coefficient

- When the correlation is *strong*, the data points on a scatter plot will be close together (tight).
 - The closer "r" is to -1 or 1, the stronger the relationship.
 - -1 Strong inverse relationship
 - +1 Strong direct relationship
- When the correlation is *weak*, the data points are spread apart more (loose).
 - The closer the correlation is to 0 the weaker the relationship.



Pearson's Correlation Coefficient

- Pearson's correlation coefficient is only sensitive to the *linear* dependence between two variables.
- It is possible that two variables have a perfect non-linear relationship when the correlation coefficient is low.
- Notice the scatter plots below with correlation equal to 0. There are clearly relationships but they are not linear and therefore can not be determined with Pearson's correlation coefficient.





Correlation and Causation

- Correlation does not imply causation.
- If variable A is highly correlated with variable B, it does not necessarily mean A causes B or vice versa. It is possible that an unknown third variable C is causing both A and B to change.
- For example, if ice cream sales at the beach are highly correlated with the number of shark attacks, it does not imply that increased ice cream sales causes increased shark attacks. They are triggered by a third factor: summer.



Correlation and Dependence

- If two variables are independent, the correlation coefficient is zero.
- WARNING! If the correlation coefficient of two variables is zero, it does not imply they are independent.
- The correlation coefficient only indicates the linear dependence between two variables. When variables are non-linearly related, they are not independent of each other but their correlation coefficient could be zero.



Correlation Coefficient and X-Y Diagram

- The correlation coefficient indicates the direction and strength of the linear dependence between two variables but it does not cover all the existing relationship patterns.
- With the same correlation coefficient, two variables might have completely different dependence patterns.
- A scatter plot or X-Y diagram can help to discover and understand additional characteristics of the relationship between variables.
- Correlation coefficient is not a replacement for examining the scatter plot to study the variables' relationship.



Statistical Significance of the Correlation Coefficient

- The correlation coefficient could be high or low by chance (randomness). It may have been calculated based on two small samples that do not provide good inference on the correlation between two populations.
- In order to test whether there is a statistically significant relationship between two variables, we need to run a hypothesis test to determine whether the correlation coefficient is statistically different from zero.
 - Hypothesis Test Statements
 - H_0 : r = 0: Null Hypothesis: There is *no* correlation.
 - H_1 : $r \neq 0$: Alternate Hypothesis: There is a correlation.



Statistical Significance of the Correlation Coefficient

- Hypothesis tests will produce p-values as a result of the statistical significance test on r.
 - When the p-value for a test is low (less than 0.05), we can reject the null hypothesis and conclude that "r" is significant; there is a correlation.
 - When the p-value for a test is > 0.05, then we fail to reject the null hypothesis; there is no correlation.

 We can also use the t statistic to draw the same conclusions regarding our test for significance of the correlation coefficient.



Statistical Significance of the Correlation Coefficient

• Test Statistic: $t = \frac{r}{\sqrt{\frac{1-r^2}{n-2}}}$

- Critical Statistic: t-value in t-table with (n − 2) degrees of freedom
- If $|t| \le t_{critical}$, we fail to reject the null. There is no statistically significant linear relationship between X and Y.
- If |t| > t_{critical}, we reject the null. There is a statistically significant linear relationship between X and Y.



Using Software to Calculate the Correlation Coefficient

- We are interested in understanding whether there is linear dependence between a car's MPG and its weight and if so, how they are related.
- The MPG and weight data are stored in the "Correlation Coefficient" tab in "Sample Data.xlsx." We will discuss three ways to get the results.

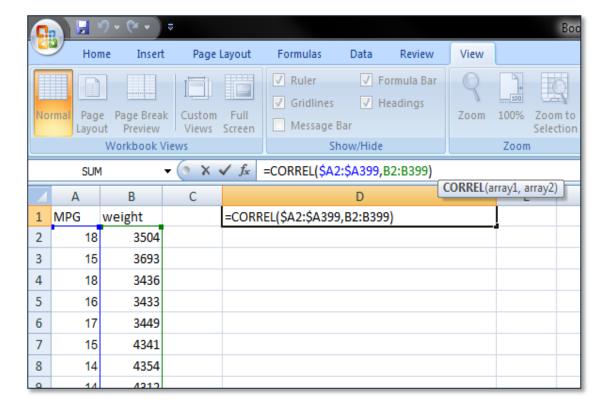


Use Excel to Calculate the Correlation Coefficient

 The formula CORREL in Excel calculates the sample correlation coefficient of two data series.

• The correlation coefficient between the two data series is -0.83, which indicates a strong negative linear relationship between MPG and

weight.



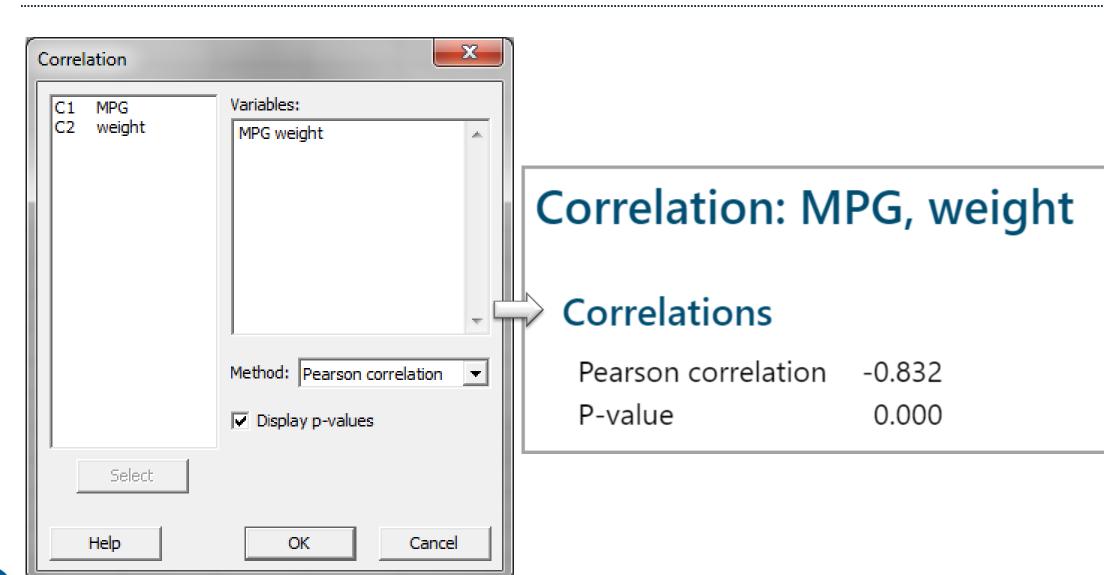


Use Minitab to Calculate the Correlation Coefficient

- Step 1: Stat → Basic Statistics → Correlation.
- Step 2: Select the two variables of interest in the pop-up window "Correlation" and click "OK."
- The correlation coefficient result (-0.832) appears in the session window. The p-value (0.000) is lower than the alpha level (0.05), indicating the linear correlation is statistically significant.



Use Minitab to Calculate the Correlation Coefficient





Interpreting Results

- How do we interpret results and make decisions based Pearson's correlation coefficient (r) and p-values?
 - Let us look at a few examples:
 - r = -0.832, p = 0.000 (previous example). The two variables are inversely related and the linear relationship is strong. Also, this conclusion is significant as supported by p-value of 0.00.
 - r = -0.832, p = 0.71. Based on r, you should conclude the linear relationship between the two variables is strong and inversely related. However, with a p-value of 0.71, you should then conclude that r is not significant and that your sample size may be too small to accurately characterize the relationship.
 - r = 0.5, p = 0.00. Moderately positive linear relationship, r is statistically significant.
 - r = 0.92, p = 0.61. Strong positive linear relationship but r is not statistically significant. Get more data.
 - r = 1.0, p = 0.00. The two variables have a perfect linear relationship and r is significant.



Correlation Coefficient Calculation

Population Correlation Coefficient (ρ)

$$\rho_{X,Y} = \frac{\text{cov}(X,Y)}{\sigma_X \sigma_Y} = \frac{E[(X - \mu_X)(Y - \mu_Y)]}{\sigma_X \sigma_Y}$$

Sample Correlation Coefficient (r)

$$r = \frac{\sum_{i=1}^{n} (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^{n} (X_i - \bar{X})^2} \sqrt{\sum_{i=1}^{n} (Y_i - \bar{Y})^2}}$$

- It is only defined when the standard deviations of both X and Y are non-zero and finite.
- When covariance of X and Y is zero, the correlation coefficient is zero.



4.1.2 X-Y Diagram



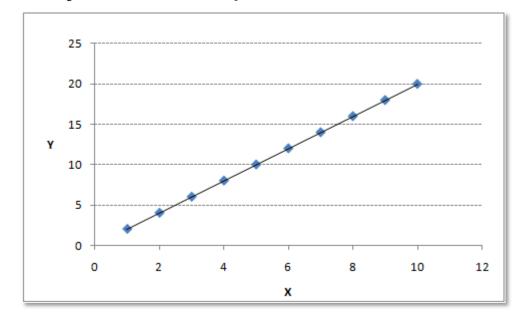
What is an X-Y Diagram?

- An **X-Y diagram** is a scatter plot depicting the relationship between two variables (i.e., X and Y).
- Each point on the X-Y diagram represents a pair of X and Y values,
 with X plotted on the horizontal axis and Y plotted on the vertical axis.
- With an X-Y diagram, you can qualitatively assess both the strength and direction of the relationship between X and Y.
- To quantitatively measure the relationship between X and Y, you may need to calculate the correlation coefficient.



Example 1: Perfect Linear Correlation

- In the chart below, there are 10 data points depicted (10 pairs of X and Y values), and they were created using the equation Y = 2X.
- As a result, all the data points fall onto the straight line of Y = 2X.
- The chart demonstrates a perfect positive linear correlation between X and Y since the relationship between X and Y can be perfectly described by a linear equation in a format of Y = a × X + b where a ≠ 0.

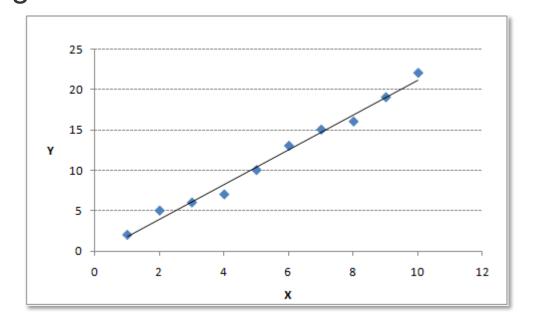


X	Υ
1	2
2	4
3	6
4	8
5	10
6	12
7	14
8	16
9	18
10	20



Example 2: Strong Linear Correlation

- In this chart, the data points scatter closely around a straight line.
- When X increases, Y increases accordingly.
- This chart demonstrates a strong positive linear correlation between X and Y.
- The straight line is the trend line showing how Y's trend goes with changes in X.

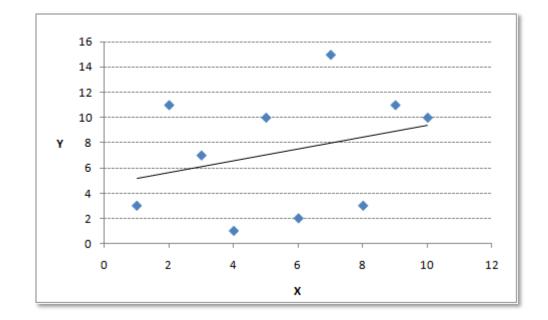


X	Υ
1	2
2	5
3	6
4	7
5	10
6	13
7	15
8	16
9	19
10	22



Example 3: Weak Linear Correlation

- In this chart, the data points scatter remotely around a straight line.
- When X increases, Y increases accordingly.
- This chart demonstrates a weak positive linear correlation between X and Y since the distance between the data points and the trend line is relatively far on average.

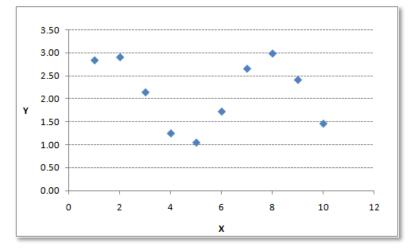


X	Υ
1	3
2	11
3	7
4	1
5	10
6	2
7	15
8	3
9	11
10	10



Example 4: Non-Linear Correlation

 The X-Y diagram also helps to identify any nonlinear relationship between X and Y.



35	; -	 				
30	,	 				
25	; +	 				
Y 20	·	 •			•	
15	;	 ·····		***************************************		
10	·	 		•		
5	;	 				
0) 	 -		-		

Х	Υ
1	35
2	26
3	19
4	14
5	11
6	10
7	11
8	14
9	19
10	26

X

6

9

10

2.84

2.91

2.14 1.24

1.04

1.72

2.66

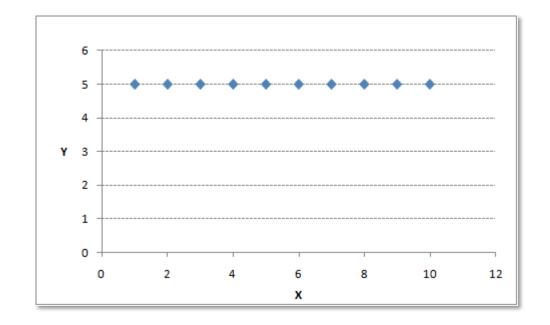
2.41

1.46



Example 5: Uncorrelated

- In this chart, the Y value of each data point is a constant regardless of what the X value is.
- Changes in X do not show any relative impact on Y. As a result, there is no correlation between X and Y.

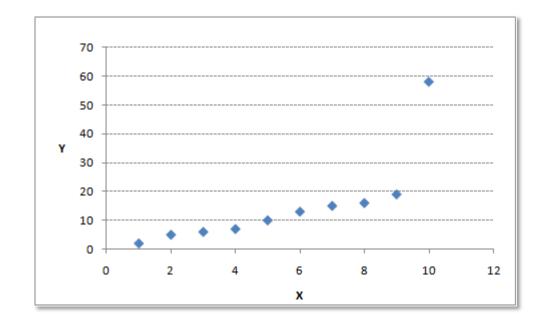


X	Υ
1	5
2	5
3	5
4	5
5	5
6	5
7	5
8	5 5 5
9	5
10	5



Example 6: Outlier Identification

- Using X-Y diagram, you may identify outliers in the data.
- In this chart, the last data point does not seem to follow the trend of other data points.
- This should require further investigation on the last data point to determine whether it is an outlier.



Х	Υ
1	2
2	5
3	6
4	7
5	10
6	13
7	15
8	16
9	19
10	58



Benefits of Using an X-Y Diagram

- An X-Y diagram graphically demonstrates the relationship between two variables.
- It suggests whether two variables are associated and helps to identify the linear or nonlinear correlation between X and Y.
- It captures the strength and direction of the relationship between X and Y.
- It helps identify any outliers in the data.



Limitations of the X-Y Diagram

- Although the X-Y diagram helps to "spot" interesting features in the data, it does not provide any quantitative conclusions about the data and further statistical analysis is needed to:
 - Assess whether the association between variables is statistically significant.
 - Measure the strength of the relationship between variables.
 - Determine whether outliers exist in the data.
 - Quantitatively describe the pattern of the data.



4.1.3 Regression Equations



Correlation and Regression Analysis

- The correlation coefficient answers the following questions:
 - Are two variables correlated?
 - How strong is the relationship between two variables?
 - When one variable increases, does the other variable increase or decrease?
- The correlation coefficient cannot address the following questions:
 - How much does one variable changes when the other variable changes by one unit?
 - How can we set the value of one variable to obtain a targeted value of the other variable?
 - How can we use the relationship between two variables to make predictions?
- The simple linear regression analysis helps to answer these questions.



What is Simple Linear Regression?

- Simple linear regression is a statistical technique to fit a straight line through the data points.
- It models the quantitative relationship between two variables.
- It describes how one variable changes according to the change of another variable.
- Both variables need to be continuous.
- It is simple because only one predictor variable is involved.



Simple Linear Regression Equation

 The simple linear regression analysis fits the data to a regression equation in the form

$$Y = \alpha \times X + \beta + e$$

where:

- Y is the dependent variable (the response) and X is the single independent variable (the predictor).
- α is the slope describing the steepness of the fitting line. β is the intercept indicating the Y value when X is equal to 0.
- e stands for error (residual). It is the difference between the actual Y and the fitted Y (i.e., the vertical difference between the data point and the fitting line).



Ordinary Least Squares

- The **ordinary least square** is a statistical method used in linear regression analysis to find the best fitting line for the data points.
- It estimates the unknown parameters of the regression equation by minimizing the sum of squared residuals (i.e. the vertical difference between the data point and the fitting line).
- In mathematical language, we look for α and β that satisfy the following criteria:

$$\min_{\alpha,\beta} Q(\alpha,\beta) \text{ where } Q(\alpha,\beta) = \sum_{i=1}^{n} e_i^2 = \sum_{i=1}^{n} (y_i - \alpha - \beta x_i)^2$$



Ordinary Least Squares

The actual value of the dependent variable:

$$Y_i = \alpha * X_i + \beta + e_i$$
 where $i = 1, 2, ..., n$

The fitted value of the dependant variable:

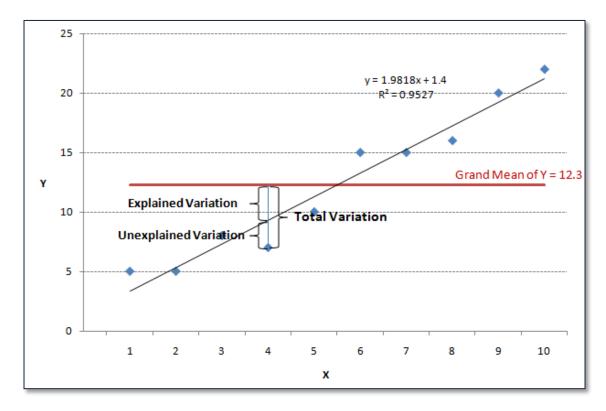
$$\hat{Y}_i = \alpha * X_i + \beta$$
 where $i = 1, 2, ..., n$

 By using calculus, it can be shown the sum of squared error is minimal when

$$\beta = \frac{\sum_{i=1}^{n} (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_{i=1}^{n} (X_i - \bar{X})^2} \qquad \alpha = \bar{Y} - \beta \bar{X}$$



ANOVA in Simple Linear Regression



X: the independent variable that we use to predict;

Y: the dependent variable that we want to predict.

Х	Υ
1	5
2	5
3	8
4	7
5	10
6	15
7	15
8	16
9	20
10	22

Total Variation = Total Sums of Squares =
$$\sum_{i=1}^{n} (Y_i - \bar{Y})^2$$

Explained Variation = Regression Sums of Squares =
$$\sum_{i=1}^{n} (\hat{Y}_i - \bar{Y})^2$$

Unexplained Variation = Error Sums of Squares =
$$\sum_{i=1}^{n} (Y_i - \hat{Y})^2$$



ANOVA in Simple Linear Regression

- Linear regression is also analysis of variance (ANOVA).
 - Variation Components:
 - Total Variation = Explained Variation + Unexplained Variation
 i.e., Total Sums of Squares = Regression Sums of Squares + Error Sums of Squares
 - Degrees of Freedom Components
 - Total Degrees of Freedom = Regression Degrees of Freedom + Residual Degrees of Freedom

i.e., n - 1 = (k - 1) + (n - k), where n is the number of data points, k is the number of predictors



ANOVA in Simple Linear Regression

- Whether the overall model is statistically significant can be tested by using Ftest of ANOVA.
 - H₀: The model is not statistically significant.
 - H_a: The model is statistically significant.

• Test Statistic:
$$F = \frac{MSR}{MSE} = \frac{\sum_{i=1}^{n} (\hat{Y}_{i} - \bar{Y})^{2} / (k-1)}{\sum_{i=1}^{n} (\hat{Y}_{i} - \hat{Y}_{i})^{2} / (n-k)}$$

- Critical Statistic: F value in F table with (k 1) degrees of freedom in the numerator and (n – k) degrees of freedom in the denominator.
- If $F \le F_{critical}$, we fail to reject the null. There is no statistically significant relationship between X and Y.
- If F > F_{critical}, we reject the null. There is a statistically significant relationship between X and Y.



Coefficient of Determination

- R² (also called coefficient of determination) measures the proportion of variability in the data that can be explained by the model.
- R² ranges from 0 to 1. The higher R² is, the better the model can fit the actual data.
- How to calculate R²:

$$R^{2} = \frac{SS_{regression}}{SS_{total}} = 1 - \frac{SS_{error}}{SS_{total}} = 1 - \frac{\sum_{i=1}^{n} (Y_{i} - Y)^{2}}{\sum_{i=1}^{n} (Y_{i} - \bar{Y})^{2}}$$

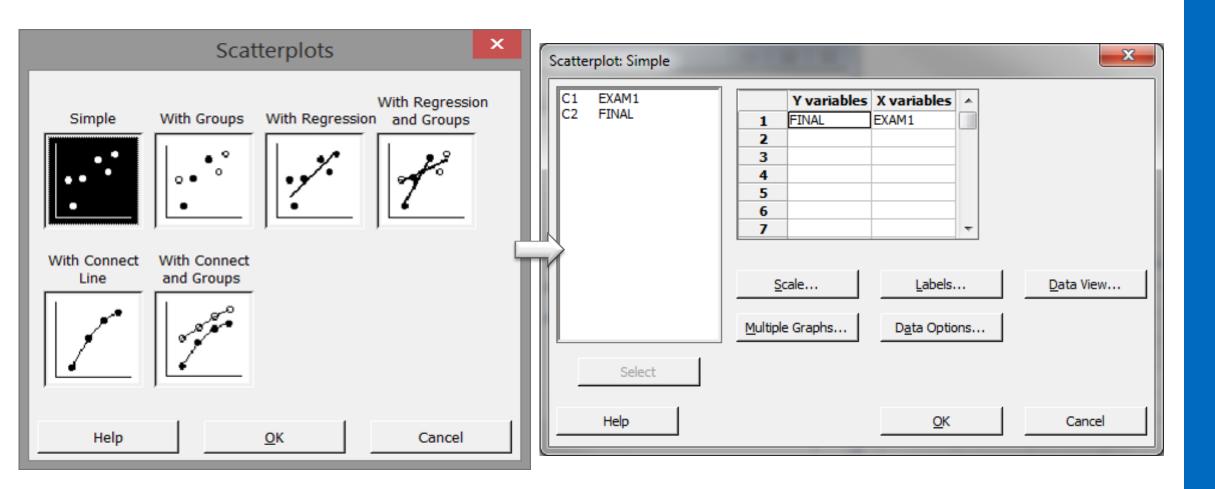


- Case study:
 - We want to see whether the score on exam one has any statistically significant relationship with the score on the final exam. If yes, how much impact does exam one have on the final exam?
 - Data File: "Simple Linear Regression" tab in "Sample Data.xlsx"
- Step 1: Determine the dependent and independent variables. Both should be continuous variables.
 - Y (dependent variable) is the score of final exam.
 - X (independent variable) is the score of exam one.
 - Both variables are continuous.



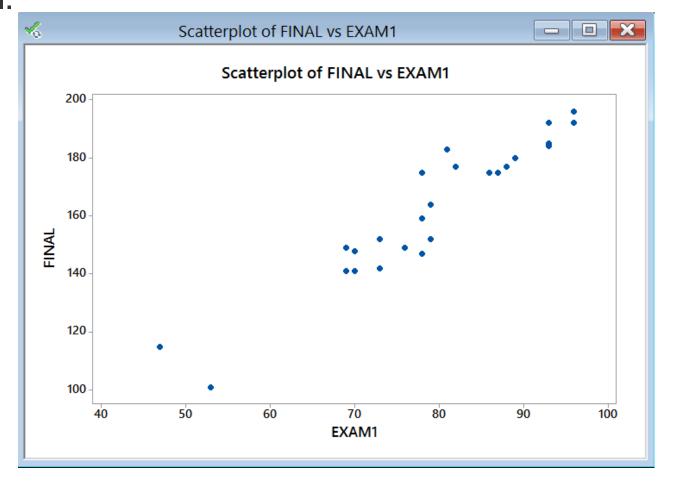
- Step 2: Create a scatter plot to visualize whether there seems to be a linear relationship between X and Y.
 - Click Graph → Scatterplot.
 - 2) A new window named "Scatterplots" pops up.
 - Click "OK."
 - 4) A new window named "Scatterplot Simple" pops up.
 - 5) Select "FINAL" as "Y variables" and "EXAM1" as "X variables."
 - 6) Click "OK."
 - 7) A scatter plot is generated in a new window.







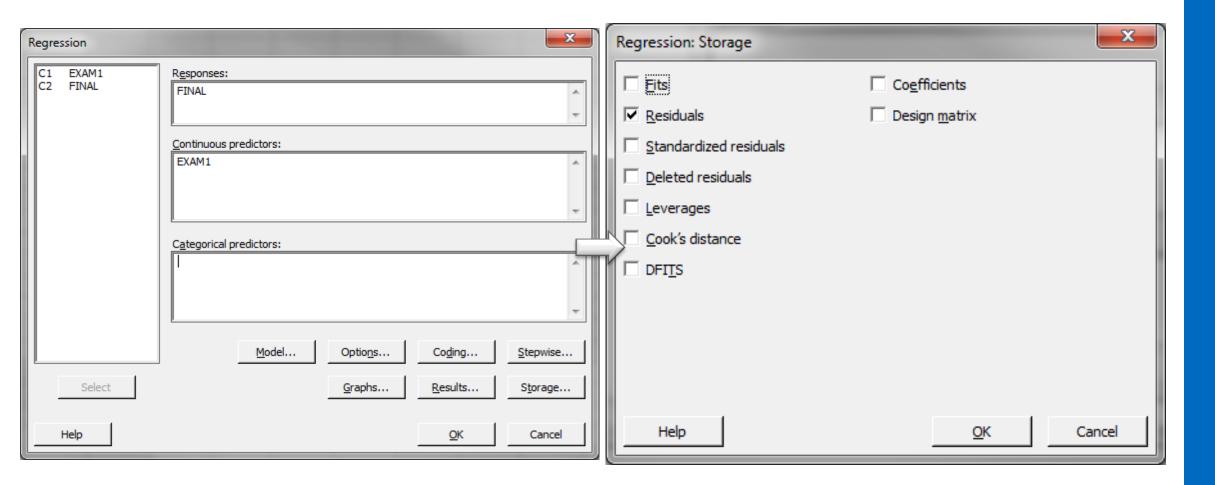
 Based on the scatter plot, the relationship between exam one and final seems linear. The higher the score on exam one, the higher the score on the final.





- Step 3: Run the simple linear regression analysis.
 - 1) Click Stat → Regression → Regression → Fit Regression Model
 - 2) A new window named "Regression" pops up.
 - 3) Select "FINAL" as "Response" and "EXAM1" as "Continuous Predictors."
 - 4) Click the "Storage" button.
 - 5) Check the box of "Residuals" so that the residuals can be saved automatically in the last column of the data table.
 - 6) Click "OK."
 - 7) The regression analysis results appear in the new window.







- Step 4: Check whether the model is statistically significant. If not significant, we will need to re-examine the predictor or look for new predictors before continuing.
- R2 measures the percentage of variation in the data set that can be explained by the model. 89.5% of the variability in the data can be accounted for by this linear regression model.
- "Analysis of Variance" section provides a ANOVA table covering degrees of freedom, sum of squares, and mean square information for total, regression and error.
- The p-value of the F-test is lower than the α level (0.05), indicating that the model is statistically significant.

Regression Analysis: FINAL versus EXAM1

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	1	12418.8	12418.8	196.15	0.000
EXAM1	1	12418.8	12418.8	196.15	0.000
Error	23	1456.2	63.3		
Lack-of-Fit	14	837.0	59.8	0.87	0.607
Pure Error	9	619.2	68.8		
Total	24	13875.0			

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
7.95689	89.51%	89.05%	86.61%

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	15.6	10.6	1.48	0.153	
EXAM1	1.852	0.132	14.01	0.000	1.00

Regression Equation

FINAL = 15.6 + 1.852 EXAM1

Fits and Diagnostics for Unusual Observations

				Std		
Obs	FINAL	Fit	Resid	Resid		
8	115.00	102.69	12.31	1.88		Х
18	183.00	165.67	17.33	2.22	R	



- Step 5: Understand regression equation
 - The estimates of slope and intercept are shown in the "Parameter Estimate" section.
 - In this example, $Y = 15.6 + 1.85 \times X$.
 - One unit increase in the score of Exam1 would increase the final score by 1.85.

Regression Analysis: FINAL versus EXAM1

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	1	12418.8	12418.8	196.15	0.000
EXAM1	1	12418.8	12418.8	196.15	0.000
Error	23	1456.2	63.3		
Lack-of-Fit	14	837.0	59.8	0.87	0.607
Pure Error	9	619.2	68.8		
Total	24	13875.0			

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
7.95689	89.51%	89.05%	86.61%

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Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	15.6	10.6	1.48	0.153	
EXAM1	1.852	0.132	14.01	0.000	1.00

Regression Equation

FINAL = 15.6 + 1.852 EXAM1

Fits and Diagnostics for Unusual Observations

				Std		
Obs	FINAL	Fit	Resid	Resid		
8	115.00	102.69	12.31	1.88		Χ
18	183.00	165.67	17.33	2.22	R	



Interpreting the Results

- Rsquare Adj = 89.05%
 - 89% of the variation in FINAL can be explained by EXAM1
- P-value of the F-test = 0.000
 - We have a statistically significant model
- Prediction Equation: 15.6 + 1.85 × EXAM1
 - 15.6 is the Y intercept, all equations will start with 15.6
 - 1.85 is the EXAM1 Coefficient: multiply it by EXAM1 score



 Let us say you are the professor and you want to use this prediction equation to estimate what two of your students might get on their final exam.



Interpreting the Results

- Let us assume the following:
 - Student "A" exam 1 results were: 79
 - Student "B" exam 1 results were: 94.
- Remember our prediction equation?
 - 15.6 + 1.85 × Exam1
 - Now apply the equation to each student
 - Student "A" Estimate: 15.6 + (1.85 × 79) = 161.8
 - Student "B" Estimate: $15.6 + (1.85 \times 94) = 189.5$



Now you can use your "magic" as the professor and allocate your time appropriately to the student(s) who you predict need the most help Nice Work!

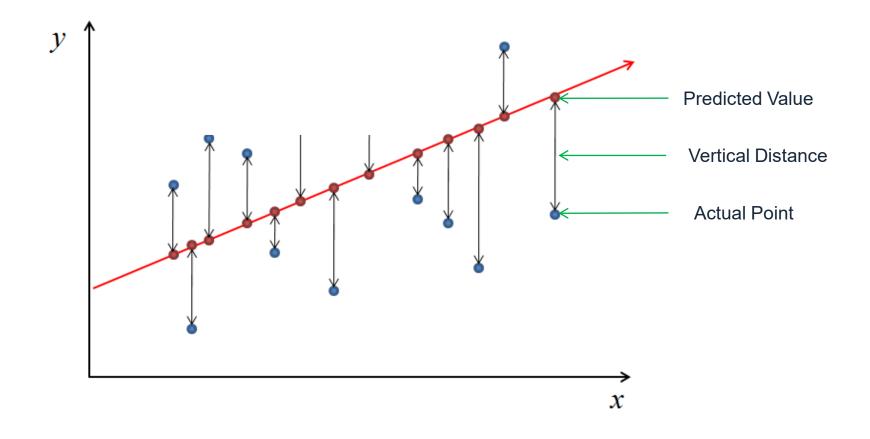


4.1.4 Residuals Analysis



What are Residuals?

• **Residuals** are the vertical differences between actual values and the predicted values or the "fitted line" created by the regression model.





Why Perform Residuals Analysis?

- Regression equations are generated on the basis of certain statistical assumptions.
- Residuals analysis helps to determine the validity of these assumptions.
- The assumptions are:
 - The residuals are normally distributed, mean equal to zero.
 - The residuals are independent.
 - The residuals have a constant variance.
 - The underlying population relationship is linear.
- If residuals performance does not meet the requirements, we will need to rebuild the model by replacing the predictor with a new one, adding new predictors, building non-linear models, and so on.



Use Minitab to Perform Residuals Analysis

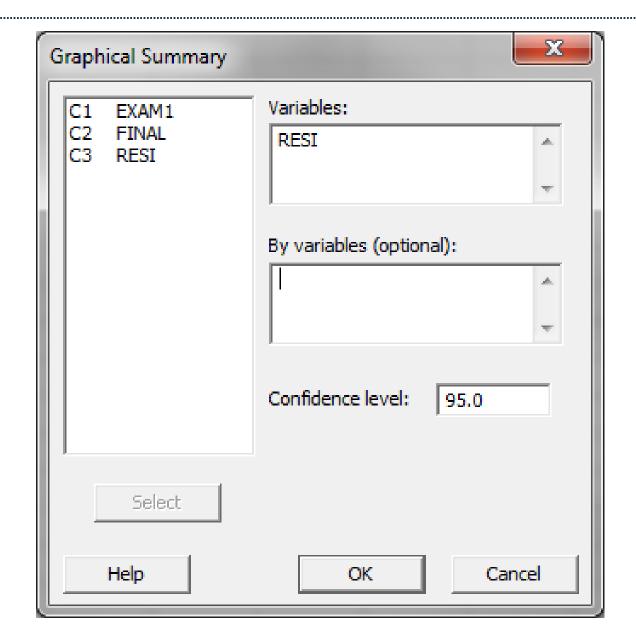
The residuals of the model are saved in a column of your data table

Worksheet 9 ***					
+	C1	C2 🗾	C3		
	EXAM1	FINAL	RESI		
1	73	152	1.1488		
2	93	185	-2.9003		
3	89	180	-0.4905		
4	96	196	2.5424		
5	73	142	-8.8512		
6	53	101	-12.8021		
7	69	149	5.5586		
8	47	115	12.3127		
9	87	175	-1.7855		
10	79	164	2.0341		
11	69	141	-2.4414		



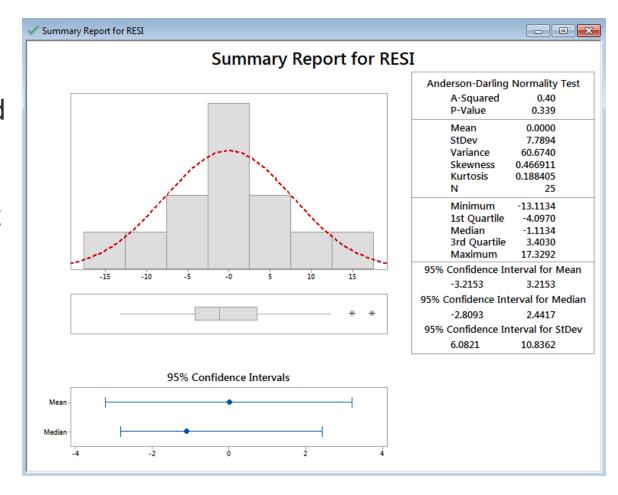
- Step 1: Check whether residuals are normally distributed around the mean of zero.
 - 1) Click Stat → Basic Statistics → Graphical Summary.
 - 2) A new window named "Graphical Summary" pops up.
 - 3) Select "RESI" as "Variables."
 - 4) Click "OK."
 - 5) The histogram and the normality test of the residuals are displayed in the new window.







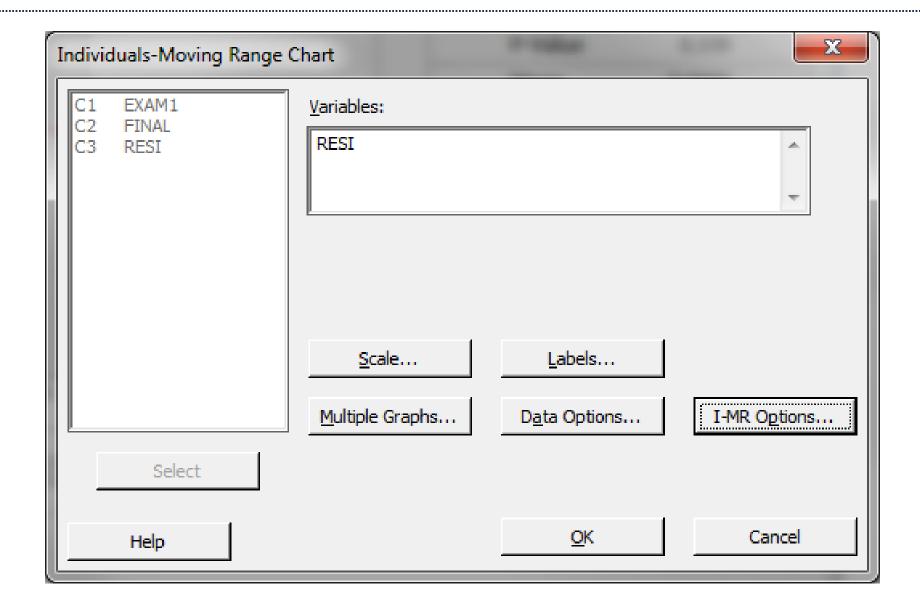
- The mean of residuals is -0.0000.
- The Anderson-Darling test is used to test the normality. Since the p-value (0.339) is greater than the alpha level (0.05), we fail to reject the null hypothesis; the residuals are normally distributed.
 - H₀: The residuals are normally distributed.
 - H₁: The residuals are not normally distributed.





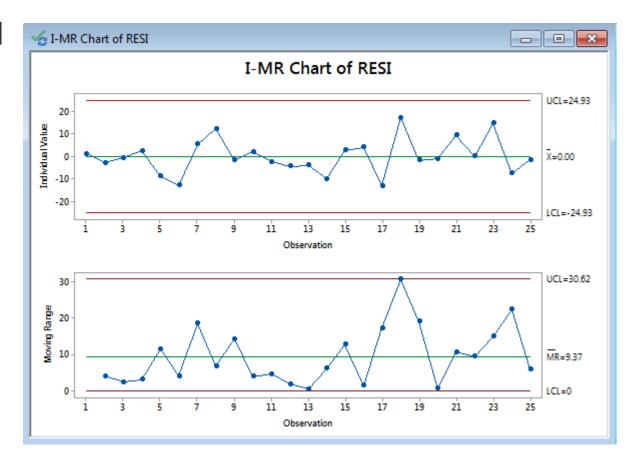
- Step 2: If the data are in time order, run the I-MR chart to check whether residuals are independent.
 - 1) Click Stat \rightarrow Control Charts \rightarrow Variable Charts for Individuals \rightarrow I-MR.
 - 2) A new window named "Individuals Moving Range Chart" pops up.
 - Select "Residuals" as "Variables" and check the box of "Test for special causes."
 - 4) Click "OK."
 - 5) The control charts are shown automatically in the new window.







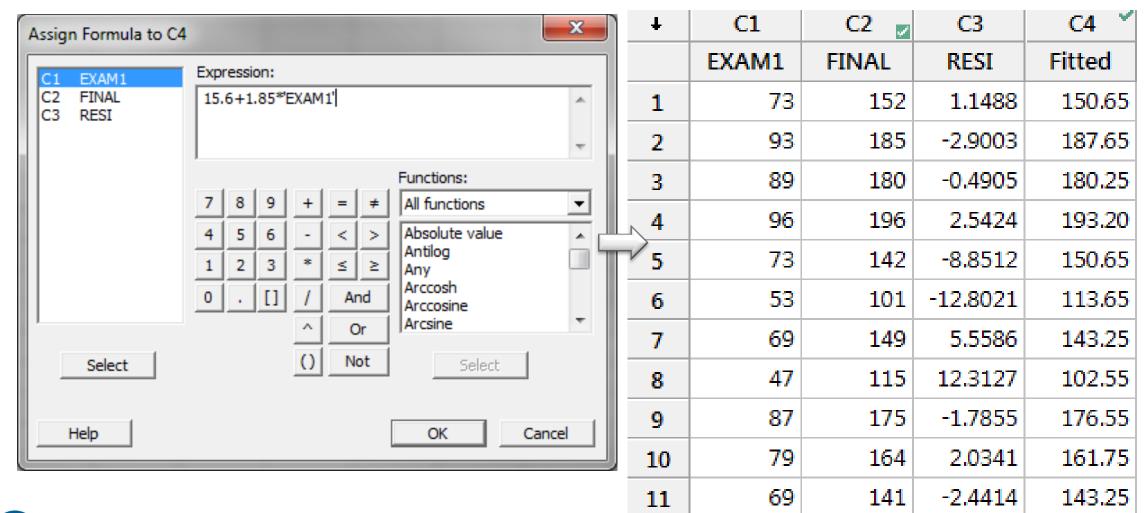
- If no data points are out of control in both the I-chart and MR chart, the residuals are independent of each other.
- If the residuals are not independent, it is possible that some important predictors are not included in the model.
- In this example, since the I-MR chart is in control, residuals are independent.



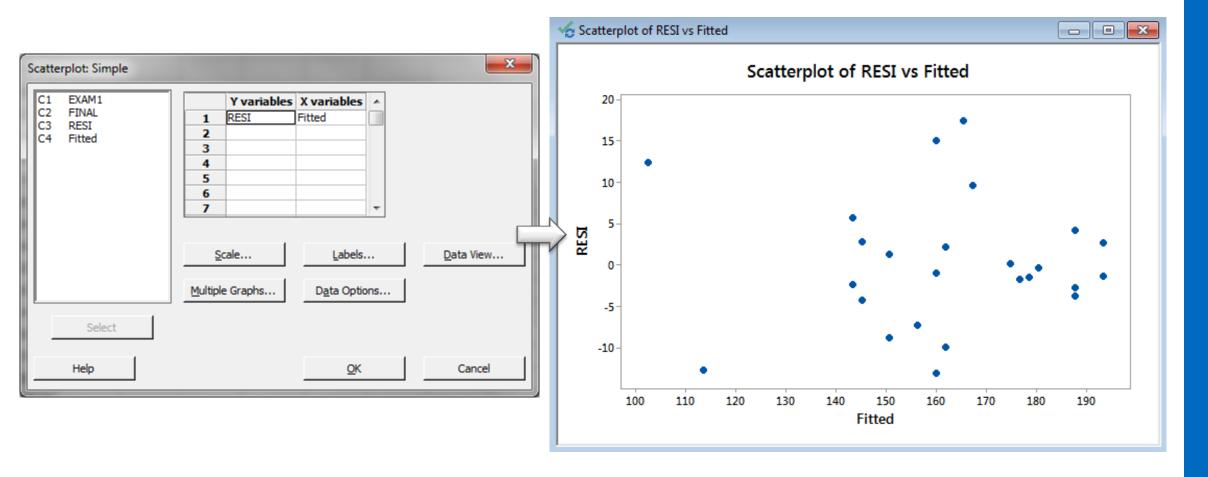


- Step 3: Check whether residuals have equal variance across the predicted responses.
 - Create a column of the fitted response using the regression formula returned in the session window.
 - 1) Right click on the last column in the data table.
 - 2) Select Formulas → Assign Formula to Column.
 - 3) A new window named "Assign Formula to C4" appears.
 - 4) Enter the regression equation "15.6+1.85*EXAM1" into the "Expression" box.
 - 5) Click "OK."
 - 6) The column of the fitted values is created in the data table.
 - Create a scatter plot with Y being the residuals and the X being the fitted values.
 - 1) Click Graph → Scatterplot.
 - 2) A new window named "Scatterplots" appears.
 - 3) Select "RESI" as the "Y variables" and "Fitted" as the "X variables."
 - 4) Click "OK."
 - 5) The scatter plot appears in a new window.
 - We are looking for the pattern in which residuals spread out evenly around zero from the top to the bottom.











4.2 Multiple Regression Analysis



Black Belt Training: Improve Phase

4.1 Simple Linear Regression

- 4.1.1 Correlation
- 4.1.2 X-Y Diagram
- 4.1.3 Regression Equations
- 4.1.4 Residuals Analysis

4.2 Multiple Regression Analysis

- 4.2.1 Non-Linear Regression
- 4.2.2 Multiple Linear Regression
- 4.2.3 Confidence Intervals
- 4.2.4 Residuals Analysis
- 4.2.5 Data Transformation, Box Cox



4.2.1 Non-Linear Regression



Linear and Non-Linear

- The word *linear* originally comes from Latin word *linearis* meaning "created by lines."
- A linear function in mathematics follows the following pattern (i.e., the output is proportional to its input):

$$f(x) = \alpha * x + \beta$$

$$f(x_1, x_2, ..., x_n) = \alpha_1 * x_1 + \alpha_2 * x_2 + ... + \alpha_n * x_n + \beta$$

• A non-linear function does not follow the above pattern. There are usually exponents, logarithms, power, polynomial components, and other non-linear functions of the independent variables and parameters.



Non-Linear Relationships Using Linear Models

- Many non-linear relationships can be transformed into linear relationships, and from there we can use linear regression methods to model the relationship.
- Some non-linear relationships cannot be transformed to linear ones and we need to apply other methods to build the non-linear models.
- In this section, we will focus on building non-linear regression models using linear transformation (i.e., transforming the independent or dependent variables or parameters to generate a linear function).



Assumptions in Using Non-Linear Regression

- The population relationship is non-linear based on a reliable underlying theory.
- Across the range of all the possible values of the independent variables, the non-linear relationship applies. It is possible that at some extreme values the relationship between variables changes dramatically.



Non-Linear Functions: Transforming to Linear

- Examples of non-linear functions that can be transformed to linear functions:
 - Exponential Function
 - Inverse Function
 - Polynomial Function
 - Power Function.



Exponential Function

Exponential Function

$$Y = a \times b^X$$

$$\log Y = \log a + X \times \log b$$



Inverse Function

Inverse Function

$$Y = a + b \times \frac{1}{X}$$

$$Y = a + b \times Z$$
 where $Z = \frac{1}{X}$



Polynomial Function

Polynomial Function

$$Y = a + b \times X + c \times X^2$$

$$Y = a + b \times X + c \times Z$$
 where $Z = X^2$



Power Function

Power Function

$$Y = a \times X^b$$

$$\log Y = \log a + b \times \log X$$



4.2.2 Multiple Linear Regression



What is Multiple Linear Regression?

- Multiple linear regression is a statistical technique to model the relationship between one dependent variable and two or more independent variables by fitting the data set into a linear equation.
- The difference between simple linear regression and multiple linear regression:
 - Simple linear regression only has one predictor.
 - Multiple linear regression has two or more predictors.



Multiple Linear Regression Equation

$$Y = \alpha_1 * X_1 + \alpha_2 * X_2 + ... + \alpha_p * X_p + \beta + e$$

- Y is the dependent variable (response).
- $X_1, X_2, ..., X_p$ are the independent variables (predictors). There are p predictors in total.
- Both dependent and independent variables are continuous.
- β is the intercept indicating the Y value when all the predictors are zeros.
- α_1 , α_2 , ..., α_p are the coefficients of predictors. They reflect the contribution of each independent variable in predicting the dependent variable.
- *e* is the residual term indicating the difference between the actual and the fitted response value.

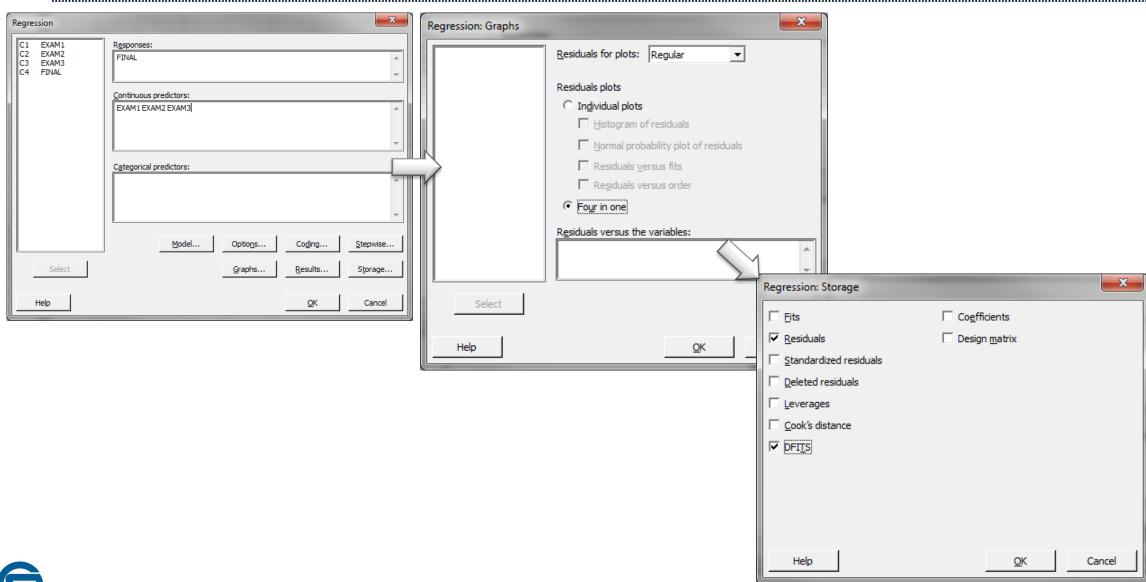


- Case study:
 - We want to see whether the scores in exam one, two, and three have any statistically significant relationship with the score in final exam. If so, how are they related to final exam score? Can we use the scores in exam one, two, and three to predict the score in final exam?
 - Data File: "Multiple Regression Analysis" tab in "Sample Data.xlsx."
- Step 1: Determine the dependent and independent variables. All should be continuous.
 - Y (dependent variable) is the score of final exam.
 - X₁, X₂, and X₃ (independent variables) are the scores of exam one, two, and three respectively.
 - All the variables are continuous.



- Step 2: Start building the multiple linear regression model
 - 1) Click Stat → Regression → Regression → Fit Regression Model
 - 2) A new window named "Regression" pops up.
 - 3) Select "FINAL" as "Response" and "EXAM1", "EXAM2" and "EXAM3" as "Continuous Predictors."
 - 4) Click the "Graph" button, select the radio button "Four in one" and click "OK."
 - 5) Click the "Storage" button, check the boxes of "Residuals" and "DFITS" and click "OK."
 - 6) Click "OK" in the window named "Regression."
 - 7) The regression analysis results appear in the session window and the four residual plots appear in another window named "Residual Plots for FINAL."







- Step 3: Check whether the whole model is statistically significant. If not, we need to re-examine the predictors or look for new predictors before continuing.
 - H₀: The model is not statistically significant (i.e., all the parameters of predictors are not significantly different from zeros).
 - H₁: The model is statistically significant (i.e., at least one predictor parameter is significantly different from zero).
 - In this example, p-value is much smaller than alpha level (0.05), hence we reject the null hypothesis; the model is statistically significant.

Regression Analysis: FINAL versus EXAM1, EXAM2, EXAM3

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	3	13731.5	4577.17	670.09	0.000
EXAM1	1	58.7	58.73	8.60	0.008
EXAM2	1	197.7	197.67	28.94	0.000
EXAM3	1	877.3	877.30	128.43	0.000
Error	21	143.4	6.83		
Total	24	13875.0			

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
2.61357	98.97%	98.82%	98.51%

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	-4.34	3.76	-1.15	0.262	
EXAM1	0.356	0.121	2.93	0.008	7.81
EXAM2	0.543	0.101	5.38	0.000	5.59
EXAM3	1.167	0.103	11.33	0.000	5.16

Regression Equation

FINAL = -4.34 + 0.356 EXAM1 + 0.543 EXAM2 + 1.167 EXAM3

Fits and Diagnostics for Unusual Observations

				Std	
Obs	FINAL	Fit	Resid	Resid	
23	175.00	167.69	7.31	2.93	R



- Step 4: Check whether multicollinearity exists in the model.
 - The VIF information is automatically generated in the table Coefficients.

Regression Analysis: FINAL versus EXAM1, EXAM2, EXAM3

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	3	13731.5	4577.17	670.09	0.000
EXAM1	1	58.7	58.73	8.60	0.008
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Regression Equation

FINAL = -4.34 + 0.356 EXAM1 + 0.543 EXAM2 + 1.167 EXAM3

Fits and Diagnostics for Unusual Observations

				Std	
Obs	FINAL	Fit	Resid	Resid	
23	175.00	167.69	7.31	2.93	R



Multicollinearity

- Multicollinearity is the situation when two or more independent variables in a multiple regression model are correlated with each other.
- Although multicollinearity does not necessarily reduce the predictability for the model as a whole, it may mislead the calculation for individual independent variables.
- To detect multicollinearity, we use VIF (Variance Inflation Factor) to quantify its severity in the model.



Variance Inflation Factor

- VIF quantifies the degree of multicollinearity for each individual independent variable in the model.
- VIF calculation:
 - Assume we are building a multiple linear regression model using p predictors. $Y = \alpha_1 \times X_1 + \alpha_2 \times X_2 + ... + \alpha_p \times X_p + \beta$
 - Two steps are needed to calculate VIF for X₁.
 - Step 1: Build a multiple linear regression model for X_1 by using $X_2, X_3, ..., X_p$ as predictors. $X_1 = a_2 \times X_2 + a_3 \times X_3 + ... + a_p \times X_p + b$
 - Step 2: Use the R² generated by the linear model in step 1 to calculate the VIF for X₁. $VIF = \frac{1}{1 R^2}$
 - Apply the same methods to obtain the VIFs for other Xs. The VIF value ranges from one to positive infinity.



Variance Inflation Factor

- Rules of thumb to analyze variance inflation factor (VIF):
 - If VIF = 1, there is no multicollinearity.
 - If 1 < VIF < 5, there is small multicollinearity.
 - If VIF ≥ 5, there is medium multicollinearity.
 - If VIF ≥ 10, there is large multicollinearity.



How to Deal With Multicollinearity

- Increase the sample size.
- Collect samples with a broader range for some predictors.
- Remove the variable with high multicollinearity and high p-value.
- Remove variables that are included more than once.
- Combine correlated variables to create a new one.
- In this section, we will focus on removing variables with high VIF and high p-value.



- Step 5: Deal with multicollinearity:
 - Step 5.1: Identify a list of independent variables with VIF higher than 5. If no variable has VIF higher than 5, go to Step 6 directly.
 - Step 5.2: Among variables identified in Step 5.1, remove the one with the highest p-value.
 - Step 5.3: Run the model again, check the VIFs and repeat Step 5.1.
 - Note: we only remove one independent variable at a time.



- Step 6: Identify the statistically insignificant predictors. Remove one insignificant predictor at a time and run the model again. Repeat this step until all the predictors in the model are statistically significant.
 - Insignificant predictors are the ones with p-value higher than alpha level (0.05). When p > alpha level, we fail to reject the null hypothesis; the predictor is not significant.
 - H₀: The predictor is not statistically significant.
 - H₁: The predictor is statistically significant.
- As long as the p-value is greater than 0.05, remove the insignificant variables one at a time in the order of the highest p-value.
- Once one insignificant variable is eliminated from the model, we need to run the model again to obtain new p-values for other predictors left in the new model.



• In this example, all three predictors have VIF higher than 5. Among them, EXAM1 has the highest p-value.

 We will remove EXAM1 from the equation and run the model again.

Regression Analysis: FINAL versus EXAM1, EXAM2, EXAM3

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	3	13731.5	4577.17	670.09	0.000
EXAM1	1	58.7	58.73	8.60	0.008
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Error	21	143.4	6.83		
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EXAM3	1.167	0.103	11.33	0.000	5.16

Regression Equation

FINAL = -4.34 + 0.356 EXAM1 + 0.543 EXAM2 + 1.167 EXAM3

Fits and Diagnostics for Unusual Observations

				Std	
Obs	FINAL	Fit	Resid	Resid	
23	175.00	167.69	7.31	2.93	R



- Run the new multiple linear regression with only two predictors (i.e., EXAM2 and EXAM3).
- Check the VIFs of EXAM2 AND EXAM3. They are both smaller than 5; hence, there is little multicollinearity existing in the model.



Use Minitab to Run a Multiple Linear Regression

• In this example, both predictors' p-values are smaller than alpha level (0.05).

 As a result, we do not need to eliminate any variables from the model.

Regression Analysis: FINAL versus EXAM2, EXAM3

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	13672.8	6836.39	743.91	0.000
EXAM2	1	555.1	555.11	60.41	0.000
EXAM3	1	1686.0	1685.97	183.46	0.000
Error	22	202.2	9.19		
Total	24	13875.0			

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
3.03146	98.54%	98.41%	98.18%

Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VIF
Constant	-4.34	4.37	-0.99	0.331	
EXAM2	0.7222	0.0929	7.77	0.000	3.53
EXAM3	1.3375	0.0987	13.54	0.000	3.53

Regression Equation

FINAL = -4.34 + 0.7222 EXAM2 + 1.3375 EXAM3

Fits and Diagnostics for Unusual Observations

				Std	
Obs	FINAL	Fit	Resid	Resid	
17	147.00	140.78	6.22	2.23	R



Use Minitab to Run a Multiple Linear Regression

- Step 7: Interpret the regression equation
 - The multiple linear regression equation appears automatically at the top of the session window.
 - "Coefficients" section provides the estimates of parameters in the linear regression equation.

Regression Analysis: FINAL versus EXAM2, EXAM3

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Regression	2	13672.8	6836.39	743.91	0.000
EXAM2	1	555.1	555.11	60.41	0.000
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Regression Equation

FINAL = -4.34 + 0.7222 EXAM2 + 1.3375 EXAM3

Fits and Diagnostics for Unusual Observations

				Std	
Obs	FINAL	Fit	Resid	Resid	
17	147.00	140.78	6.22	2.23	R



Interpreting the Results

- Rsquare Adj = 98.4%
 - 98% of the variation in FINAL can be explained by the predictor variables EXAM2 & EXAM3.
- P-value of the F-test = 0.000
 - We have a statistically significant model.
- Variables p-value:
 - Both are significant (less than 0.05).
- VIF
 - EXAM2 & EXAM3 are both below 5; we're in good shape!
- Equation: -4.34 + 0.722*EXAM2 + 1.34*EXAM3
 - -4.34 is the Y intercept, all equations will start with -4.34.
 - 0.722 is the EXAM2 coefficient; multiply it by EXAM2 score.
 - 1.34 is the EXAM3 coefficient; multiply it by EXAM3 score.



Interpreting the Results



- Let us say you are the professor again, and this time you want to use your prediction equation to estimate what one of your students might get on their final exam.
- Assume the following:
 - Exam 2 results were: 84
 - Exam 3 results were: 102.
- Use your equation: -4.34 + 0.722*EXAM2 + 1.34*EXAM3
- Predict your student's final exam score:
 - -4.34 + (0.722*84) + (1.34*102) = -4.34 + 60.648 + 136.68 =**192.988**



Nice work again! Now you can use your "magic" as the smart and efficient professor and allocate your time to other students because this one projects to perform much better than the average score of 162.



4.2.3 Confidence & Prediction Intervals



Prediction

- The purpose of building a regression model is not only to understand what happened in the past but more importantly to *predict* the future based on the past.
- By plugging the values of independent variables into the regression equation, we obtain the estimation/prediction of the dependent variable.



Uncertainty of Prediction

- We build the regression model using the sample data to describe as close as possible the true population relationship between dependent and independent variables.
- Due to noise in the data, the prediction will probably differ from the true response value.
- However, the true response value might fall in a range around the prediction with some certainty.
- To measure the uncertainty of the prediction, we need confidence interval and prediction interval.



Confidence Interval

- The **confidence interval** of the prediction is a range in which the population mean of the dependent variable would fall with some certainty, given specified values of the independent variables.
- The width of confidence interval is related to:
 - Sample size
 - Confidence level
 - Variation in the data.
- We build the model based on a sample set $\{y_1, y_2, ..., y_n\}$. The confidence interval is used to estimate the value of the population mean μ of the underlying population.
- The focus of the confidence interval are the unobservable population parameters.
- The confidence interval accounts for the uncertainty in the estimates of regression parameters.



Prediction Interval

- The prediction interval is a range in which future values of the dependent variable would fall with some certainty, given specified values of the independent variables.
- We build the model based on a sample set $\{y1, y2,..., yn\}$. The prediction interval is used to estimate the value of future observation y_{n+1} .
- The focus of the prediction interval are the future observations.
- Prediction interval is wider than confidence interval because it accounts for the uncertainty in the estimates of regression parameters and the uncertainty of the new measurement.

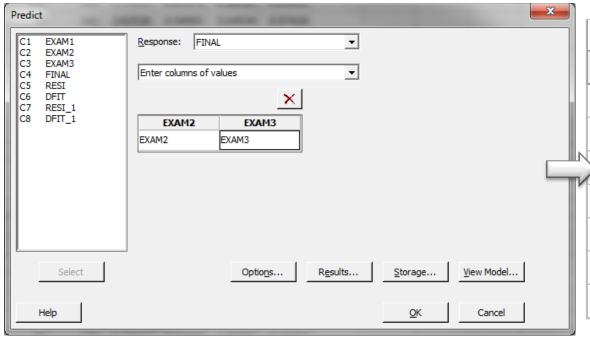


Use Minitab to Obtain Prediction & Confidence Interval

- Steps in Minitab to obtain prediction, confidence interval and prediction interval after running Regression tests.
 - 1. Click Stat → Regression → Predict
 - 2. Choose "FINAL" as Response
 - 3. Select "Enter columns of values" from second dropdown menu
 - 4. Click on "Exam 2" for the first column and "Exam 3" for the second column
 - Click "OK"
 - 6. The fitted response (PFIT), the confidence interval (CLIM), and the prediction interval (PLIM) are automatically added to your data table.



Use Minitab to Obtain Prediction & Confidence Interval



	C9	C10	C11	C12	C13	C14
	PFITS	PSEFITS	CLIM	CLIM_1	PLIM	PLIM_1
	153.748	0.91198	151.857	155.640	147.183	160.314
	183.601	0.87406	181.788	185.414	177.058	190.144
/	181.755	0.83474	180.024	183.486	175.234	188.276
	200.185	1.16029	197.779	202.591	193.453	206.917
	136.951	0.90445	135.075	138.826	130.390	143.511
	102.445	1.76442	98.786	106.104	95.171	109.719
	152.091	0.66428	150.713	153.468	145.654	158.527

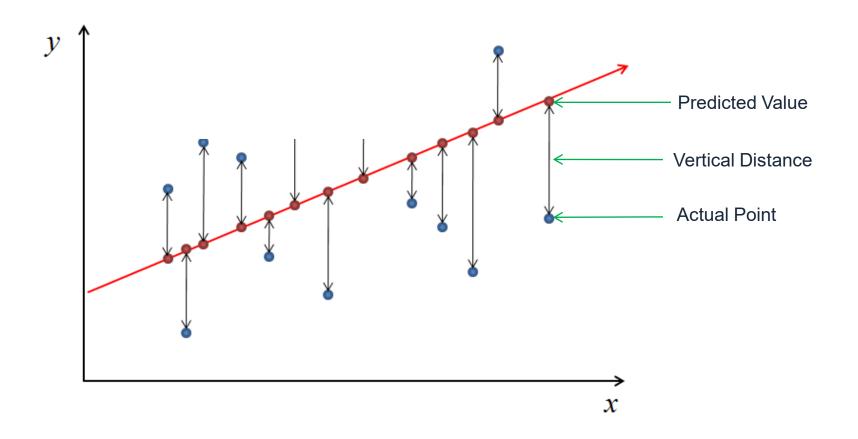


4.2.4 Residuals Analysis



Remember what Residuals Are?

• **Residuals** are the vertical difference between actual values and the predicted values or the "fitted line" created by the regression model.





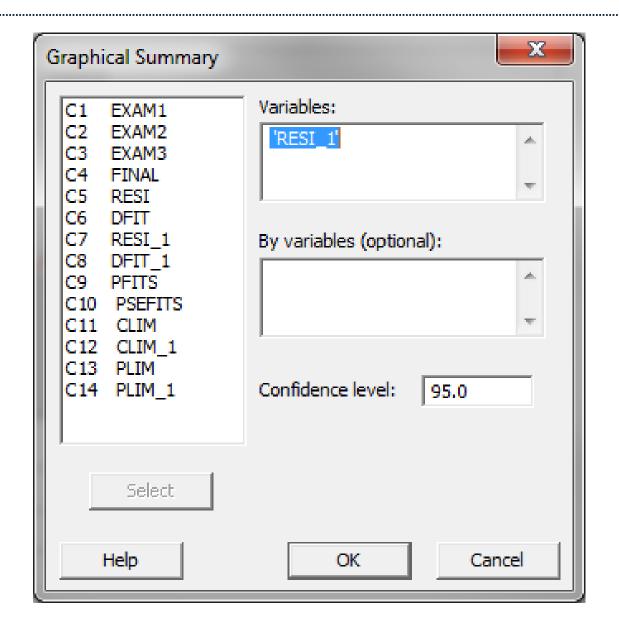
Why Perform Residuals Analysis?

- The regression equation generated based on the sample data can make accurate statistical inference only if certain assumptions are met. Residuals analysis can help to validate these assumptions. The following assumptions must be met to ensure the reliability of the linear regression model:
 - The errors are normally distributed with mean equal to zero.
 - The errors are independent.
 - The errors have a constant variance.
 - The underlying population relationship is linear.
- If the residuals performance does not meet the requirement, we will need to rebuild the model by replacing the predictors with new ones, adding new predictors, building non-linear models, and so on.



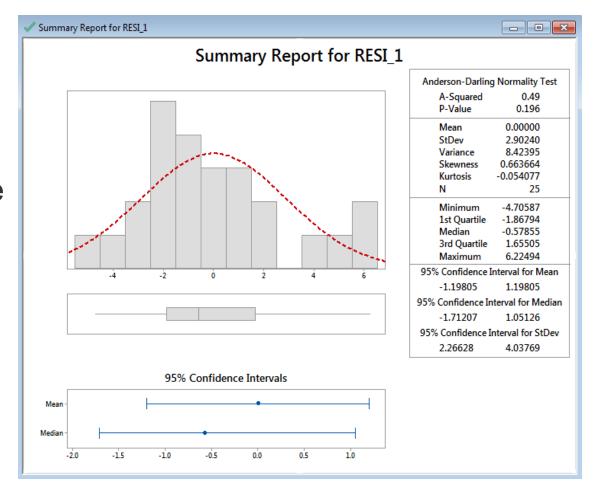
- Step 1: Check whether residuals are normally distributed around the mean of zero.
 - 1) Click Stat → Basic Statistics → Graphical Summary.
 - 2) A new window named "Graphical Summary" pops up.
 - 3) Select "RESI_1" as the "Variables." We are selecting RESI_1 because we ran our regression two times, the first with all Exams which gave us our first set of residuals (RESI), the second model we ran without Exam1 and then we got residuals added to our data table and that column was automatically labeled RESI_1.
 - 4) Click "OK."
 - 5) The histogram and analysis results are shown automatically in the new window.







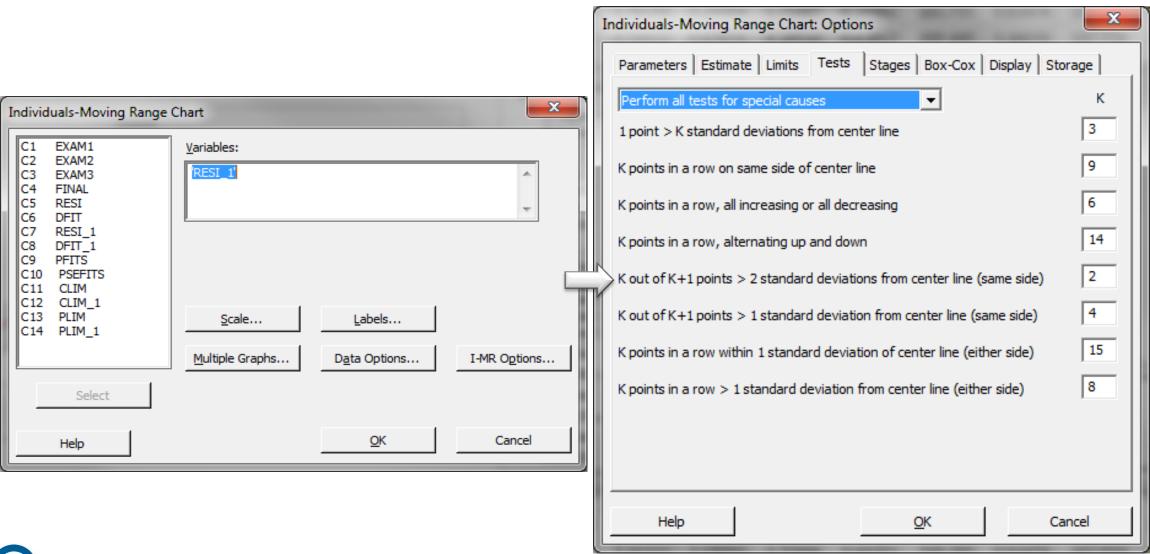
- The mean of residuals is 0.0000.
- The Anderson-Darling test is used to test the normality. Since the p-value (0.196) is greater than the alpha level (0.05), we fail to reject the null hypothesis; the residuals are normally distributed.
 - H₀: The residuals are normally distributed.
 - H₁: The residuals are not normally distributed.





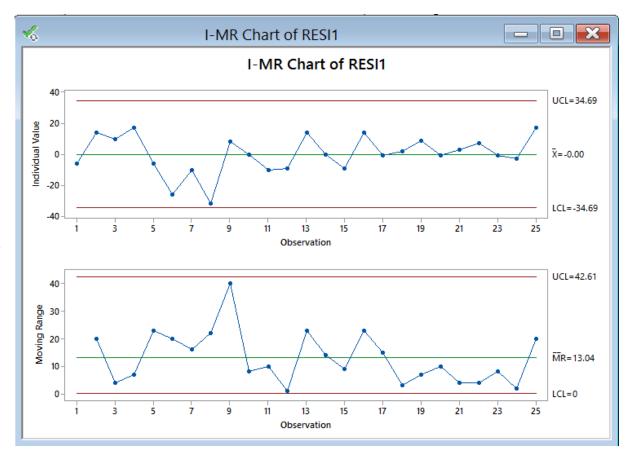
- Step 2: If the data are in time order, run the I-MR chart to check whether residuals are independent.
 - 1) Click Stat \rightarrow Control Charts \rightarrow Variable Charts for Individuals \rightarrow I-MR.
 - 2) A new window named "Individuals Moving Range Chart" pops up.
 - 3) Select "RESI1" as "Variables
 - 4) Click "I-MR Options"
 - 5) Click on "Tests" tab and select "Perform all tests for special causes"
 - 6) Click "OK." to close "I-MR" Options
 - 7) Click OK
 - 8) The control charts are shown automatically in the new window.







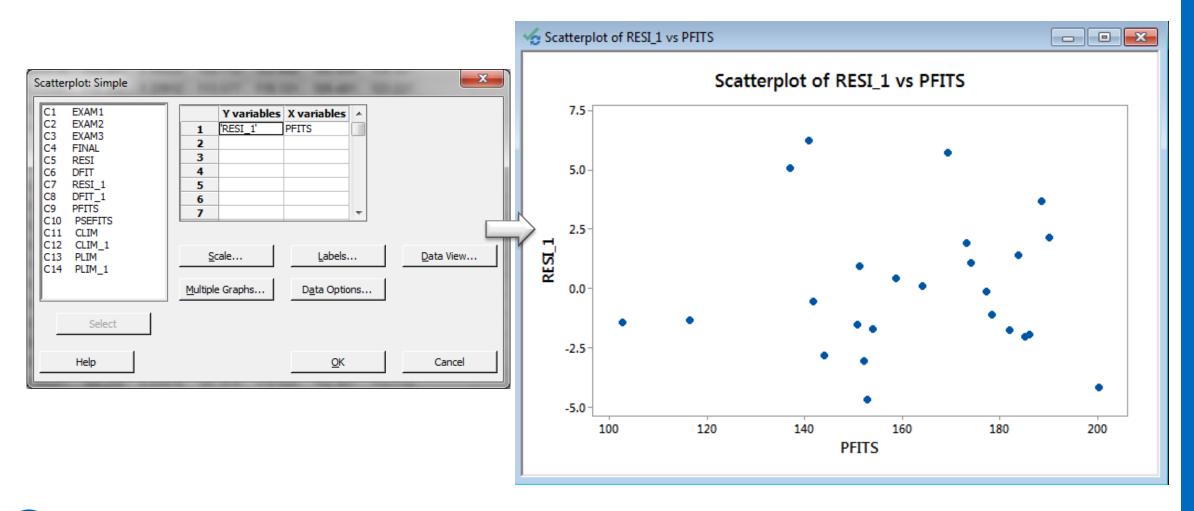
- If no data points are out of control in both the I-chart and MR chart, the residuals are independent of each other.
- If the residuals are not independent, it is possible that some important predictors are not included in the model.
- In this example, since the IMR chart is in control, residuals are independent.





- Step 3: Check whether residuals have equal variance across the predicted responses.
 - Create a scatterplot with Y being the residuals and the X being the fitted values.
 - 1. Click Graph → Scatterplot.
 - 2. A new window named "Scatterplots" appears.
 - 3. Select "RESI1" as the "Y variables" and "Fitted" as the "X variables."
 - 4. Click "OK."
 - 5. The scatterplot appears in a new window.
 - We are looking for the pattern in which residuals spread out evenly around zero from the top to the bottom.







4.2.5 Data Transformation



What is the Box-Cox Transformation?

- When a response does not fit the model well, sometimes using a transformation of the response can improve the fit.
- **Power transformation** is a class of transformation functions that raise the response to some power. For example, a square root transformation converts X to X^{1/2}.
- Box-Cox transformation is a popular power transformation method developed by George E. P. Box and David Cox.



Box-Cox Transformation Formula

The formula of the Box-Cox transformation is:

$$y = \frac{x^{\lambda} - 1}{\lambda} \quad \text{where} \quad \lambda \neq 0$$

$$y = \ln x \quad \text{where} \quad \lambda = 0$$

y is the transformation result.

x is the variable under transformation.

 λ is the transformation parameter.

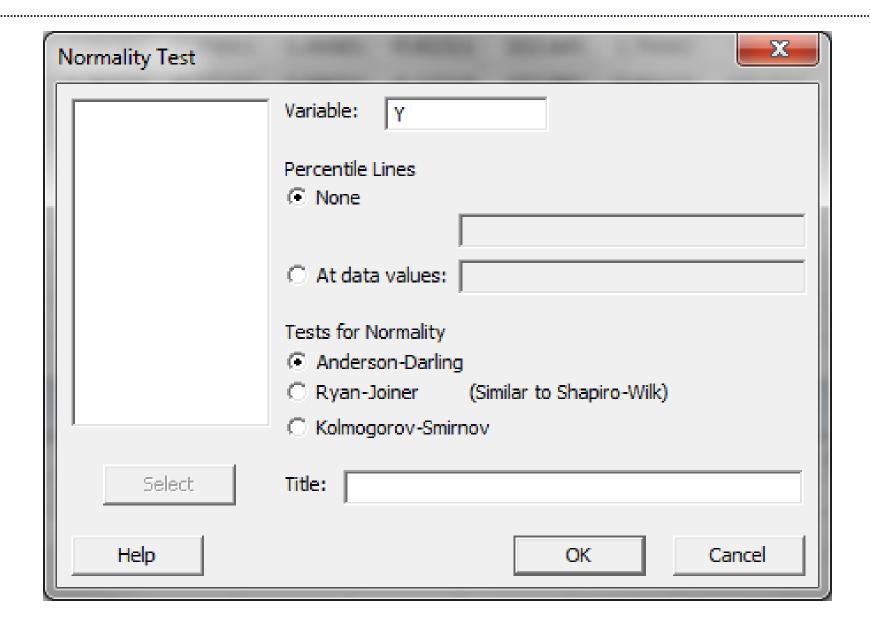


- Minitab provides the best Box-Cox transformation with an optimal λ that minimizes the model SSE (sum of squared error).
- Here is an example of how we transform the non-normally distributed response to normal data using Box-Cox method.
- Data File: "Box-Cox" tab in "Sample Data.xlsx"



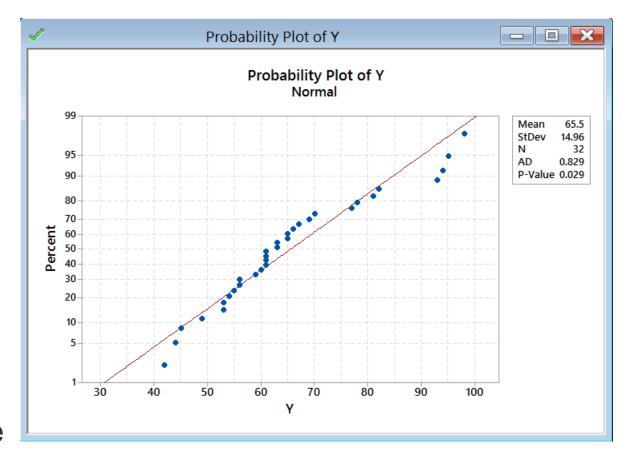
- Step 1: Test the normality of the original data set.
 - 1) Click Stat → Basic Statistics → Normality Test.
 - 2) A new window named "Normality Test" pops up.
 - 3) Select "Y" as "Variable."
 - 4) Click "OK."
 - 5) The normality test results are shown automatically in the new window.







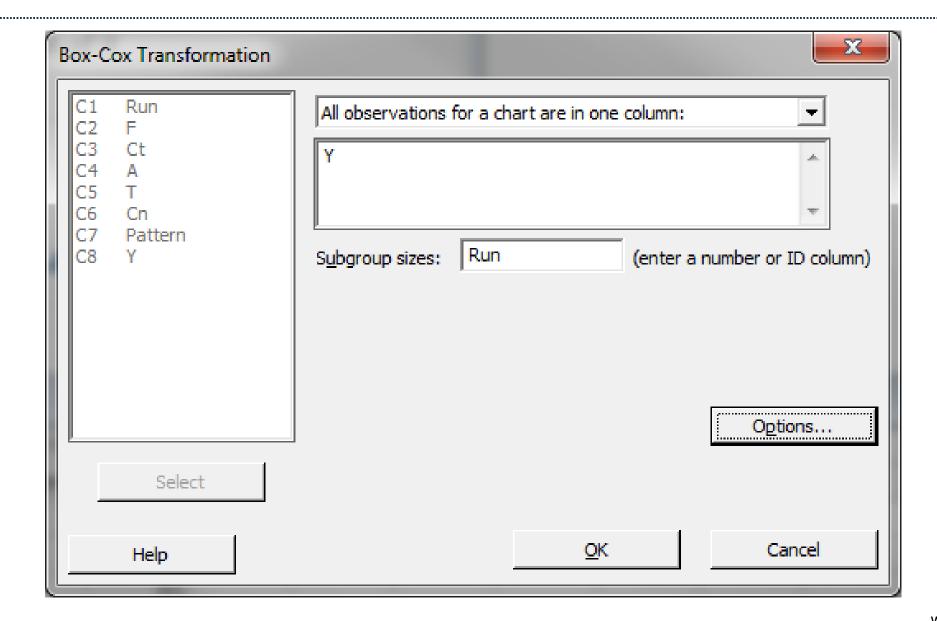
- Normality Test:
 - H0: The data are normally distributed.
 - H1: The data are not normally distributed.
- If p-value > alpha level (0.05), we fail to reject the null hypothesis. Otherwise, we reject the null.
- In this example, p-value = 0.029 < alpha level (0.05). The data are not normally distributed.





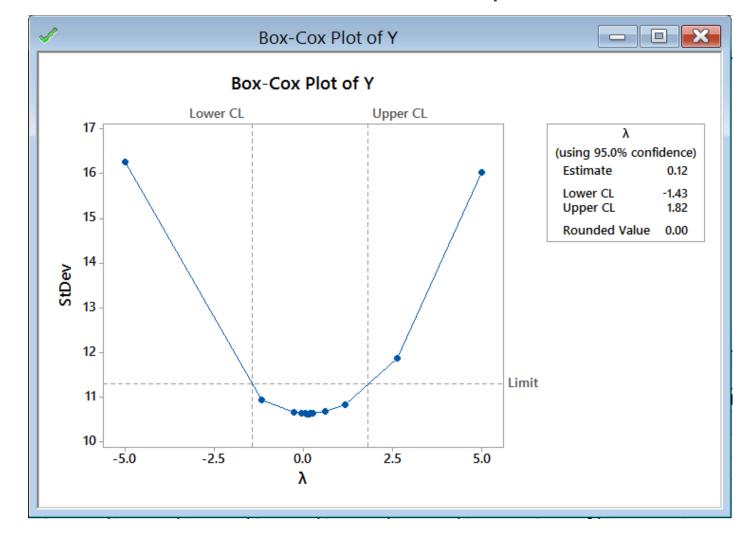
- Step 2: Run the Box-Cox Transformation:
 - 1) Click Stat → Control Charts → Box-Cox Transformation.
 - 2) A new window named "Box-Cox Transformation" pops up.
 - 3) Click into the blank list box below "All observations for a chart are in one column."
 - 4) Select "Y" as the variable.
 - 5) Select "Run" into the box next to "Subgroup sizes (enter a number or ID column)."
 - 6) Click "OK."
 - 7) The analysis results are shown automatically in the new window.







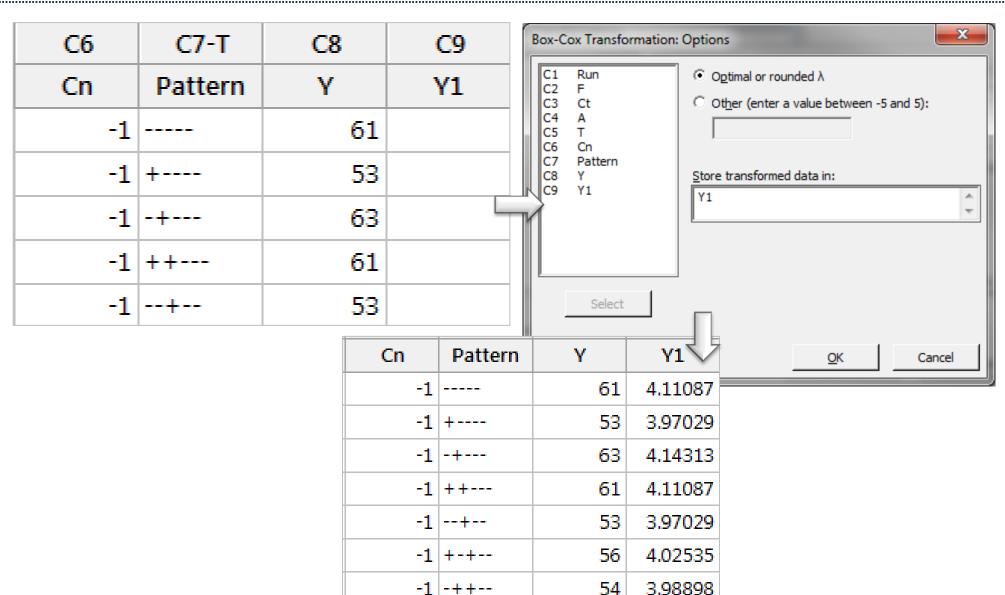
• When λ is 0.12, the transformation is the optimal with minimum SSE.





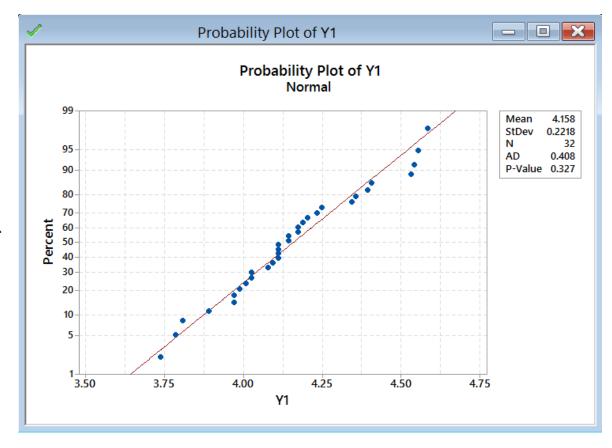
- The transformed Y can also be saved in another column.
 - 1) Create a new column named "Y1" in the data table.
 - 2) Click on the "Options" button in the "Box-Cox Transformation" window.
 - 3) A new window named "Box-Cox Transformation Options" appears.
 - 4) Click in the blank box under "Store transformed data in" and all the columns pop up in the list box on the left.
 - 5) Select "Y1" in "Store transformed data in."
 - 6) Click "OK" in the window "Box-Cox Transformation Options."
 - 7) Click "OK" in the window "Box-Cox Transformation."
 - 8) The transformed column is stored in the column "Y1."







- Run the normality test to check whether the transformed data are normally distributed.
- Use the Anderson-Darling test to test the normality of the transformed data
 - H0: The data are normally distributed.
 - H1: The data are not normally distributed.
- If p-value > alpha level (0.05), we fail to reject the null. Otherwise, we reject the null.
- In this example, p-value = 0.327 > alpha level (0.05). The data are normally distributed.





5.0 Control Phase



Black Belt Training: Control Phase

5.1 Lean Controls

- 5.1.1 Control Methods for 5S
- 5.1.2 Kanban
- 5.1.3 Poka-Yoke (Mistake Proofing)

5.2 Statistical Process Control (SPC)

- 5.2.1 Data Collection for SPC
- 5.2.2 I-MR Chart
- 5.2.3 Xbar-R Chart
- 5.2.4 U Chart
- 5.2.5 P Chart
- 5.2.6 NP Chart

- 5.2.7 X-S chart
- 5.2.8 CumSum Chart
- 5.2.9 EWMA Chart
- 5.2.10 Control Methods
- 5.2.11 Control Chart Anatomy
- 5.2.12 Subgroups, Variation, Sampling

5.3 Six Sigma Control Plans

- 5.3.1 Cost Benefit Analysis
- 5.3.2 Elements of the Control Plan
- 5.3.3 Elements of the Response Plan



5.1 Lean Controls



Black Belt Training: Control Phase

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5.1.1 Control Methods for 5S



What is 5S?

- 5S is a systematic method to organize, order, clean, and standardize a workplace...and keep it that way!
 - 5S is a methodology of organizing and improving the work environment.
- 5S is summarized in five Japanese words, all starting with the letter S:
 - Seiri (sorting)
 - Seiton (straightening)
 - Seiso (shining)
 - Seiketsu (standardizing)
 - Shisuke (sustaining).
- 5S was originally developed in Japan, and is widely used to optimize the workplace to increase productivity and efficiency.



5S Goals

- Reduced waste
- Reduced cost
- Establish a work environment that is:
 - self-explaining
 - self-ordering
 - self-regulating
 - self improving.
 - Where there is/are no more:
 - Wandering and/or searching
 - Waiting or delays
 - Secrets hiding spots for tools
 - Obstacles or detours
 - Extra pieces, parts, materials, etc.
 - Injuries
 - Waste.



5S Benefits

- Reduced changeovers
- Reduced defects
- Reduced waste
- Reduced delays
- Reduced injuries
- Reduced breakdowns
- Reduced complaints
- Reduced red ink
- Higher quality
- Lower costs
- Safer work environment
- Greater associate and equipment capacity



5S Systems Reported Results

•	Cut in floor space:	60%
---	---------------------	-----

• Cut in flow distance: 80%	%
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• Cut in accidents: 70	9/	, O
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 Cut in rack storage: 	68%
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•	Cut in	number	of forklifts:	45%
---	--------	--------	---------------	-----

•	Cut in	machine	changeover	time:	62%
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•	Cut in annual	physical	inventory time:	50%
	Out iii aiiiiaai	priyordar	mivoritory tillio.	00 / 0

•	Cut in classroom	raining training	requirements:	55%
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•	Cut in	nonconformance	e in assen	nbly:	96%
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•	Increase in	test yields:	50%)
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• Increase in throughput: 15%



Sorting (Seiri)



- Go through all tools, parts, equipment, supply, and materials in the workplace.
- Categorize into two major groups: needed and unneeded.
- Eliminate unneeded items from the workplace. Dispose of or recycle those items.
- Keep need items and sort in order of priority. When in doubt...throw it out!



Straightening (Seiton)



- Straightening in 5S is also called setting in order.
 - Label each needed item.
 - Store items at their best locations so that the workers can find them easily whenever they needed any item.
 - Reduce the motion and time required to locate and obtain any item whenever it is needed.
 - Promote an efficient work flow path.
 - Use visual aids like the tool board image on this page.



Shining (Seiso)

• Shining in 5S is also called sweeping.



- Clean the workplace thoroughly.
- Maintain the tidiness of the workplace.
- Make sure every item is located at the specific location where it should be.
- Create the ownership in the team to keep the work area clean and organized.



Standardizing (Seiketsu)

- Standardize the workstation and the layout of tools, equipment, and parts.
- Create identical workstations with a consistent way of storing the items at their specific locations so that workers can be moved around to any workstation any time and perform the same task.



Sustaining (Shisuke)

- Sustaining in 5S is also called self-discipline.
- Create the culture in the team to follow the first four S's consistently.
- Avoid falling back to the old ways of cluttered and unorganized work environment.
- Keep the momentum of optimizing the workplace.
- Promote innovations of workplace improvement.
- Sustain the first fours S's using:
 - 5S Maps
 - 5S Schedules
 - 5S Job cycle charts
 - Integration of regular work duties
 - 5S Blitz schedules
 - Daily workplace scans.



Simplified Summary of 5S

- Sort "when in doubt, move it out."
- Set in Order Organize all necessary tools, parts, and components of production. Use visual ordering techniques wherever possible.
- Shine Clean machines and/or work areas. Set regular cleaning schedules and responsibilities.
- **Standardize** Solidify previous three steps, make 5S a regular part of the work environment and everyday life.
- Sustain Audit, manage, and comply with established five-s guidelines for your business or facility



5.1.2 Kanban



What is Kanban?

- The Japanese word "Kanban" means "signboard."
- Kanban system is a "pull" production scheduling system to determine when to produce, what to produce, and how much to produce based on the demand.
- It was originally developed by Taiichi Ohno in order to reduce the waste in inventory and increase the speed of responding to the immediate demand.



Kanban System

- Kanban system is a demand-driven system.
- The customer demand is the signal to trigger or pull the production.
- Products are made only to meet the immediate demand. When there is no demand, there is no production.
- It is designed to minimize the in-process inventory and to have the right material with the right amount at the right location at the right time.



Kanban System

- Principles of the Kanban System:
 - Only produce products with exactly the same amount that customers consume.
 - Only produce products when customers consume.
- The production is driven by the actual demand from the customer side instead of the forecasted demand planned by the staff.



Kanban Card

- The Kanban card is the ticket or signal to authorize the production or movement of materials. It is the message of asking for more.
- It is sent from the end customer up to the chain of production.
- Upon receiving of a Kanban card, the production station would start to produce goods.
- The Kanban card can be a physical card or an electronic signal.



Kanban System Example

- The simplest example of a Kanban system is the supermarket operation.
- Customers visit the supermarkets and buy what they need.
- The checkout scanners send electronic Kanban cards to the local warehouse asking for more when the items are sold to customers.
- When the warehouse receives the Kanban cards, it starts to replenish the exact goods being sold.
- It the warehouse prepares more than what Kanban cards require, the goods would become obsolete. If it prepares less, the supermarket would not have the goods available when customers need them.



Kanban System Benefits

- Minimize in-process inventory
- Free up space occupied by unnecessary inventory
- Prevent overproduction
- Improve responsiveness to dynamic demand
- Avoid the risk of inaccurate demand forecast
- Streamline the production flow
- Visualize the work flow.



5.1.3 Poka-Yoke



What is Poka-Yoke?

- The Japanese term "poka-yoke" means "mistake-proofing."
- It is a mechanism to eliminate defects as early as possible in the process.
- It was originally developed by Shigeo Shingo and was initially called "baka-yoke" (fool-proofing).



Two Types of Poka-Yoke

Prevention

- Preventing defects from occurring
- Removing the possibility that an error could occur
- Making the occurrence of an error impossible.

Detection

- Detecting defects once they occur
- Highlighting defects to draw workers' attention immediately
- Correcting defects so that they would not reach the next stage.



Three Methods of Poka-Yoke

- Contact Method
 - Use of shape, color, size, or any other physical attributes of the items.
- Constant Number Method
 - Use of a fixed number to make sure a certain number of motions are completed.
- Sequence Method
 - Use of a checklist to make sure all the prescribed process steps are followed in the right order.



Poka-Yoke Devices

- We are surrounded by poka-yoke devices daily.
 - Prevention Devices
 - Example: the dishwasher does not start to run when the door is open.
 - Detection Devices
 - Example: the car starts to beep when the passengers do not buckle their seatbelts.
- Poka-yoke devices can be in any format that can quickly and effectively prevent or detect mistakes.
 - Visual, electrical, mechanical, procedural, human etc.



Steps to Apply Poka-Yoke

- Step 1: Identify the process steps in need of mistake proofing.
- Step 2: Use the 5-why's method to analyze the possible mistakes or failures for the process step.
- Step 3: Determine the type of poka-yoke: prevention or detection.
- Step 4: Determine the method of poka-yoke: contact, constant number, or sequence.
- Step 5: Pilot the poka-yoke approach and make any adjustments if needed.
- Step 6: Implement poka-yoke in the operating process and maintain the performance.



5.2 Statistical Process Control



Black Belt Training: Control Phase

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5.2.1 Data Collection for SPC



What is SPC?

- Statistical process control (SPC) is a statistical method to monitor the performance of a process using control charts in order to keep the process in statistical control.
- Statistical process control can be used to distinguish between special cause variation and common cause variation in the process.
- It presents the voice of the process.



Common Cause Variation

- Common cause variation (also called chance variation) is the inherent natural variation in any processes.
- It is the random background noise, which cannot be controlled or eliminated from the process.
- Its presence in the process is expected and acceptable due to its relatively small influence on the process.



Special Cause Variation

- Special cause variation (also called assignable cause variation) is the unnatural variation in the process.
- It is the cause of process instability and leads to defects of the products or services.
- It is the signal of unanticipated change (either positive or negative) in the process.
- It is possible to eliminate the special cause variation from the process.



Process Stability

- A process is stable when:
 - There is not any special cause variation involved in the process
 - The process is in statistical control
 - The future performance of the process is predictable within certain limits
 - The changes happening in the process are all due to random inherent variation
 - There are not any trends, unnatural patterns, and outliers in the control chart of the process.



SPC Benefits

- Statistical process control can be used in different phases of Six Sigma projects to:
 - Understand the stability of a process
 - Detect the special cause variation in the process
 - Identify the statistical difference between two phases
 - Eliminate or apply the unnatural change in the process
 - Improve the quality and productivity.

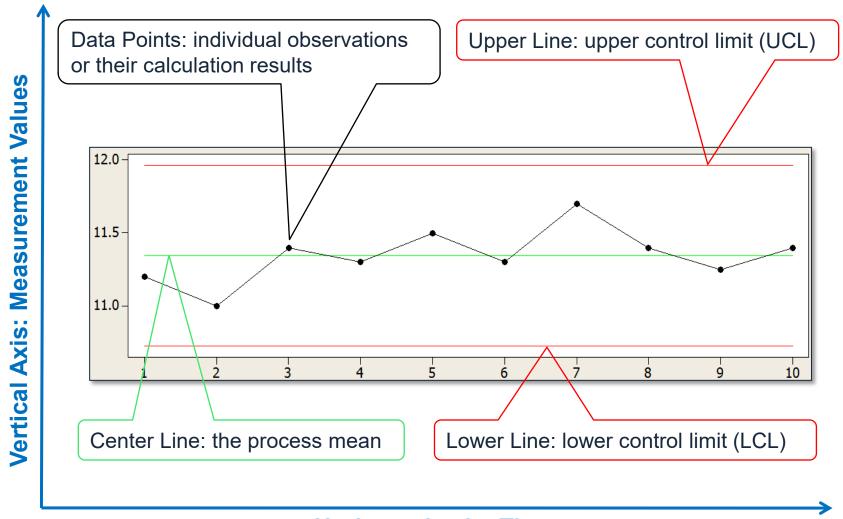


Control Charts

- Control charts are graphical tools to present and analyze the process performance in statistical process control.
- Control charts are used to detect special cause variation and determine whether the process is in statistical control (stable).
- Variation solutions:
 - Minimize the common cause variation
 - Eliminate the special cause variation when it leads to unanticipated negative changes in the outcome
 - Implement the special cause variation when it leads to unanticipated positive changes in the outcome.



Control Charts Elements





Control Charts Elements

- Control charts can work for both continuous data and discrete or count data.
- Control limits are approximately three sigma away from the process mean.
- A process is in statistical control when all the data points on the control charts fall within the control limits and have random patterns only.
- Otherwise, the process is out of control and we need to investigate the special cause variation in the process.



Possible Errors in SPC

There are two types of possible errors in interpreting controls charts.

		Interpretation	
		Common Cause	Special Cause Variation
Truth	Common Cause Variation		Type I Error (False Positive)
	Special Cause Variation	Type II Error (False Negative)	



Possible Errors in SPC

- It is similar to the way of defining the type I and type II errors in hypothesis testing.
- Control charts can be interpreted as a way of testing the hypothesis about the process stability.
 - Null Hypothesis (H₀): The process is stable (i.e., in statistical control).
 - Alternative Hypothesis (H_A): The process is unstable (i.e., out of statistical control).



Possible Errors in SPC

- Type I Error
 - False positive
 - False alarm
 - Considering true common cause variation as special cause variation
 - Type I errors waste resources spent on investigation.
- Type II Error
 - False negative
 - Miss
 - Considering true special cause variation as common cause variation
 - Type II errors neglect the need to investigate critical changes in the process.



Data Collection Considerations

- To collect data for plotting control charts, we need to consider:
 - What is the measurement of interest?
 - Are the data discrete or continuous?
 - How many samples do we need?
 - How often do we sample?
 - Where do we sample?
 - What is the sampling strategy?
 - Do we use the raw data collected or transfer them to percentages, proportions, rates, etc.?



Subgroups and Rational Subgrouping

- When sampling, we randomly select a group of items (i.e., a subgroup) from the population of interest.
- The subgroup size is the count of samples in a subgroup. It can be constant or variable.
- Depending on the subgroup sizes, we select different control charts accordingly.
- Rational subgrouping is the basic sampling scheme in SPC.
- The goal of rational subgrouping is to maximize the likelihood of detecting special cause variation. In other words, the control limits should only reflect the variation between subgroups.
- The number of subgroups, subgroup size, and frequency of sampling have great impact on the quality of control charts.



Impact of Variation

- The rational subgrouping strategy is designed to minimize the opportunity of having special cause variation *within* subgroups.
- If there is only random variation (background noise) within subgroups, all the special cause variation would be reflected between subgroups. It is easier to detect an out-of-control situation.
- Random variation is inherent and indelible in the process. We are more interested in identifying and taking actions on special cause variation.



Frequency of Sampling

- The frequency of sampling in SPC depends on whether we have sufficient data to signal the changes in a process with reasonable time and costs.
- The more frequently we sample, the higher costs it may trigger.
- We need the subject matter experts' knowledge on the nature and characteristics of the process to make good decisions on sampling frequency.



5.2.2 I-MR Chart



I-MR Chart

- The **I-MR chart** (also called individual-moving range chart or I-MR chart) is a popular control chart for continuous data with subgroup size equal to one.
- The I chart plots an individual observation as a data point.
- The MR chart plots the absolute value of the difference between two consecutive observations in individual charts as a data point.
- If there are n data points in the I chart, there are n-1 data points in and MR chart with a moving range of 2.
- The I chart is valid only if the MR chart is in control.
- The underlying distribution of the I-MR chart is normal distribution.



I Chart Equations

I Chart (Individuals Chart)

• Data Point: X_i

• Center Line: $\sum_{i=1}^{n} x_i$

• Control Limits: $\frac{\sum_{i=1}^{n} x_i}{n} \pm 2.66 \times \overline{MR}$

where *n* is the number of observations



MR-Chart Equations

MR Chart (Moving Range Chart)

• Data Point:
$$\left| x_{i+1} - x_i \right|$$

• Center Line: $\frac{\sum |x_{i+1} - x_i|}{n-1}$

• Upper Control Limit:
$$3.267 \times \frac{\sum |x_{i+1} - x_i|}{n-1}$$

Lower Control Limit: 0
 where n is the number of observations

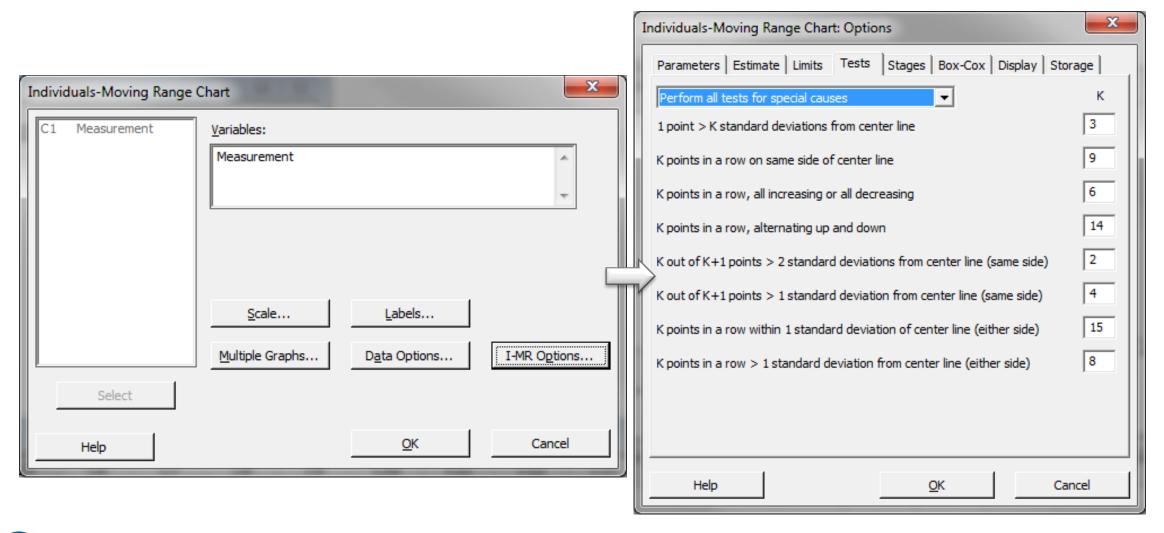


Use Minitab to Plot I-MR Charts

- Data File: "IR" tab in "Sample Data.xlsx"
- Steps to plot I-MR charts in Minitab
 - 1) Click Stat \rightarrow Control Charts \rightarrow Variable Charts for Individuals \rightarrow I-MR.
 - 2) A new window named "Individuals Moving Range Chart" appears.
 - 3) Select "Measurement" as the "Variables."
 - 4) Click "I-MR Options" button and a new window "Individual Moving Range Chart Options" appears.
 - 5) Click on the tab "Tests."
 - 6) Select the item "Perform all tests for special causes" in the dropdown box.
 - 7) Click "OK" in the window "Individuals Moving Range Chart Options."
 - 8) Click "OK."
 - 9) The I-MR charts appear in the newly-generated window.



Use Minitab to Plot I-MR Charts





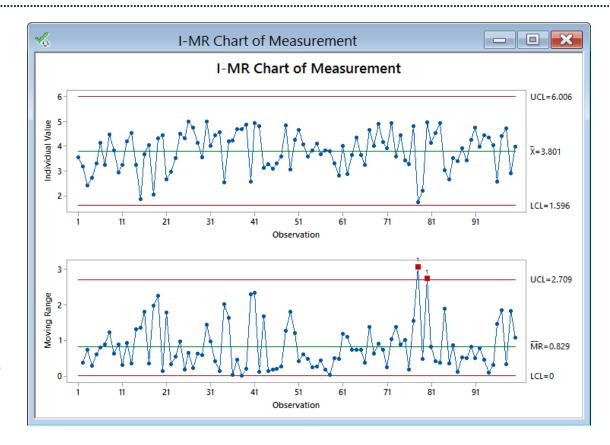
I-MR Charts Diagnosis

I Chart (Individuals' Chart):

 Since the MR chart is out of control, the I chart is invalid.

MR Chart (Moving Range Chart):

- Two data points fall beyond the upper control limit. This indicates the MR chart is out of control (i.e., the variations between every two contiguous individual samples are not stable over time).
- We need to further investigate the process, identify the root causes that trigger the outliers, and correct them to bring the process back in control.





5.2.3 Xbar-R Chart



Xbar-R Chart

- The **Xbar-R** chart is a control chart for continuous data with a constant subgroup size between two and ten.
- The Xbar chart plots the average of a subgroup as a data point.
- The R chart plots the difference between the highest and lowest values within a subgroup as a data point.
- The Xbar chart monitors the process mean and the R chart monitors the variation within subgroups.
- The Xbar is valid only if the R chart is in control.
- The underlying distribution of the Xbar-R chart is normal distribution.



Xbar Chart Equations

Xbar chart

• Data Point:
$$\overline{X}_i = \frac{\sum_{j=1}^{m} x_{ij}}{m}$$

• Center Line:
$$\overline{\overline{X}} = \frac{\sum_{i=1}^{k} \overline{X}_i}{k}$$

• Control Limits $\overline{\overline{X}} \pm A_2 \overline{R}$



where m is the subgroup size and k is the number of subgroups. A_2 is a constant depending on the subgroup size.

R Chart Equations

R chart (Rage Chart)

• Data Point: $R_i = \underset{j \in [1,m]}{Max}(x_{ij}) - \underset{j \in [1,m]}{Min}(x_{ij})$

- Center Line: $\overline{R} = \frac{\sum_{i=1}^{k} R_i}{k}$
- Upper Control Limit: $D_4 \times \overline{R}$
- Lower Control Limit: $D_3 \times \overline{R}$

where m is the subgroup size and k is the number of subgroups. D_3 and D_4 are constants depending on the subgroup size.

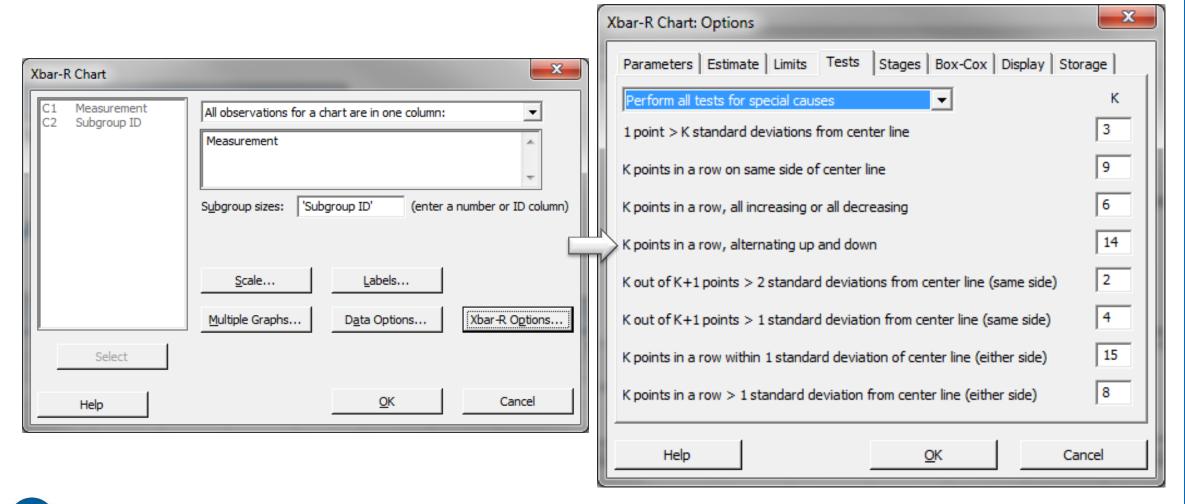


Use Minitab to Plot Xbar-R Charts

- Data File: "Xbar-R" tab in "Sample Data.xlsx"
- Steps to plot Xbar-R charts in Minitab
 - Click Stat → Control Charts → Variable Charts for Subgroups → Xbar-R.
 - A new window named "Xbar-R Chart" appears.
 - Click in the blank box right below "All observations for a chart are in one column" and the variables appear in the list box on the left.
 - Select the "Measurement" into the box below "All observations for a chart are in one column."
 - Select the "Subgroup ID" as the "Subgroup size (enter a number or ID column).
 - Click "Xbar-R Options" button and a new window "Xbar-R Chart Options" appears.
 - Click on the tab "Tests."
 - Select the item "Perform all tests for special causes" in the dropdown menu.
 - Click "OK" in the window "Xbar-R Chart Options."
 - Click "OK."
 - The Xbar-R charts appear in the newly-generated window.



Use Minitab to Plot Xbar-R Charts

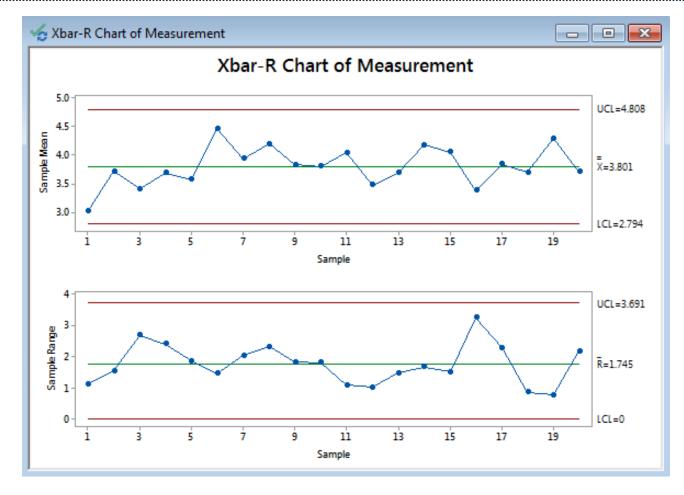




Xbar-R Charts Diagnosis

Xbar-R Chart Interpretation:

- Since the R chart is in control, the Xbar chart is valid.
- In both charts, there are not any data points failing any tests for special causes (i.e., all the data points fall between the control limits and spread around the center line with a random pattern).
- We conclude that the process is in control.





5.2.4 U Chart



Defect vs. Defective

- A **defect** of a unit is the unit's characteristic that does not meet the customers' requirements.
- A defective is a unit that is not acceptable to the customers.
- One defective might have multiple defects.
- One unit might have multiple defects but be still usable to the customers.



U Chart

- The U chart is a control chart monitoring the average defects per unit.
- The U chart plots the count of defects per unit of a subgroup as a data point.
- It considers the situation when the subgroup size of inspected units for which the defects would be counted is not constant.
- The underlying distribution of the U chart is Poisson distribution.



U Chart Equations

U chart

• Data Point: $u_i = \frac{x_i}{n_i}$

• Center Line: $u = \frac{\sum_{i=1}^{k} u_i}{k}$

• Control Limits: $u \pm 3 \times \sqrt{\frac{u}{n_i}}$

where n_i is the subgroup size for the ith subgroup; k is the number of subgroups; x_i is the number of defects in the ith subgroup.

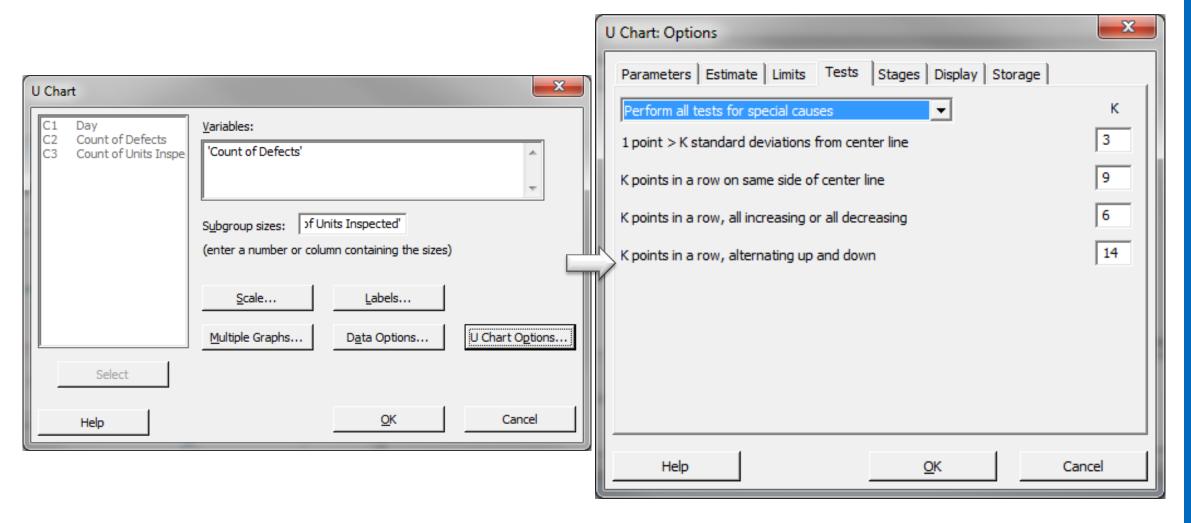


Use Minitab to Plot a U Chart

- Data File: "U" tab in "Sample Data.xlsx"
- Steps to plot a U chart in Minitab
 - 1) Click Stat → Control Charts → Attributes Charts → U.
 - 2) A new window named "U Chart" appears.
 - 3) Select "Count of Defects" as the "Variables."
 - 4) Select "Count of Units Inspected" as the "Subgroup Sizes."
 - 5) Click the button "U Chart Options" and another window named "U Chart Options" pops up.
 - 6) Click the tab "Tests."
 - 7) Select the item "Perform all tests for special causes" in the dropdown menu.
 - 8) Click "OK" in the window "U Chart Options."
 - 9) Click "OK" in the window "U Chart."
 - 10) The U chart appears in the newly-generated window.



Use Minitab to Plot a U Chart

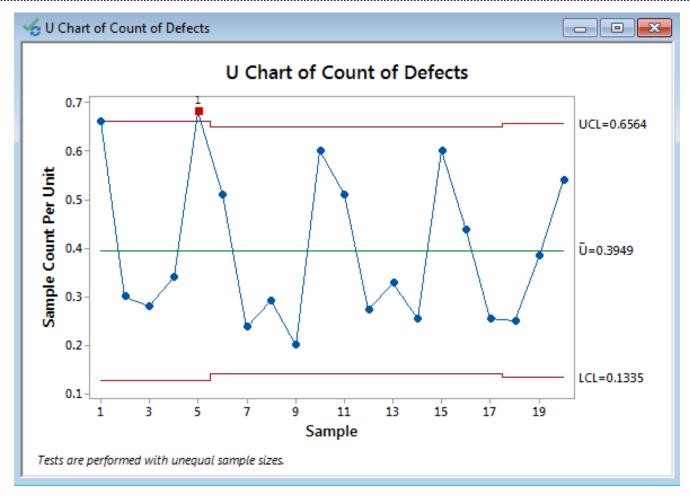




Use Minitab to Plot a U Chart

U Chart Interpretation:

- Since the sample sizes are not constant over time, the control limits are adjusted to different values accordingly.
- The data point circled in red falls beyond the upper control limit. We conclude that the process is out of control.
- Further investigation is needed to determine the special causes that triggered the unnatural pattern of the process.





5.2.5 P Chart



P Chart

- The **P** chart is a control chart monitoring the percentages of defectives.
- The P chart plots the percentage of defectives in one subgroup as a data point.
- It considers the situation when the subgroup size of inspected units is not constant.
- The underlying distribution of the P chart is binomial distribution.



P Chart Equations

P chart

• Data Point:
$$p_i = \frac{x_i}{n_i}$$

• Center Line: $p = \frac{\sum_{i=1}^{k} x_i}{\sum_{i=1}^{k} n_i}$

• Control Limits: $\overline{p} \pm 3\sqrt{\frac{\overline{p}(1-\overline{p})}{n_i}}$

where n_i is the subgroup size for the ith subgroup; k is the number of subgroups; x_i is the number of defectives in the ith subgroup.

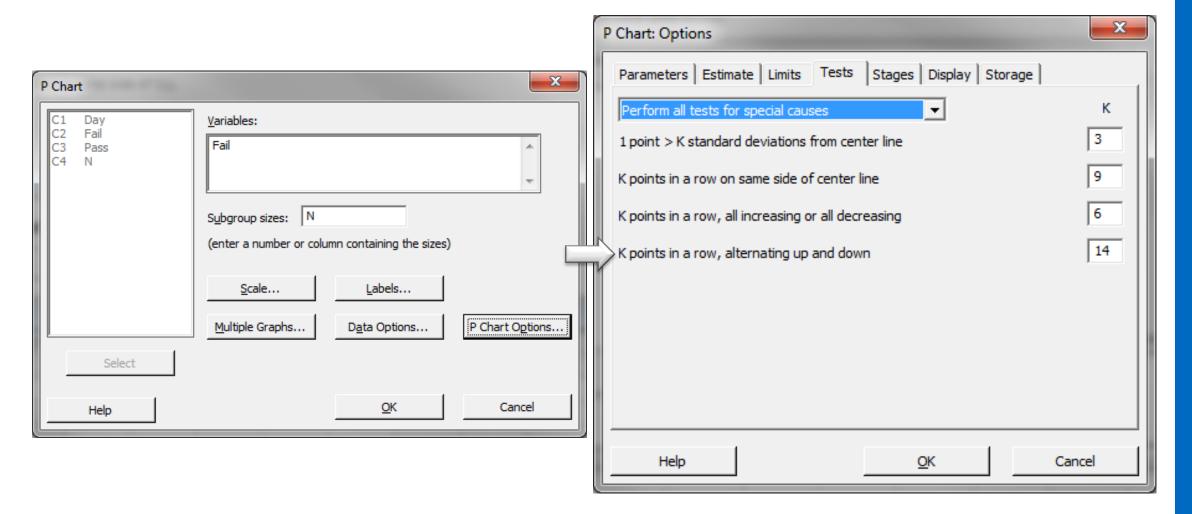


Use Minitab to Plot a P Chart

- Data File: "P" tab in "Sample Data.xlsx"
- Steps to plot a P chart in Minitab
 - 1) Click Stat \rightarrow Control Charts \rightarrow Attributes Charts \rightarrow P.
 - 2) A new window named "P Chart" appears.
 - 3) Select "Fail" as the "Variables."
 - 4) Select "N" as the "Subgroup Sizes."
 - 5) Click the button "P Chart Options" and another window named "P Chart Options" pops up.
 - 6) Click the tab "Tests."
 - 7) Select the item "Perform all tests for special causes" in the dropdown menu.
 - 8) Click "OK" in the window "P Chart Options."
 - 9) Click "OK."
 - 10) The P chart appears in the newly-generated window.



Use Minitab to Plot a P Chart

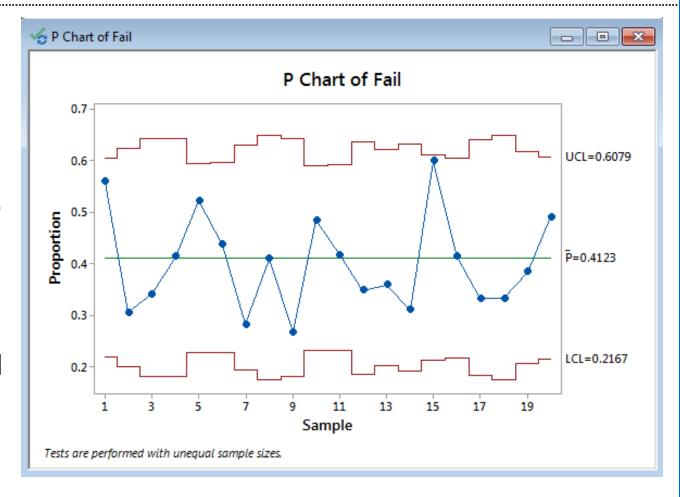




P Chart Diagnosis

• P Chart Interpretation:

- Since the sample sizes are not constant over time, the control limits are adjusted to different values accordingly.
- All the data points fall within the control limits and spread randomly around the mean.
 We conclude that the process is in control.





5.2.6 NP Chart



NP Chart

- The NP chart is a control chart monitoring the count of defectives.
- The NP chart plots the number of defectives in one subgroup as a data point.
- The subgroup size of the NP chart is constant.
- The underlying distribution of the NP chart is binomial distribution.



NP Chart Equations

- NP chart
 - Data Point X_i

• Center Line: $np = \frac{\sum_{i=1}^{k} x_i}{k}$

• Control Limits: $n\overline{p} \pm 3 \times \sqrt{n\overline{p}(1-\overline{p})}$

where n is the constant subgroup size; k is the number of subgroups; x_i is the number of defectives in the ith subgroup.

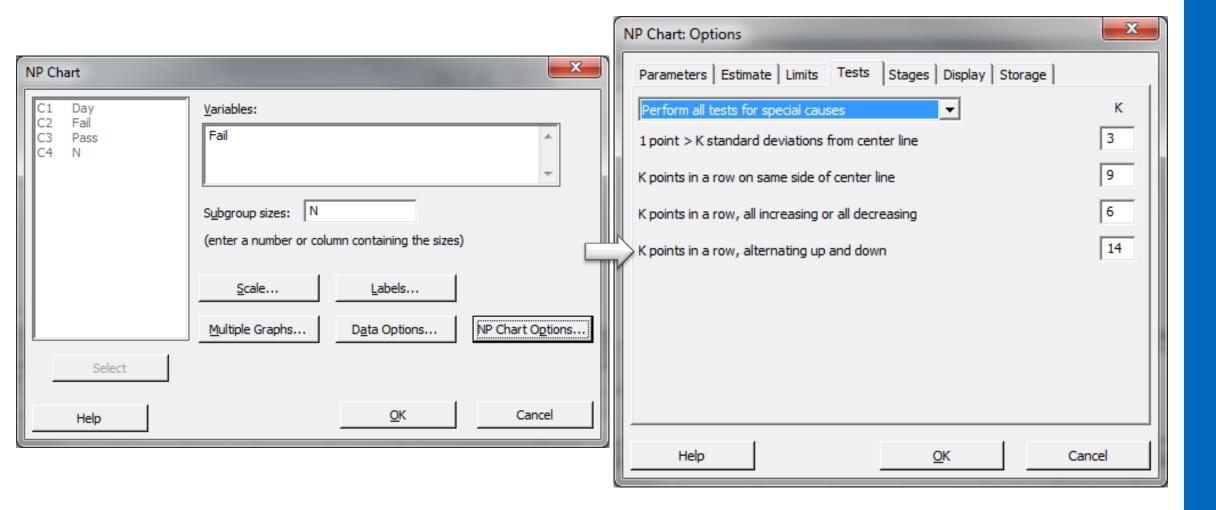


Use Minitab to Plot an NP Chart

- Data File: "NP" tab in "Sample Data.xlsx"
- Steps to plot a NP chart in Minitab
 - 1. Click Stat \rightarrow Control Charts \rightarrow Attributes Charts \rightarrow NP.
 - 2. A new window named "NP Chart" appears.
 - 3. Select "Fail" as the "Variables."
 - 4. Enter "50" as the "Subgroup Sizes."
 - 5. Click the button "NP Chart Options" and another window named "NP Chart Options" pops up.
 - 6. Click the tab "Tests."
 - 7. Select the item "Perform all tests for special causes" in the dropdown menu.
 - 8. Click "OK" in the window "NP Chart Options."
 - 9. Click "OK."
 - 10. The NP chart appears in the newly-generated window.



Use Minitab to Plot an NP Chart

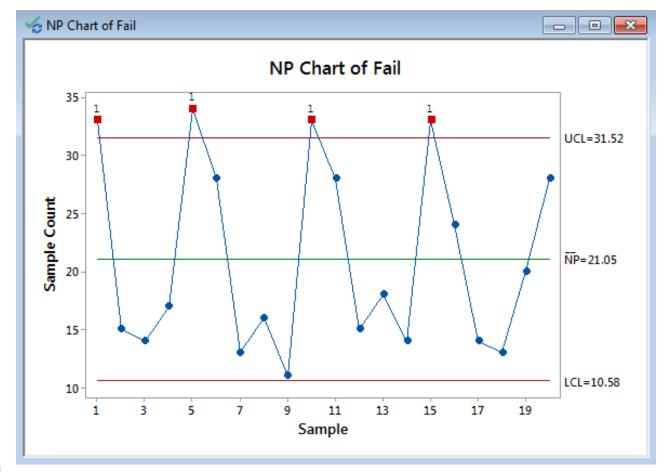




NP Chart Diagnosis

NP Chart Interpretation:

- Four data points, circled in red, fall beyond the upper control limit.
- We conclude that the NP chart is out of control.
- Further investigation is needed to determine the special causes that triggered the unnatural pattern of the process.





5.2.7 X-S Chart



X-S Chart

- The **X-S chart** (also called Xbar-S chart) is a control chart for continuous data with a constant subgroup size greater than ten.
- The Xbar chart plots the average of a subgroup as a data point.
- The S chart plots the standard deviation within a subgroup as a data point.
- The Xbar chart monitors the process mean and the S chart monitors the variability within subgroups.
- The Xbar is valid only if the S chart is in control.
- The underlying distribution of the Xbar-S chart is normal distribution.



Xbar Chart Equations

Xbar Chart

• Data Point:
$$\overline{X}_i = \frac{\sum_{j=1}^m x_{ij}}{m}$$

• Center Line: $\overline{\overline{X}} = \frac{\sum_{i=1}^{k} \overline{X}_{i}}{k}$

• Control Limits: $\overline{\overline{X}} \pm A_3 \overline{s}$

where m is the subgroup size and k is the number of subgroups. A₃ is a constant depending on the subgroup size.



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S Chart Equations

S chart

• Data Point:
$$s_i = \sqrt{\frac{\sum_{j=1}^{m} (x_{ij} - \bar{x}_i)^2}{m-1}}$$

- Center Line: $\overline{s} = \frac{\overline{i=1}}{i=1}$
- Upper Control Limit: $B_{\Delta} \times S$
- Lower Control Limit: $B_3 \times S$

where *m* is the subgroup size and *k* is the number of subgroups. B₃ and B₄ are constants depending on the subgroup size.

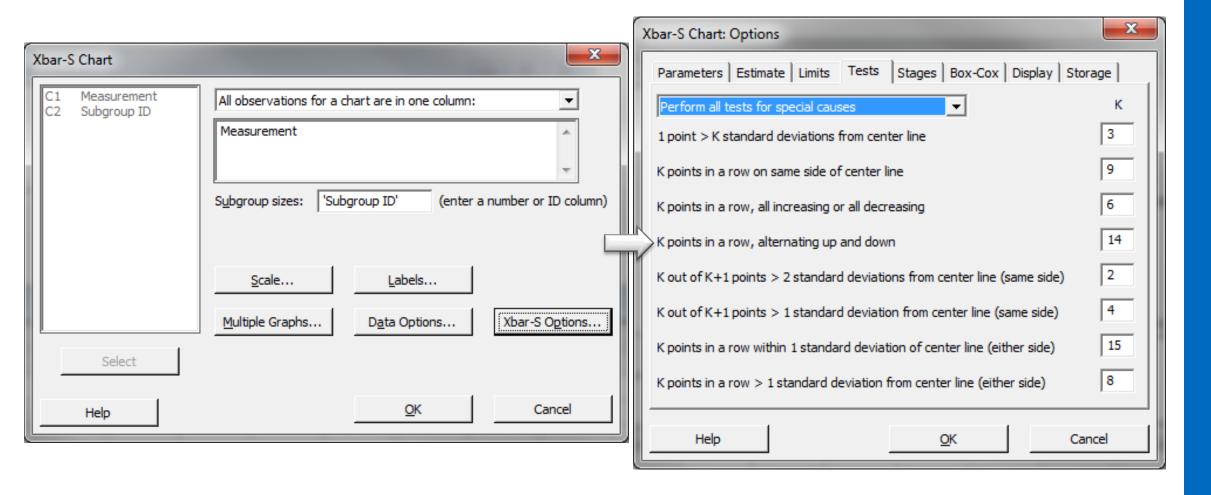


Use Minitab to Plot Xbar-S Charts

- Data File: "Xbar-S" tab in "Sample Data.xlsx"
- Steps to plot Xbar-S charts in Minitab
 - 1) Click Stat → Control Charts → Variables Charts for Subgroups → Xbar-S.
 - 2) A new window named "Xbar-S Chart" appears.
 - 3) Select "Measurement" by ticking the box below "All observations for a chart are in one column."
 - 4) Select "Subgroup ID" as the "Subgroup Sizes."
 - 5) Click the button "Xbar-S Options" and another window named "Xbar-S Chart Options" pops up.
 - 6) Click the tab "Tests."
 - 7) Select the item "Perform all tests for special causes" in the dropdown menu.
 - 8) Click "OK" in the window "Xbar-S Chart Options."
 - 9) Click "OK."
 - 10) The Xbar-S charts appear in the newly-generated window.



Use Minitab to Plot Xbar-S Charts





Use Minitab to Plot Xbar-S Charts





Xbar-S Charts Diagnosis

- Xbar-S Chart Interpretation:
 - Since the S chart is in control, the Xbar chart is valid.
 - In both charts, there are not any data points failing any tests for special cause (i.e., all the data points fall between the control limits and spread around the center line with a random pattern).

 We conclude that the process is in control.





5.2.8 CumSum Chart



CumSum Chart

- The CumSum chart (also called cumulative sum control chart or CUSUM chart) is a control chart of monitoring the cumulative sum of the subgroup mean deviations from the process target.
- It detects the shift of the process mean from the process target over time.
- The underlying distribution of the CumSum chart is normal distribution.
- There are two types of CumSum charts:
 - One two-sided CumSum charts
 - Two one-sided CumSum charts.

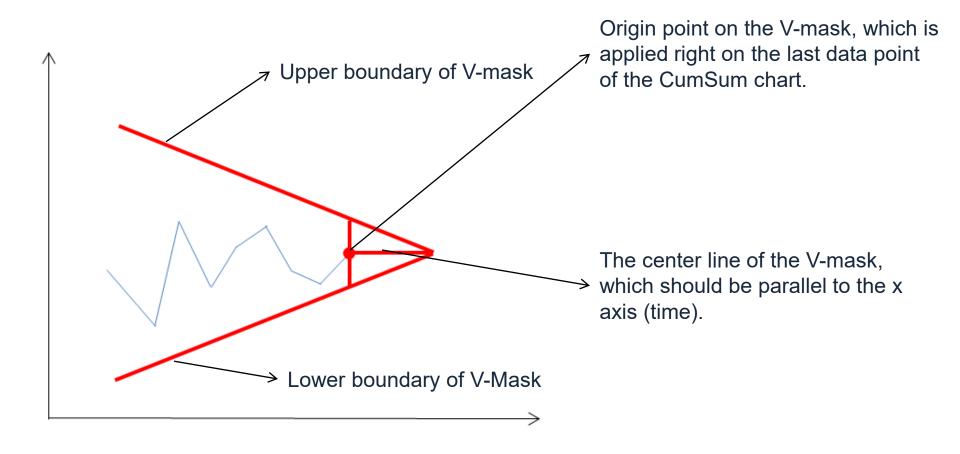


Two-Sided CumSum

- In two-sided CumSum charts, we use a V-mask to identify out-of-control data points.
- The V-mask is a transparent overlay shape of a horizontal "V" applied on the top of the CumSum chart. Its origin point is placed on the last data point of the CumSum chart and its center line is horizontal.
- If all of the data points stay inside the V-mask, we consider the process is in statistical control.



V-Mask



If any data points in the CumSum chart to the left of the origin point are outside the V-mask, the process is considered out of statistical control.



Two-Sided CumSum Equations

Two-Sided CumSum

• Data Point:
$$\begin{cases} c_i = c_{i-1} + (\overline{x}_i - T) & i > 0 \\ c_i = 0 & i = 0 \end{cases}$$

• V-Mask Slope: $k \frac{\sigma}{\sqrt{m}}$

• V-Mask Width at the Origin Point: $2h\frac{\sigma}{\sqrt{m}}$

where x_i is the mean of the ith subgroup; T is the process target; σ is the estimation of process standard deviation; m is the subgroup size.



One-Sided CumSum

- We can also use two one-sided CumSum charts to detect the shift of the process mean from the process target.
- The upper one-sided CumSum detects the upward shifts of the process mean.
- The lower one-sided CumSum detects the downward shifts of the process mean.



One-Sided CumSum Equations

One-Sided CumSum

• Data Point:
$$\begin{cases} c_i = c_{i-1} + (\overline{x}_i - T) & i > 0 \\ c_i = 0 & i = 0 \end{cases}$$

Center Line: T'

• Upper Control Limit: $c_i^+ = \max(0, x_i^- - (T+k) + c_{i-1}^+)$

• Lower Control Limit: $c_i^- = \max(0, (T-k) - x_i^- + c_{i-1}^-)$

where $\bar{\chi}_i$ is the mean of the ith subgroup; *T* is the process target; k is the slope of the lower boundary of the V-mask.

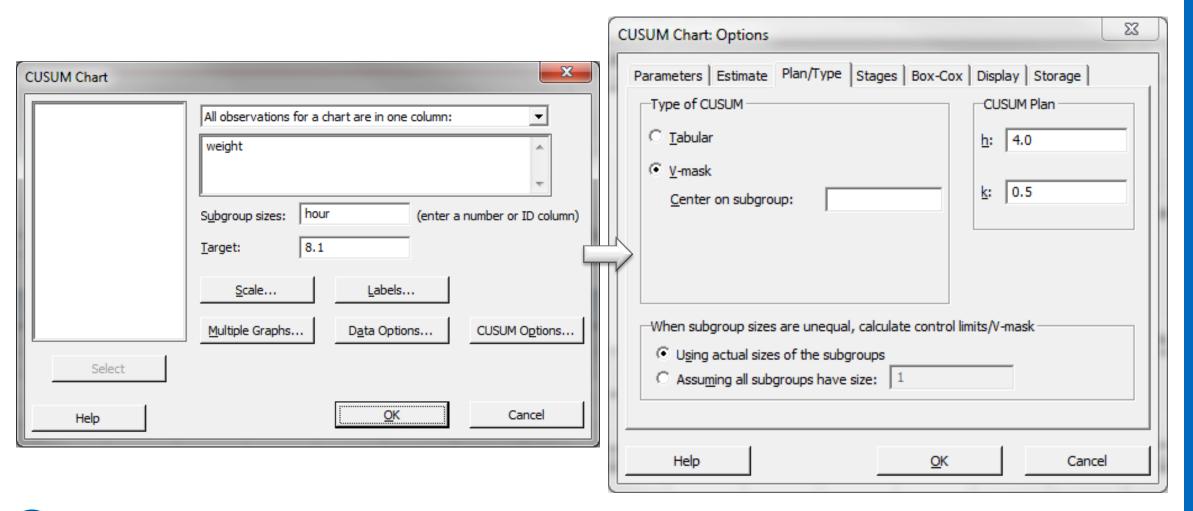


Use Minitab to Plot a CumSum Chart

- Data File: "CumSum" tab in "Sample Data.xlsx"
- Steps in Minitab to plot CumSum charts
 - 1. Click Stat → Control Charts → Time Weighted Charts → CUSUM.
 - 2. A new window named "CUSUM Chart" appears.
 - 3. Select "weight" in the box below "All observations for a chart are in one column."
 - 4. Select "hour" as the "Subgroup sizes."
 - 5. Enter "8.1" as "Target."
 - 6. Click the button "CUSUM Options" and another window named "CUSUM Chart Options" pops up.
 - 7. Click the tab "Plan/Type."
 - 8. Select the radio button "V-mask."
 - 9. Click "OK" in the window "CUSUM Chart Options."
 - 10. Click "OK."
 - 11. The CUSUM chart appears in the newly-generated window.



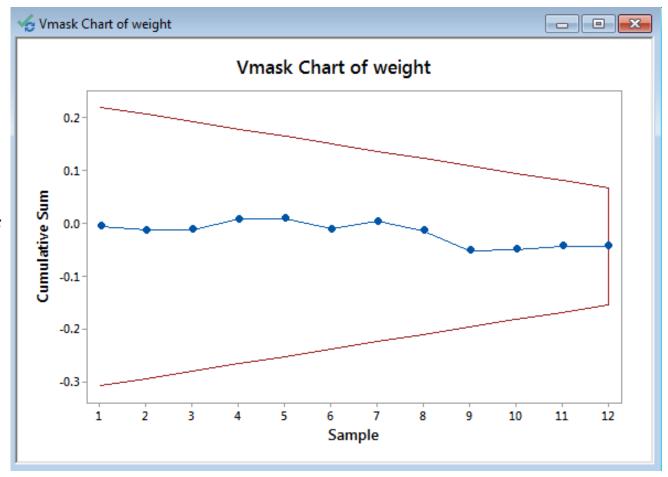
Use Minitab to Plot a CumSum Chart





Use Minitab to Plot a CumSum Chart

- CumSum Chart Interpretation:
 - The process is in control since all of the data points fall between the two arms of the V-mask.





5.2.9 EWMA Chart



EWMA Chart

- The **EWMA chart** (Exponentially-Weighted Moving Average Chart) is a control chart monitoring the exponentially-weighted average of previous and present subgroup means.
- The more recent data get more weight than more historical data.
- It detects the shift of the process mean from the process target over time.
- The underlying distribution of the EWMA chart is normal distribution.



EWMA Chart Equations

EWMA Chart

• Data Point: $z_i = \lambda \overline{x_i} + (1 - \lambda)z_{i-1}$ where $0 < \lambda < 1$

• Center Line: \overline{X}

• Control Limits:
$$\overline{X} \pm k \cdot \frac{s}{\sqrt{n}} \sqrt{\left(\frac{\lambda}{2-\lambda}\right) \left[1 - (1-\lambda)^{2i}\right]}$$

where \bar{x}_i is the mean of the ith subgroup; λ and k are user-defined parameters to calculate the EWMA data points and the control limits.

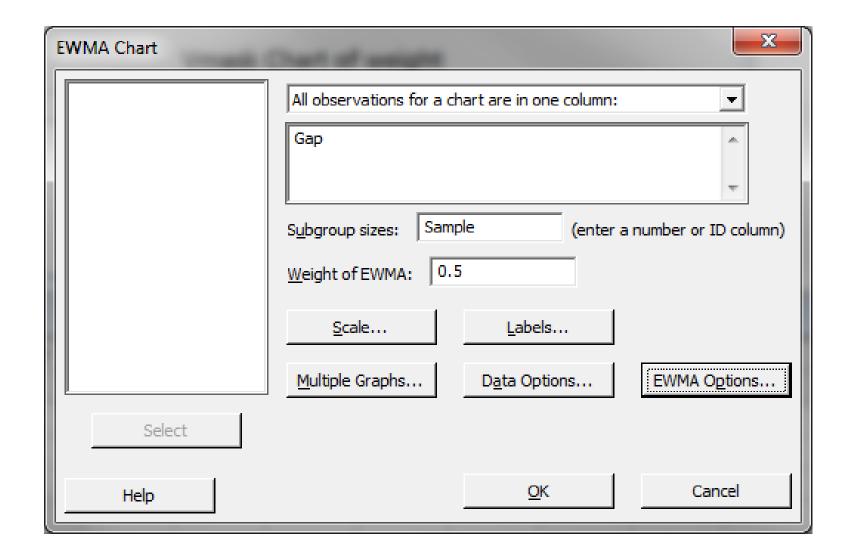


Use Minitab to Plot an EWMA Chart

- Data File: "EWMA" tab in "Sample Data.xlsx"
- Steps in Minitab to plot EWMA charts
 - 1. Click Stat → Control Charts → Time Weighted Charts → EWMA.
 - 2. A new window named "EWMA Chart" appears.
 - 3. Select "Gap" in the box below "All observations for a chart are in one column."
 - 4. Select "Sample" as the "Subgroup sizes."
 - 5. Enter "0.5" as "Weight of EWMA."
 - 6. Click "OK."
 - 7. The EWMA chart appears in the newly-generated window.



Use Minitab to Plot an EWMA Chart

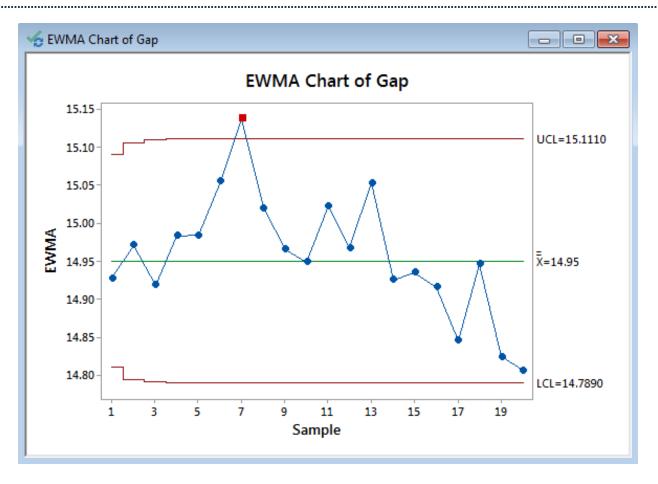




Use Minitab to Plot an EWMA Chart

• EWMA Chart Interpretation:

- One data point falls beyond the upper control limit and we conclude that the process is out of control.
- Further investigation is needed to discover the root cause for the outlier.





5.2.10 Control Methods



Control Methods

- There are many control methods available to keep the process stable and minimize the variation.
- Most common control methods:
 - SPC (statistical process control)
 - 5S method
 - Kanban
 - Poka-Yoke (mistake proofing).



SPC

- SPC (Statistical Process Control) is a quantitative control method to monitor the stability of the process performance by identifying the special cause variation in the process.
- It uses control charts to detect the unanticipated changes in the process.
- Which control chart to use depends on:
 - Whether the data are continuous or discrete
 - How large the subgroup size is
 - Whether the subgroup size is constant
 - Whether we are interested in measuring defects or defectives
 - Whether we are interested in detecting the shifts in the process mean.



5S

- 5S is a systematic approach of cleaning and organizing the workplace.
 - Seiri (sorting)
 - Seiton (straightening)
 - Seiso (shining)
 - Seiketsu (standardizing)
 - Shisuke (sustaining)



Kanban

- A Kanban system is a demand-driven system.
- The customer demand is the signal to trigger or pull the production.
- Products are made only to meet the immediate demand. When there is no demand, there is no production.
- It was designed to reduce the waste in inventory and increase the speed of responding to immediate demand.



Poka-Yoke

 Poka-yoke is a mechanism to eliminate the defects as early as possible in the process.

- Contact Method
 - Use of the shape, color, size, or any other physical attributes of the items
- Constant Number Method
 - Use of a fixed number to make sure a certain number of motions are completed
- Sequence Method
 - Use of a checklist to make sure all the prescribed process steps are followed in the right order.



5.2.11 Control Chart Anatomy



Control Chart Calculations Summary

Chart	Center Line	Control Limits	σ _x
I Chart	$\frac{\sum_{i=1}^{n} \mathbf{x}_{i}}{n}$	$\frac{\sum_{i=1}^{n} X_{i}}{n} \pm 3 \times \frac{\overline{MR}}{d_{2}}$	<u>MR</u> d,
MR Chart	$ \frac{m}{MR} = \frac{\sum_{i=1}^{n-1} x_{i+1} - x_i }{n-1} $	$UCL = D_4 \times \overline{MR}$ $LCL = D_3 \times \overline{MR}$	
Xbar Chart (Xbar-R)	$\overline{\overline{X}} = \frac{\sum_{t=1}^{k} \overline{X}_{t}}{k}$	$\overline{\overline{X}} \pm A_2 \overline{R}$	<u>R</u> d,
Xbar Chart (Xbar-S)	$\overline{\overline{X}} = \frac{\sum_{t=1}^{k} \overline{X}_{t}}{k}$	$\overline{\overline{X}} \pm A_3 \overline{s}$	<u>;</u>
R Chart	$\overline{R} = \frac{\sum_{i=1}^{k} R_i}{k}$	$UCL = D_4 \times \overline{R}$ $LCL = D_3 \times \overline{R}$	
S Chart	$\frac{1}{s} = \frac{\sum_{i=1}^{k} s_i}{k}$	$UCL = B_4 \times \overline{s}$ $LCL = B_3 \times \overline{s}$	
U Chart	$\overline{u} = \frac{\sum_{i=1}^{k} u_i}{k}$	$\frac{\overline{u}\pm 3\times\sqrt{\frac{\overline{u}}{n_i}}}{n_i}$	$\sqrt{\frac{\overline{u}}{n_i}}$
P Chart	$\overline{p} = \frac{\sum_{i=1}^{b} x_i}{\sum_{i=1}^{b} x_i}$	$\overline{p} \pm 3 \sqrt{\frac{\overline{p}(1-\overline{p})}{n_i}}$	$\sqrt{\frac{\overline{p}(1-\overline{p})}{n_i}}$
NP Chart	$n\overline{p} = \frac{\sum_{i=1}^{n} x_i}{k}$	$n\overline{p} \pm 3 \times \sqrt{n\overline{p}(1-\overline{p})}$	$\sqrt{n\overline{p}(1-\overline{p})}$



Control Chart Constants

Subgroup Size	A2	A 3	В3	B4	c4	d2	D3	D4
2	1.88	2.659	_	3.267	0.7979	1.128	-	3.267
3	1.023	1.954	-	2.568	0.8862	1.693	-	2.574
4	0.729	1.628	-	2.266	0.9213	2.059	_	2.282
5	0.577	1.427	-	2.089	0.94	2.326	_	2.114
6	0.483	1.287	0.03	1.97	0.9515	2.534	-	2.004
7	0.419	1.182	0.118	1.882	0.9594	2.704	0.076	1.924
8	0.373	1.099	0.185	1.815	0.965	2.847	0.136	1.864
9	0.337	1.032	0.239	1.761	0.9693	2.97	0.184	1.816
10	0.308	0.975	0.284	1.716	0.9727	3.078	0.223	1.777
15	0.223	0.789	0.428	1.572	0.9823	3.472	0.347	1.653
25	0.153	0.606	0.565	1.435	0.9896	3.931	0.459	1.541



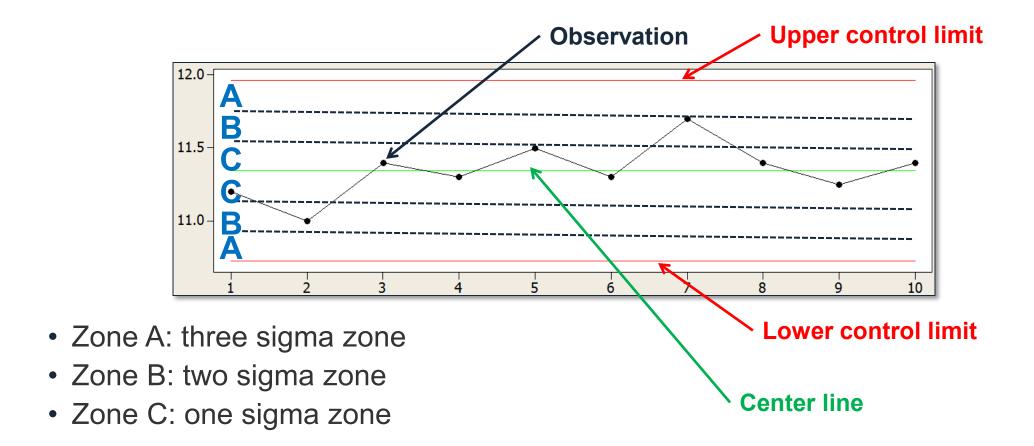
Unnatural Patterns

- If there are *unnatural patterns* in the control chart of a process, we consider the process out of statistical control.
- Typical unnatural patterns in control charts:
 - Outliers
 - Trending
 - Cycling
 - Auto-correlative
 - Mixture.
- A process is *in control* if all the data points on the control chart are randomly spread out within the control limits.



- Western Electric Rules are the most popular decision rules to detect unnatural patterns in the control charts. They are a group of tests for special causes in a process.
- The area between the upper and lower control limits is separated into six subzones.
 - Zone A: between two and three standard deviations from the center line
 - Zone B: between one and two standard deviations from the center line
 - Zone C: within one standard deviation from the center line.

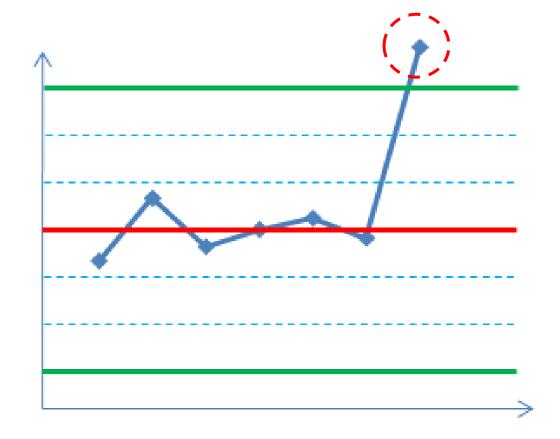




• If a data point falls onto the dividing line of two consecutive zones, the point belongs to the outer zone.

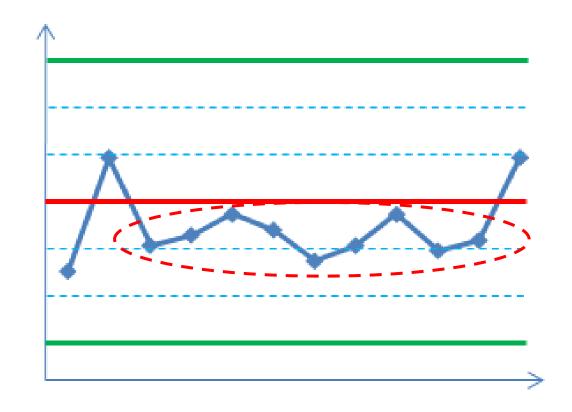


 Test 1: 1 point more than 3 standard deviations from the center line (i.e., 1 point beyond zone A)



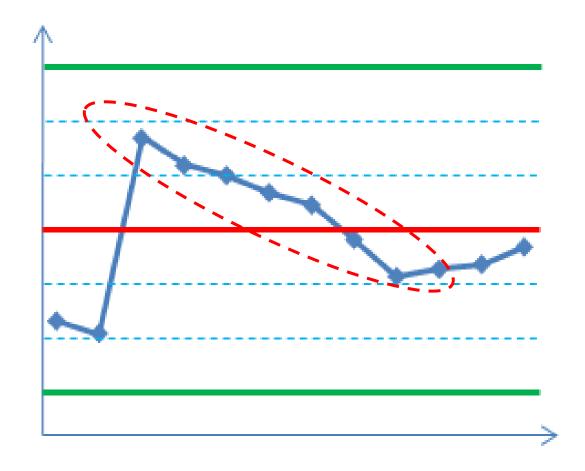


• Test 2: 9 points in a row on the same side of the center line



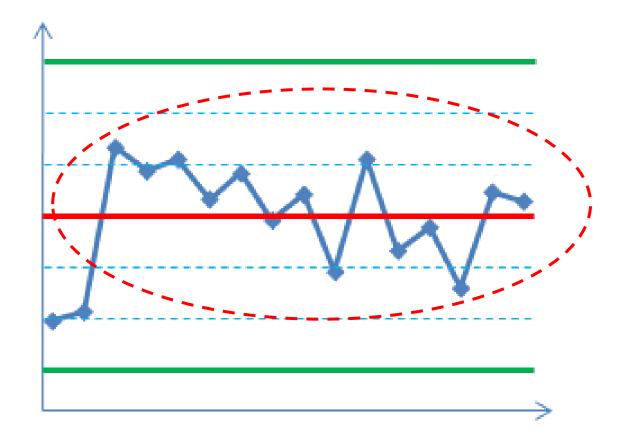


• Test 3: 6 points in a row steadily increasing or steadily decreasing



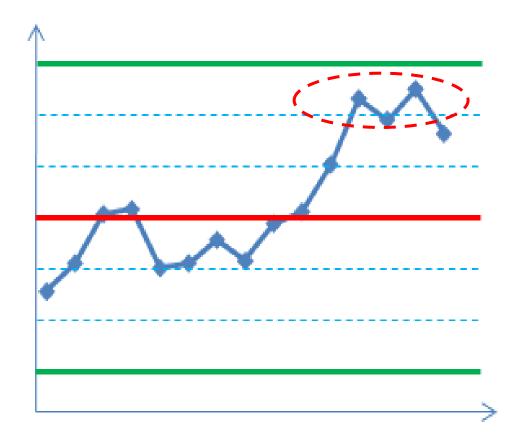


• Test 4: 14 points in a row alternating up and down



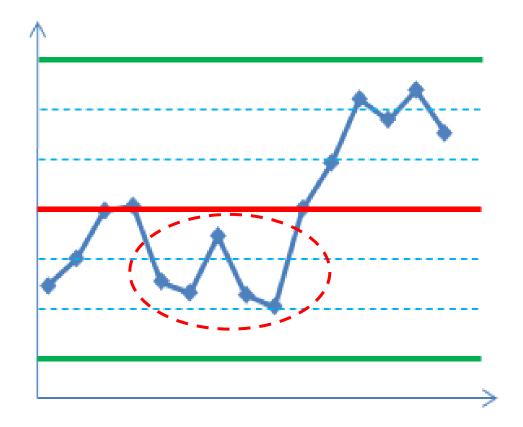


 Test 5: 2 out of 3 points in a row at least 2 standard deviations from the center line (in zone A or beyond) on the same side of the center line



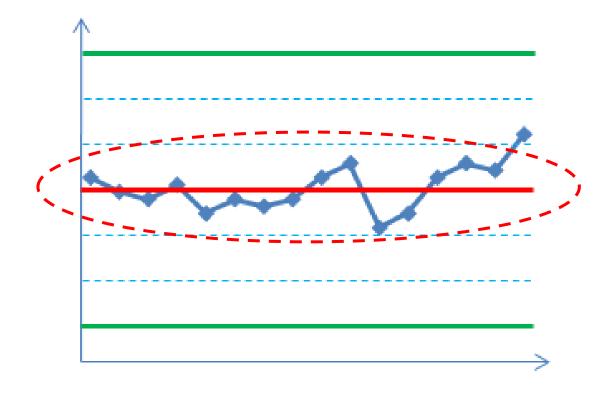


 Test 6: 4 out of 5 points in a row at least 1 standard deviation from the center line (in zone B or beyond) on the same side of the center line



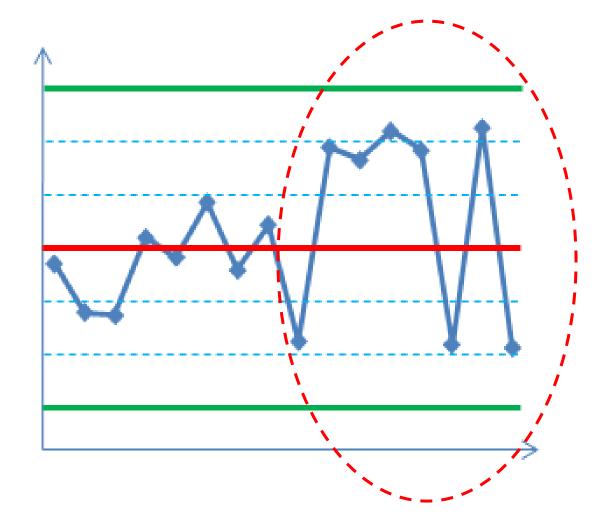


 Test 7: 15 points in a row within 1 standard deviation from the center line (in zone C) on either side of the center line





 Test 8: 8 points in a row beyond 1 standard deviation from the center line (beyond zone C) on either side of the center line





Next Steps

- If no data points fail any tests for special causes, the process is in statistical control.
- If any data point fails any tests for special causes, the process is unstable and we will need to investigate the observation thoroughly to discover and take actions on the special causes leading to the changes.
- Process stability is the prerequisite of process capability analysis.



5.2.12 Subgroups & Sampling



Subgroups

- Rational subgrouping is the basic sampling scheme in SPC (Statistical Process Control).
- When sampling, we randomly select a group (i.e., a subgroup) of items from the population of interest.
- The subgroup size is the count of samples in a subgroup. It can be constant or variable.
- Depending on the subgroup sizes, we select different control charts accordingly.



Impact of Variation

- The rational subgrouping strategy is designed to minimize the opportunity of having special cause variation within subgroups.
- If there is only random variation (background noise) within subgroups, all the special cause variation would be reflected between subgroups. It is easier to detect the out-of-control situation.
- Random variation is inherent and indelible in the process. We are more interested in identifying and taking actions on special cause variation.



Frequency of Sampling

- The frequency of sampling in SPC depends on whether we have sufficient data to signal the changes in a process with reasonable time and costs.
- The more frequently we sample, the higher the costs sampling may trigger.
- We need the subject matter experts' knowledge on the nature and characteristics of the process to make good decisions on sampling frequency.



5.3 Six Sigma Control Plans



Black Belt Training: Control Phase

5.1 Lean Controls

- 5.1.1 Control Methods for 5S
- 5.1.2 Kanban
- 5.1.3 Poka-Yoke (Mistake Proofing)

5.2 Statistical Process Control (SPC)

- 5.2.1 Data Collection for SPC
- 5.2.2 I-MR Chart
- 5.2.3 Xbar-R Chart
- 5.2.4 U Chart
- 5.2.5 P Chart
- 5.2.6 NP Chart

- 5.2.7 X-S chart
- 5.2.8 CumSum Chart
- 5.2.9 EWMA Chart
- 5.2.10 Control Methods
- 5.2.11 Control Chart Anatomy
- 5.2.12 Subgroups, Variation, Sampling

5.3 Six Sigma Control Plans

- 5.3.1 Cost Benefit Analysis
- 5.3.2 Elements of the Control Plan
- 5.3.3 Elements of the Response Plan



5.3.1 Cost Benefit Analysis



What is Cost-Benefit Analysis?

- The cost-benefit analysis is a systematic method to assess and compare the financial costs and benefits of multiple scenarios in order to make sound economic decisions.
- A cost-benefit analysis is recommended to be done at the beginning of the project based on estimations of the experts from the finance team in order to determine whether the project is financially feasible.
- It is recommended to update the cost-benefit analysis at each DMAIC phase of the project.



Why Cost-Benefit Analysis?

- In the Define phase of the project, the cost-benefit analysis helps us understand the financial feasibility of the project.
- In the middle phases of the project, updating and reviewing the costbenefit analysis helps us compare potential solutions and make robust data-driven decisions.
- In the Control phase of the project, the cost-benefit analysis helps us track the project's profitability.



Return on Investment

• Return on investment (also called ROI, rate of return, or ROR) is the ratio of the net financial benefits (either gain or loss) of a project or investment to the financial costs.

$$ROI = \frac{TotalNetBenefits}{TotalCosts} \times 100\%$$

where

TotalNetBenefits = TotalBenefits - TotalCosts



Return on Investment (ROI)

- The return on investment is used to evaluate the financial feasibility and profitability of a project or investment.
 - If ROI < 0, the investment is not financially viable.
 - If ROI = 0, the investment has neither gain nor loss.
 - If ROI > 0, the investment has financial gains.
- The higher the ROI, the more profitable the project.



Net Present Value (NPV)

 The net present value (also called NPV, net present worth, or NPW) is the total present value of the cash flows calculated using a discount rate.

$$NPV = \frac{NetCashFlow_t}{(1+r)^t}$$

Where

NetCashFlow_t is the net cash flow happening at time t;r is the discount rate;t is the time of the cash flow.



Cost Estimation

- Examples of costs triggered by the project:
 - Administration
 - Asset
 - Equipment
 - Material
 - Delivery
 - Real estate
 - Labor
 - Training
 - Consulting.



Benefits Estimation

- Examples of benefits generated by the project:
 - Direct revenue increase
 - Waste reduction
 - Operation cost reduction
 - Quality and productivity improvement
 - Market share increase
 - Cost avoidance
 - Customer satisfaction improvement
 - Associate satisfaction improvement.



Challenges in Cost and Benefit Estimation

- Different analysts might come up with different cost and benefit estimations due to their subjectivity in determining:
 - The discount rate
 - The time length of the project and its impact
 - Potential costs of the project
 - The tangible/intangible benefits of the project
 - The specific contribution of the project to the relevant financial gains/loss.



5.3.2 Elements of Control Plans



Control Plans

- The control plans ensure that the changes introduced by a Six Sigma project are sustained over time.
- Benefits of the Control phase:
 - Methodical roll-out of changes including standardization of processes and work procedures
 - Ensure compliance with changes through methods like auditing and corrective actions
 - Transfer solutions and learning across the enterprise
 - Plan and communicate standardized work procedures
 - Coordinate ongoing team and individual involvement
 - Standardize data collection and procedures
 - Measure process performance, stability, and capability
 - Plan actions that mitigate possible out-of-control conditions
 - Sustain changes over time.



What is a Control Plan?

- A control plan is a management planning tool to identify, describe, and monitor the process performance metrics in order to meet the customer specifications steadily.
- It proposes the plan of monitoring the stability and capability of inputs and outputs of critical process steps in the Control phase of a project.
- It covers the data collection plan of gathering the process performance measurements.
- Control plans are the most overlooked element of most projects. It is critical that a good solution be solidified with a great control plan!



Control Plan Elements

Control Plan

- The clear and concise summary document that details key process steps, CTQs metrics, measurements, and corrective actions.
- Standard Operating Procedures (SOPs)
 - Supporting documentation showing the "who does what, when, and how" in completing the tasks.
- Communication Plan
 - Document outlining messages to be delivered and the target audience.
- Training Plan
 - Document outlining the necessary training for employees to successfully perform new processes and procedures.
- Audit Checklists
 - Document that provides auditors with the audit questions they need to ask.
- Corrective Actions
 - Activities that need to be conducted when an audit fails.



Control Plan

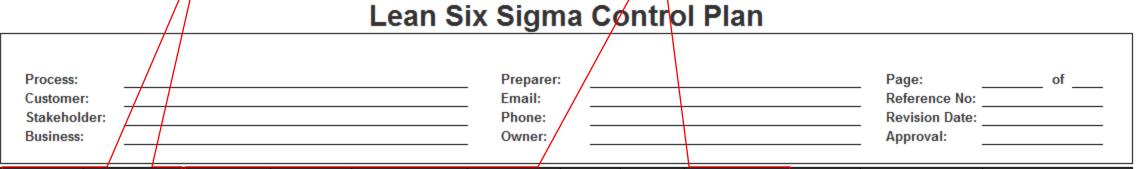
- The control plan identifies critical process steps that have significant impact on the products or services and the appropriate controls mechanisms.
- The control plan includes measurement systems that monitor and help manage key process step performance.
- Specified limits and targets of the performance metrics are clearly defined and communicated.
- Sampling plans to collect the measurements are declared:
 - How many samples are needed?
 - How often do we need to sample?
 - Where should we sample?



Control Plan

Key processes and process steps are identified

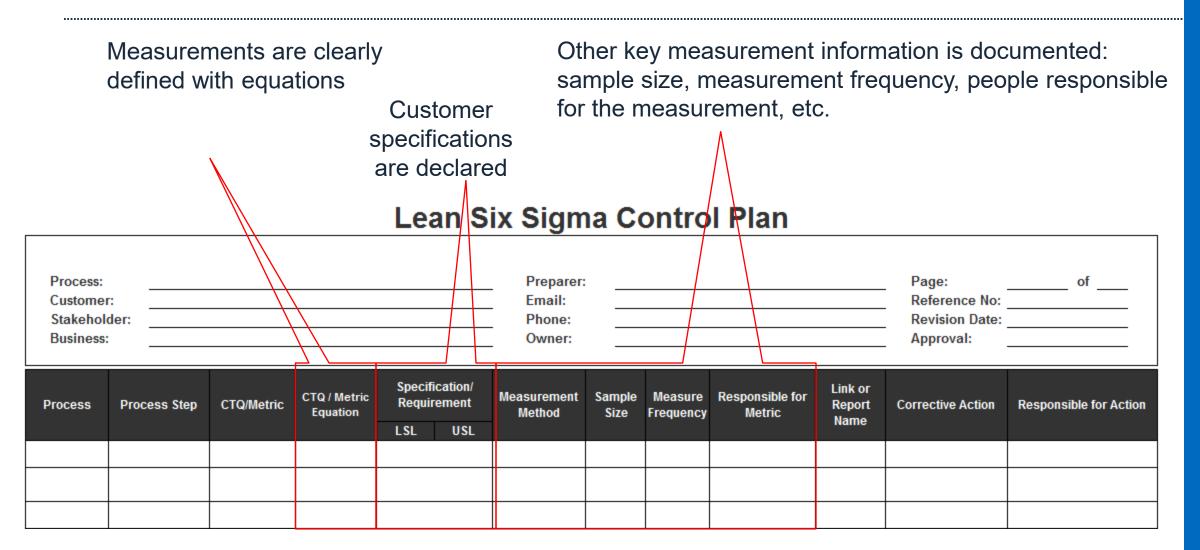
Critical information regarding key measurements is documented and clarified



Process	Process Step	CTQ/Metric	CTQ / Metric Equation	Specification/ Requirement		Measurement Sam Method Siz		Link or Report Name	Corrective Action	Responsible for Action
				LSL	USL			Name		



Control Plan





Control Plan

Where will this measurement or report be found? Good control plans provide linking information or other report reference information.

Control plans identify the mitigating action or corrective actions required in the event the measurement falls out of spec or control. Responsible parties are also declared.

Process: Customer: Stakeholder: Business: Preparer: Email: Preparer: Email: Preparer: Email: Preparer: Email: Preparer: Email: Phone: Owner: Approval:

	Process	Process Step	CTQ/Metric	CTQ / Metric Equation		ication/ rement	Measurement Method	Measure Frequency	Responsible for Metric	Link or Report Name	Corrective Action	Responsible for Action
					LSL	USL				Name		
ľ												



Standard Operating Procedures (SOPs)

- Standard Operating Procedures (SOPs) are documents that focus on process steps, activities, and specific tasks required to complete an operation.
- SOPs should not be much more than two to four pages.
- SOPs should be written to the user's level of required detail and information.
 - The level of detail is dependent on the position's required skills and training
- Good SOPs are auditable, easy to follow, and not difficult to find.
 - Auditable characteristics are: observable actions and countable frequencies.
 Results should be evident to a third party (compliance to the SOP must be measurable).



SOP Elements

- SOPs are intended to impart high value information in concise and well-documented manner.
- SOP Title and Version Number:
 - Provide a title and unique identification number with version information.
- Date:
 - List the original creation date; add all revision dates.
- Purpose:
 - State the reason for the SOP and what it intends to accomplish.
- Scope:
 - Identify all functions, jobs, positions, and/or processes governed or affected by the SOP.



SOP Elements

- Responsibilities:
 - Identify job functions and positions (not people) responsible for carrying out activities listed in the SOP.
- Materials:
 - List all material inputs: parts, files, data, information, instruments, etc.
- Process Map:
 - Show high level or level two to three process maps or other graphical representations of operating steps.
- Process Metrics:
 - Declare all process metrics and targets or specifications.
- Procedures:
 - List actual steps required to perform the function.
- References:
 - List any documents that support the SOP.



SOP Template

Standard Operating Procedures

SOP Name/Title:			
Document Storage Loca	tion/Source:	Document No:	
SOP Originator:	Approving Position:	Effective Date:	
Name:	Name:	Last Edited Date:	
Signature:	Signature:	Other:	

1. Purpose

"Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut

2. Scope

"Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut

3. Responsibilities (RACI)

Responsible	Accountable	Consulted	Informed
John Doe	Jane Doe	Jack Doe	Jill Doe
Pam Doe	Paul Doe	Phil Doe	Peggy Doe

4. Materials

"Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut

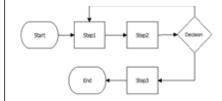
5. Related Documents

"Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut

6. Definitions

"Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut

7. Process Map



8. Procedures

Step	Action	Responsible
1		
2		
3		

9. Process Metrics

- "Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut
- "Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut

10. Resources

- "Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut
- "Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut



Communication Plans

- Communication plans are documents that focus on planning and preparing for the dissemination of information.
- Communication plans organize messages and ensure that the proper audiences receive the correct message at the right time.
- A good communication plan identifies:
 - Audience
 - Key points/message
 - Medium (how the message is to be delivered)
 - Delivery schedule
 - Messenger
 - Dependencies and escalation points
 - Follow-up messages and delivery mediums.
- Communication plans help develop and execute strategies for delivering changes to an organization.



Communication Plan Template

Process/Funct	ion Name	Project/Prog	<u>Project Lead</u>		<u>Project Sponsor/Champion</u>				
ommunication Pu	rpose:								
Target Audience	Key Message	Message Dependencies	Delivery Date	Location	Medium	Follow up Medium	Messenger	Escalation Path	Contact Information
					_				



Training Plans

- Training plans are used to manage the delivery of training for new processes and procedures.
- Most GB or BB projects will require changes to processes and/or procedures that must be executed or followed by various employees.
- Training plans should incorporate all SOPs related to performing new or modified tasks.
- Training plans use and support existing SOPs and do not supersede them.
- Training plans should include logistics:
 - One-on-one or classroom
 - Instruction time
 - Location of training materials
 - Master training reference materials
 - Instructors and intended audience
 - Trainee names.



Training Plan Template

_	1		ining Pla									
_	Sponsor		on	Business Divisio		Project Lead		Process		ject	Project	
S	Status	Trainer	Other Requirements	Technology Requirements	Supporting Docs	Training Medium	Key Change/Process	How Many	When	Where	Who	
_												
_												
-												
-												
-												
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_	\bot											
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Audits

- What is an audit?
 - ISO 9000 defines an Audit as "a systematic and independent examination to determine whether quality activities and related results comply with planned arrangements and whether these arrangements are implemented effectively and are suitable to achieve objectives."
- Audits are used to ensure actions, processes, procedures, and other tasks are performed as expected.



Audit Guidelines

- Audits should be directed by managers, supervisors, and other accountable positions.
- An audit's purpose must be well-defined and executed by independent unbiased personnel.
- Auditors must:
 - Be qualified to perform their tasks
 - Attend and successfully complete an internal auditing training session
 - Be able to identify whether or not activities are being followed according to the defined SOP
 - Base conclusions on facts and objective evidence
 - Use a well documented audit checklist.
- Audits should confirm compliance or declare non-compliance.



Audit Checklists

 Auditors should review the SOPs before preparing checklists or ensure that existing checklists properly reference SOPs.

Audit checklists:

- Serve as guides for identifying items to be examined
- Are used in conjunction with understanding of the procedure
- Ensure a well-defined audit scope
- Identify needed facts during audits
- Provide places to record gathered facts.

Checklists should include:

- A review of training records
- A review of maintenance records
- Questions or observations that focus on expected behaviors
- Questions should be open-ended where possible
- Definitive observations yes/no, true/false, present/absent, etc.



Audit Checklist Template

Target Area:	Statement of Audit Objective:	Auditor:	Audit Date
Audit Technique	Auditable Item, Observation, Procedure etc.		al Auditor ircle Rating
Observation	Have all associates been trained?	YES	NO
Observation	Is training documentation available?	YES	NO
Observation	Is training documentation current?	YES	NO
Observation	Are associates wearing proper safety gear?	YES	NO
Observation	Are SOP's available?	YES	NO
Observation	Are SOP's current?	YES	NO
Observation	Is quality being measured	YES	NO
Observation	Is sampling being conducted in random fashion	YES	NO
Observation	Is sampling meeting it's sample size target?	YES	NO
Observation	Are control charts in control	YES	NO
Observation	Are control charts current?	YES	NO
Observation	Is the process capability index >1.0?	YES	NO
Number of Out of Co	mpliance Observations		
Total Observations			
Audit Yield			#DIV/0!
Corrective Actions R	equired		
Auditor Comments			



5.3.3 Response Plan Elements



What is a Response Plan?

- A response plan should be a component of as many control plan elements as possible.
- Response plans are a management planning tool to describe corrective actions necessary in the event of out-of-control situations.
- There is never any guarantee that processes will always perform as designed. Therefore, it is wise to prepare for occasions when special causes are present.
- Response plans help us mitigate risks and, as already mentioned, should be part of several control plan elements.



Response Plan Elements

- Action triggers
 - When do we need to take actions to correct a problem or issue?
- Action recommendation
 - What activities are required in order to solve the problem in the process?
 The action recommended can be short-term (quick fix) or long-term (true process improvement).
- Action respondent
 - Who is responsible for taking actions?
- Action date
 - When did the actions happen?
- Action results
 - What actions have been taken?
 - When were actions taken?
 - What are the outcomes of the actions taken?



Response Plan Elements

Note the response plan element in this control plan template

Lean Six Sigma Control Plan

		\	
Process:	Preparer: Page	: \	of
Customer:	Email: Refer	ence No:	
Stakeholder:	Phone: Revis	ion Date:	
Business:	Owner: Appro	oval:	
		\	

Process	Process Step	CTQ/Metric	CTQ / Metric Equation		ication/ rement	Measurement Method	Measure Frequency	Responsible for Metric	Link or Report	Corrective Action	Responsible for Action
				LSL	USL				Name		



Response Plan Elements

Process/Functi	on Name	Project/Progr	<u>Project Lead</u>		Project Sponsor/Champion				
Communication Pu	rpose:								
Target Audience	Key Message	Message Dependencies	Delivery Date	Location	Medium	Follow up Medium	Messenger	Escalation Path	Contact Information
~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	A attend			~	~~~				-

Note the response plan element in this communication plan template



Audit Checklist

# Response Plan Elements

Note the response plan element in this audit checklist

Target Area:	Statement of Audit Objective:	Auditor:	Audit Date:
		Individu	al Auditor
Audit Technique	Auditable Item, Observation, Procedure etc.	Rating (C	ircle Rating
Observation	Have all associates been trained?	YES	NO
Observation	Is training documentation available?	YES	NO
Observation	Is training documentation current?	YES	NO
Observation	Are associates wearing proper safety gear?	YES	NO
Observation	Are SOP's available?	YES	NO
Observation	Are SOP's current?	YES	NO
Observation	Is quality being measured	YES	NO
Observation	Is sampling being conducted in random fashion	YES	NO
Observation	Is sampling meeting it's sample size target?	YES	NO
Observation	Are control charts in control	YES	NO
Observation	Are control charts current?	YES	NO
Observation	Is the process capability index >1.0?	YES	NO
Jumber of Out of	Compliance Observations		
otal Observations			
Audit Yield			#DIV/0!
Corrective Actions	Required		1
Auditor Commer	•		





Lean Six Sigma Green Belt Training
Featuring Examples from Minitab 18

