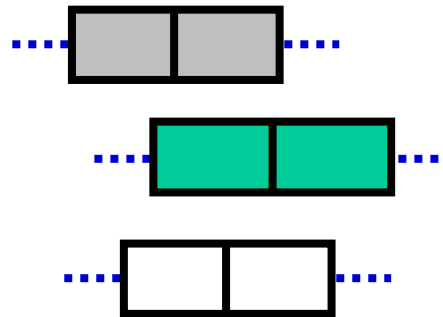


Continuous Improvement Toolkit

ANOVA



Managing Risk

PDPC

FMEA RAID Logs

Fault Tree Analysis

Risk Assessment*

Decision Tree

Traffic Light Assessment

Lean Measures

KPIs

OEE

Capability Indices

MSA

RTY

Descriptive Statistics

Cost of Quality

Probability Distributions

Reliability Analysis

Graphical Analysis

Understanding Performance

Run Charts

Control Charts

Benchmarking

Sampling

Focus groups

Interviews

Photography

Check Sheets

Measles Charts

Surveys

Data Collection

Critical Incident Technique

Observations

Deciding & Selecting

Pros and Cons

Importance-Urgency Mapping

Break-even Analysis

Cost -Benefit Analysis

Force Field Analysis

Pugh Matrix

Voting

SWOT

Decision Tree

QFD

Matrix Diagram

TPN Analysis

Kano Analysis

Prioritization Matrix

Critical-to Tree

Paired Comparison

Cause & Effect Matrix

Pareto Analysis

Confidence Intervals

ANOVA

Graphical Analysis

Hypothesis Testing

Scatter Plot

Correlation

5 Whys

Chi-Square Test

Fishbone Diagram

TRIZ***

Brainstorming

Analogy

SCAMPER***

Nominal Group Technique

Mind Mapping*

Affinity Diagram

Attribute Analysis

Lateral Thinking

Visioning

Creating Ideas**

Planning & Project Management*

RACI Matrix

Stakeholders Analysis

PEST

PERT/CPM

Activity Diagram

Roadmaps

Project Charter

Gantt Chart

PDCA

Control Planning

Gap Analysis

Hoshin Kanri

Kaizen

How-How Diagram

Tree Diagram**

Standard work

Simulation

TPM

Mistake Proofing

Pull Systems

JIT

Ergonomics

Work Balancing

Automation

Bottleneck Analysis

Visual Management

Flow

Value Analysis

5S

Wastes Analysis

SMED

Time Value Map

Process Redesign

IDEF0

Value Stream Mapping

SIPOC

Flow Process Chart

Process Mapping

Flowcharting

Service Blueprints

Designing & Analyzing Processes

Understanding Cause & Effect

Design of Experiments

Regression

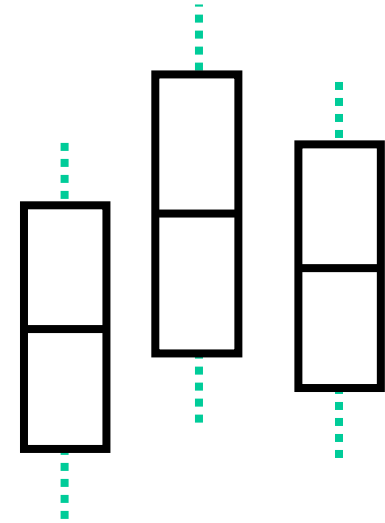
Bottleneck Analysis

Multi-Vari Charts

Relations Mapping*

- ANOVA

- ❑ Analysis of Variance.
- ❑ Used to determine whether the mean responses for two or more groups differ.
- ❑ We can use ANOVA to compare the means of three or more population.
- ❑ If we are only comparing two means, then ANOVA will give the same results as the 2-samples t-test.
- ❑ The Math is different, but the approach and interpretation of p-values is the same.



- ANOVA

- ❑ We must be clear of the hypotheses before applying the technique.
- ❑ A hypothesis test used to determine whether **two or more sample means are significantly different** by comparing the variances between groups.

The Null Hypothesis → The sample means are all the same.
The Alternative Hypothesis → They are not all the same (at least one of them differs significantly from the others).

- ❑ Be careful how you phrase: “**There is a difference**” not “**They are all different**”.

- ANOVA

Some important terms used in ANOVA:

❑ **Factor:**

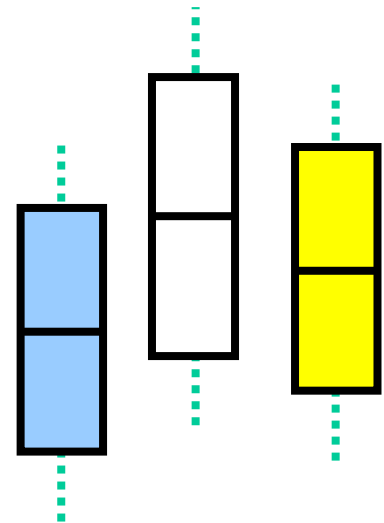
- ❑ The explanatory variable in the study.
- ❑ The factor is categorical (the data classify people, objects or events).

❑ **Levels:**

- ❑ The groups or categories that comprise a factor.

❑ **Response:**

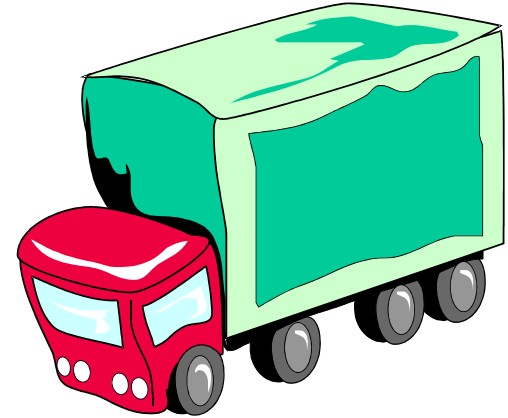
- ❑ A variable (continuous) being measured in the study.



- ANOVA

Example:

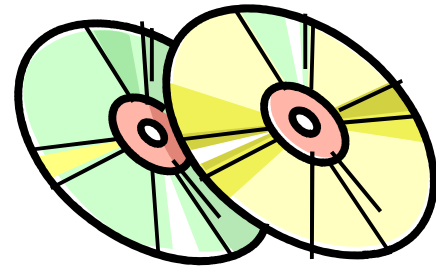
- ❑ If we select the supplier to be the **factor**, each supplier represent a **level** (the group within a factor).
- ❑ In **one-way ANOVA**, there is only one factor.
- ❑ ANOVA is used to compare the means of the factor levels to determine whether the levels differ.
- ❑ Where is the **response**?



- ANOVA

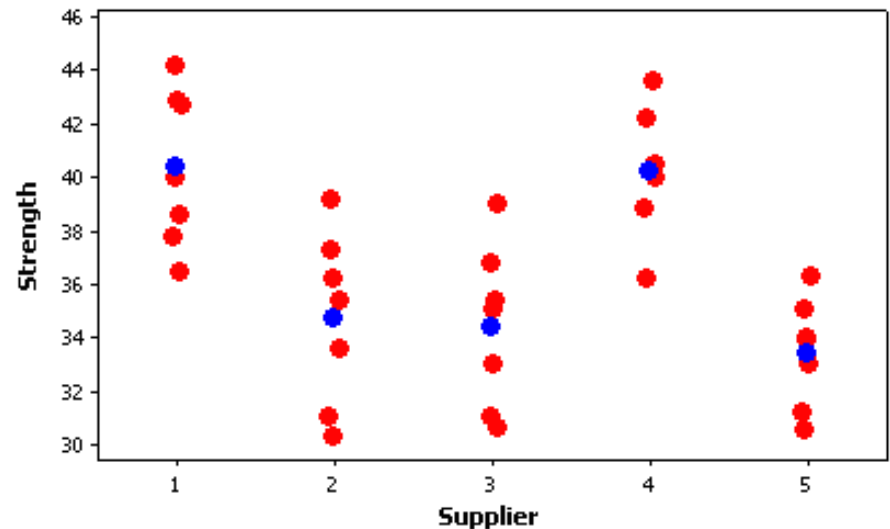
Example:

- ❑ If a company wants to purchase one of three expensive software packages:
- ❑ The software would be the **factor** because it is our explanatory variable.
- ❑ The three software packages are the **levels** that comprise the factor.
- ❑ The amount of time it takes to fill out a report would be the **response** (because it is the particular variable being measured).



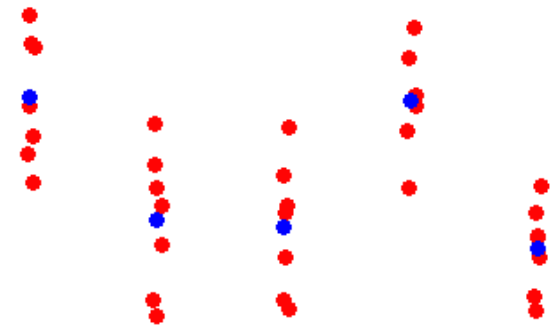
- ANOVA

- ❑ In ANOVA, it is useful to graph the data.
- ❑ We can examine the factor level means and look at the variation **within** each group and **between** all groups.
- ❑ However, the graph will give no idea if the differences between the means are statistically significant.



- ANOVA

- ❑ **Within-group variation** is the variability in measurements within individual groups.
- ❑ **Between-group variation** is the variability in measurements between all groups.
- ❑ We compare between-group variation to within-group variation to determine whether real differences between groups exist.
- ❑ If the between-group variation is large relative to the within-group variation, evidence suggests that the population means are not the same, and vice versa.



- ANOVA

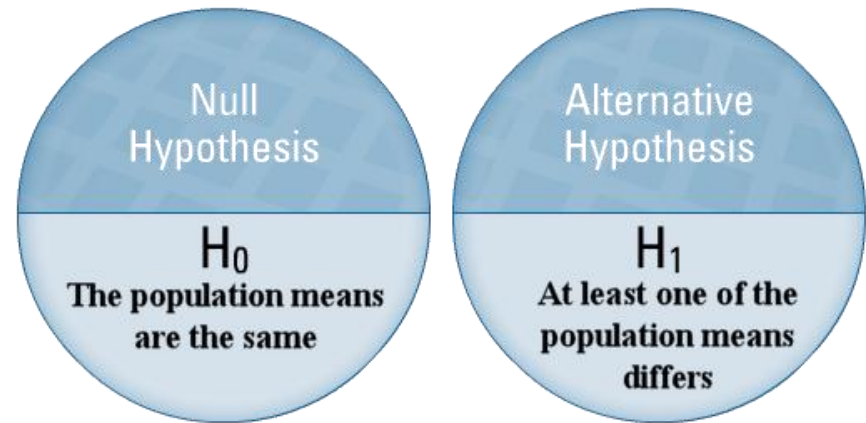
- A **test** to be conducted to decide whether differences between group means are real or simply random error.
- We can compare between and within group variations using **F-statistic ratio**.

$$\text{F-statistic} = \frac{\text{Between-group variation}}{\text{Within-group variation}}$$

- **When F is large** → between-group variation is larger than within group variation, which indicates a real difference between group means.
- **When F is small** → little or no evidence of a significant difference between group means.

- ANOVA

- ❑ When comparing the means of two population, we can use either the 2-sample t-test or **one-way ANOVA**.
- ❑ We must first define the null and alternative hypotheses.
- ❑ We need to use the **p-value** from the ANOVA output to determine whether we should reject or fail to reject the null hypothesis.



- ANOVA

Example:

- ❑ An automobile company uses nylon from five different suppliers to manufacture automobile safety belts.
- ❑ Suppose after establishing the hypothesis & collecting random samples, the results are:

Source	DF	SS	MS	F	P
Supplier	4	322.63	80.66	10.50	0.000
Error	30	230.44	7.68		
Total	34	553.06			



- ❑ As **p-value < 0.05**, we will reject the null hypotheses.
- ❑ The fiber strength for at least one supplier is different from the others.

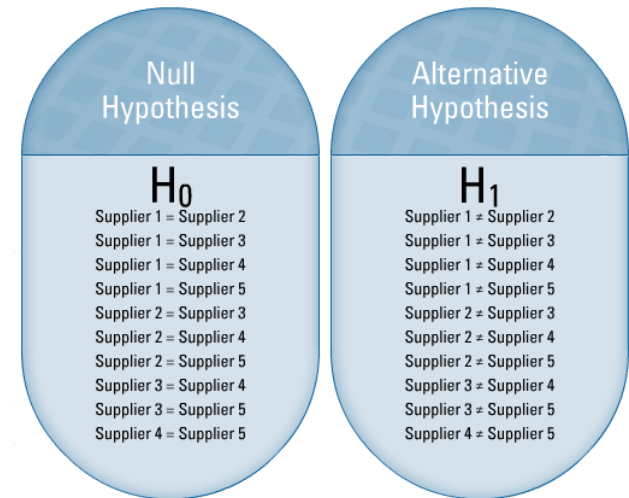
- ANOVA

- ❑ In the previous example, we need to know which suppliers produce the strongest fiber.
- ❑ For this purpose, we can use multiple comparisons.
- ❑ The **multiple comparisons** are the simultaneous testing of multiple hypotheses.
- ❑ We will use a method called **Tukey's test** multiple comparisons, which checks for differences in pairs of group means.



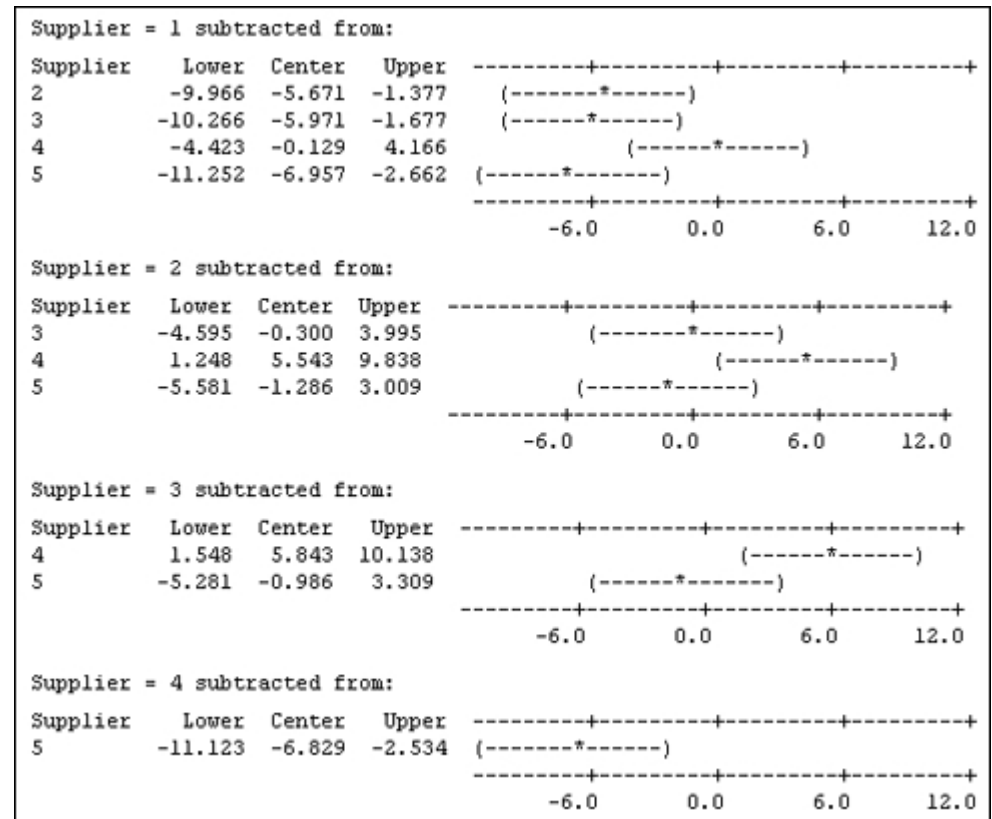
- ANOVA

- ❑ Each individual comparison is like a **2-sample t-test**.
- ❑ For each comparison, we will examine the **confidence interval** to determine whether there is a significant difference between the groups.
- ❑ Each confidence interval provides a range of likely values for the difference between the two population means.
- ❑ If the confidence interval does not contain the value **zero**, then we reject the null hypothesis and conclude that the two group are different.



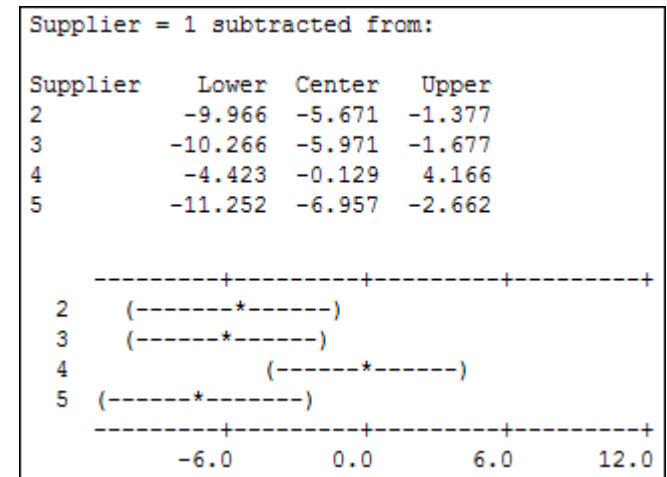
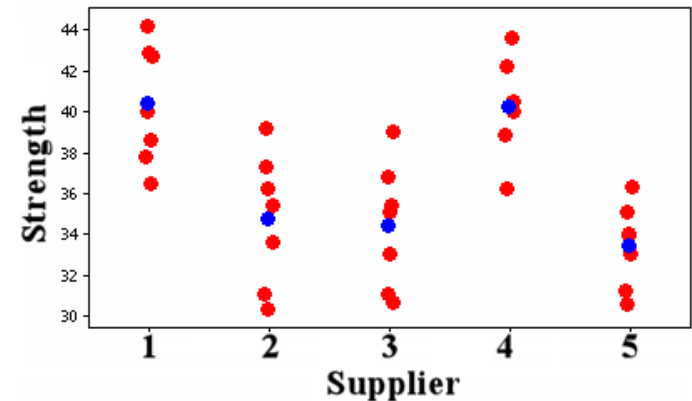
- ANOVA

- ❑ Here are the results of all the comparisons.
- ❑ For example, we will reject the null hypothesis for supplier 2 and 4.
- ❑ Therefore, there strength are different.



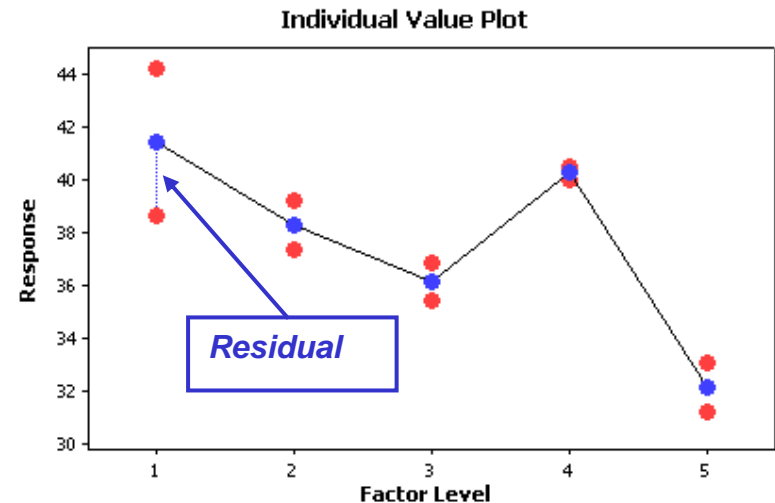
- ANOVA

- ❑ **Question:** Which two suppliers have the higher mean strength measurements than the others?
- ❑ **Answer:** 1 and 4.
- ❑ **Question:** Is there a statistical difference between suppliers 1 and 4? Or which one is the absolute strongest?
- ❑ **Answer:** No, supplier 4 contains the value zero when comparing to supplier 1.



- ANOVA

- ❑ The **residuals** estimates the error in ANOVA.
- ❑ They are calculated by subtracting the observed value from the fitted value (the group mean if the sample size is the same for each group).
- ❑ We can examine the plots of the residuals to check the ANOVA assumptions.
- ❑ Errors in ANOVA (residuals) should be random independent, normally distributed and have constant variance across all factor levels.



- ANOVA

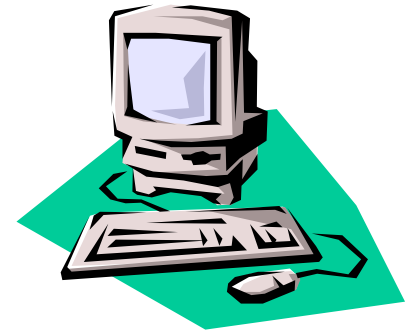
- ❑ When we have two factors, we can use two-way ANOVA to investigate differences among group means.
- ❑ In two-way ANOVA, we use the one-way ANOVA terms (factor, levels and response).

- ❑ **New terms:**
 - ❑ **Main effect:** The influence of a single factor on a response.
 - ❑ **Interaction:** An interaction between factors is present when the mean response for the levels of one factor depends on the level of the second factor.

- ANOVA

Example:

- ❑ An IT Consultancy employs a variety of software developers to provide custom software solutions.
- ❑ It has programmers, testers and system administrators.
- ❑ Company's training programs are classroom teaching, instructional videotapes, and one-on-one training.
- ❑ The manager wants to determine **how to best leverage the company's training programs.**
- ❑ It can save the company money in the long run if the right employees are trained in the best possible.



- ANOVA

Example:

❑ Factors:

- ❑ Job type & method of instruction.

❑ Levels:

- ❑ **Job type:** programmers, testers & system administrators.
- ❑ **Method of instruction:** classroom teaching, instructional videotapes, and one-on-one training.

❑ Response:

- ❑ Impact on employees (**effectiveness of training via test**).

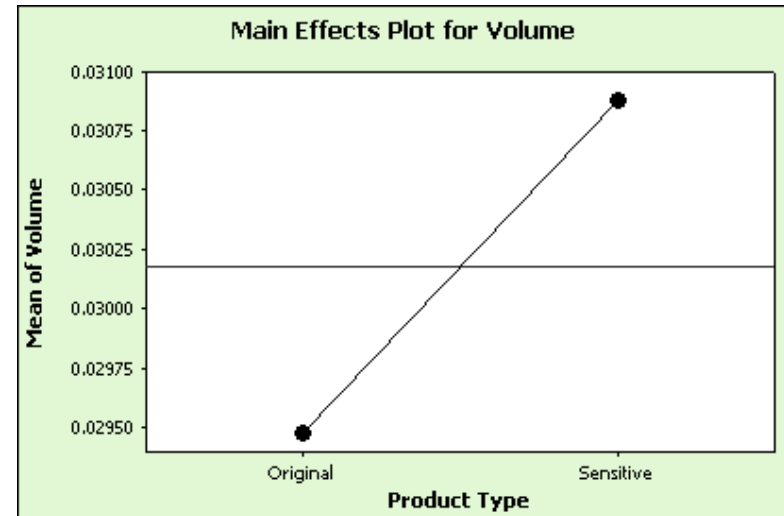
❑ Main effect and Interaction:

- ❑ If a particular job type achieves higher scores from using a given method of instruction, is this an **interaction** or a **main effect**?



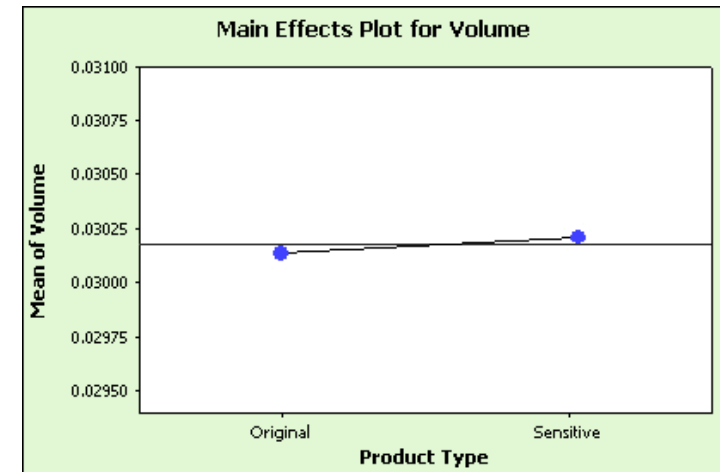
- ANOVA

- ❑ How can we get a better idea of how factors might be affecting the response?
- ❑ The main effects plot can show whether each factor **individually** influences the response.
- ❑ First we'll calculate the mean for each level of the two factors.
- ❑ Then we plot these values on the graph and draw a line to connect the points.
- ❑ Finally, we add a reference line at the overall mean of the data.



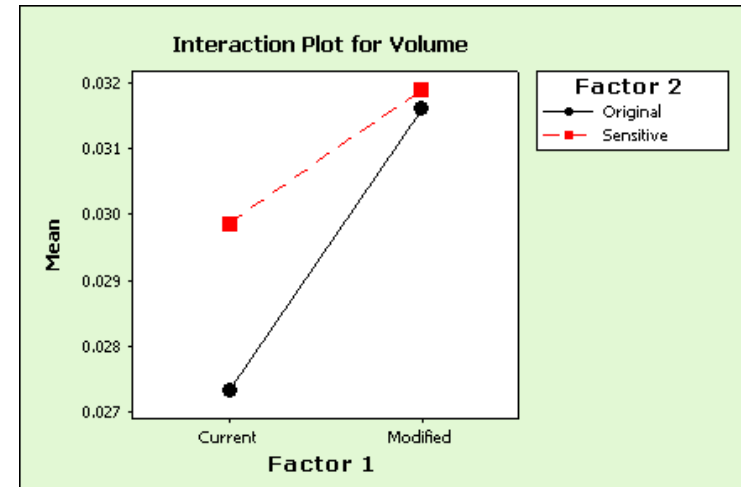
- ANOVA

- ❑ If the line connecting the points is horizontal, this indicates that no main effect is present.
- ❑ If the output line is not horizontal, the main effect is present.
- ❑ The greater the slope of the line, the stronger the effect.
- ❑ Remember to include measurements of all factor levels for the other factor.



- ANOVA

- ❑ To show how the effect of one factor interacts with the effect of another, we'll use the interaction plot.
- ❑ We need to calculate the mean of each combination of levels for the two factors.
- ❑ Then we'll connect each pair of points with a line (in different color).
- ❑ When an interaction exists, the connecting lines on the interaction plot are not parallel. They intersect or almost intersect (as shown here).



- ANOVA

2 Ways ANOVA Approach:

- ❑ Establish the hypothesis.
 - ❑ For each factor.
 - ❑ For the interaction between the factors.

The Null Hypothesis → Either no main effect is present or no interaction effect is present.

The Alternative Hypothesis → Either a main effect is present or an interaction effect is present.

- ❑ Collect random samples.
- ❑ Graph the data (main effects and interaction plots).
- ❑ Conduct the 2 Ways ANOVA and interpret the results.

- ANOVA

Example:

- ❑ Let's get back to the IT consultancy study.
- ❑ The null and the alternative hypothesis:

H0: Mean scores on the test are the same of each training method.

H1: Mean scores on the test are different for at least one of the training methods.

H0: Mean scores on the test are the same of all job types.

H1: Mean scores on the test are different for at least 1 of the job types.

H0: The Change in the mean response across levels of training methods does not depend on job type.

H1: The Change in the mean response across levels of training methods does depend on job type.

- ANOVA

Example:

- ❑ In two-way ANOVA, we first need to test whether a significant interaction effect is present.
- ❑ We will need to consider the **F-statistic** and the **p-value** to determine whether the interaction is statistically significant.
- ❑ If the interaction is significant, it does not make sense to look at the main effect for each factor individually.
- ❑ Like in one-way ANOVA, we can examine the plots of the residuals to check the ANOVA assumptions.

Source	DF	Seq SS	Adj SS	Adj MS	F	P
Factor 1	1	0.0000795	0.0000795	0.0000795	6.36	0.013
Factor 2	1	0.0004007	0.0004007	0.0004007	32.07	0.000
Factor 1 * Factor 2	1	0.0000504	0.0000504	0.0000504	4.03	0.046
Error	156	0.0019492	0.0019492	0.0000125		
Total	159	0.0024798				