

Module 6

Hypothesis

WHAT IS A HYPOTHESIS?

- hypothesis may be defined as a proposition set forth as an explanation for the occurrence of phenomena either asserted merely as a provisional guide some investigation or accepted as highly established facts.
- Quite often a research hypothesis is a prediction of being tested by scientific methods, that relates a variable to some dependent variable

Characteristics of hypothesis:

- (i) Hypothesis should be clear and precise. If the hypothesis is clear and precise, the inferences drawn on its basis will be reliable.
- (ii) Hypothesis should be capable of being tested.
- (iii) Hypothesis should state relationship between variables. It happens to be a relational hypothesis.
- (iv) Hypothesis should be limited in scope. The researcher must remember that narrower hypothesis is more testable and he should develop such hypothesis.

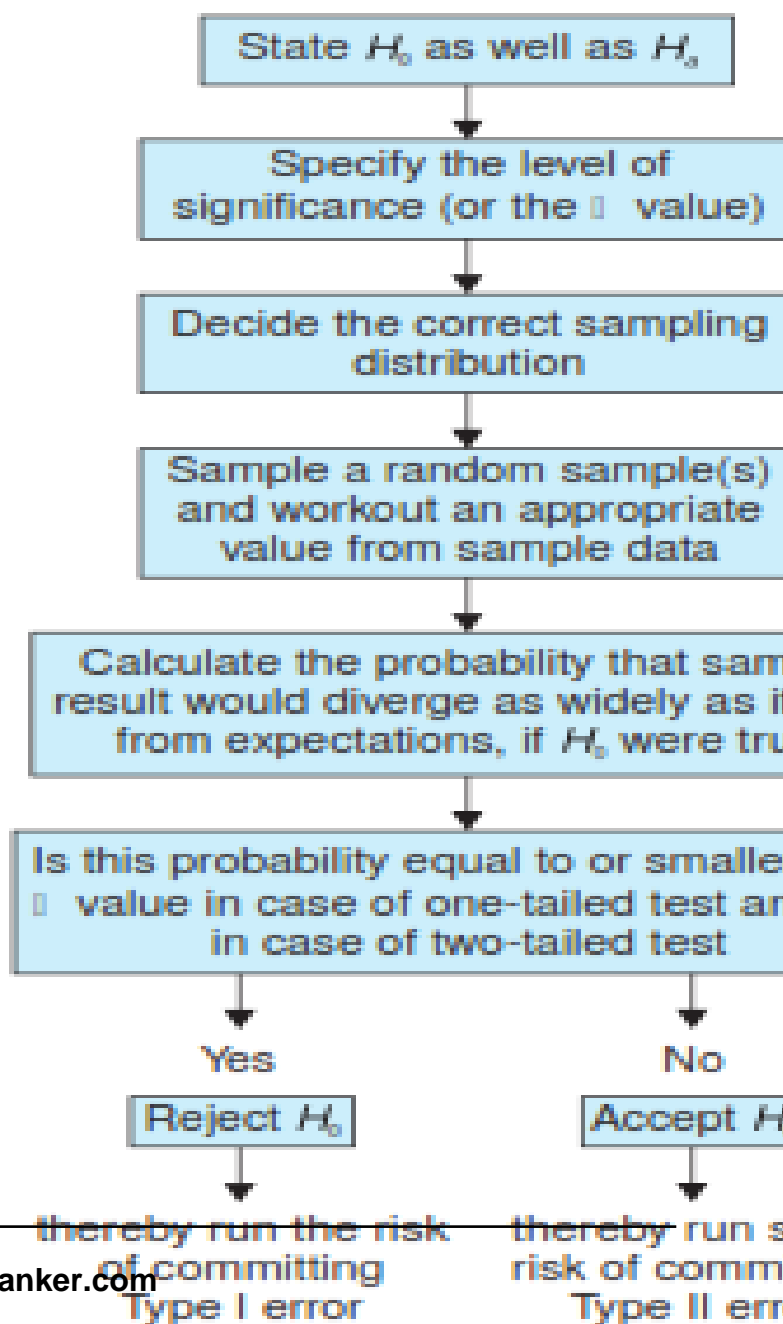
Characteristics of hypothesis:

(v) Hypothesis should be stated as far as possible so that the same is easily understandable by all. One must remember that simplicity of hypothesis is of great significance.

(vi) Hypothesis should be consistent with most known facts. It should be consistent with a substantial body of established facts. In other words, it should be one which judges accept as true.

PROCEDURE FOR HYPOTHESIS

FLOW DIAGRAM FOR HYPOTHESIS TESTING



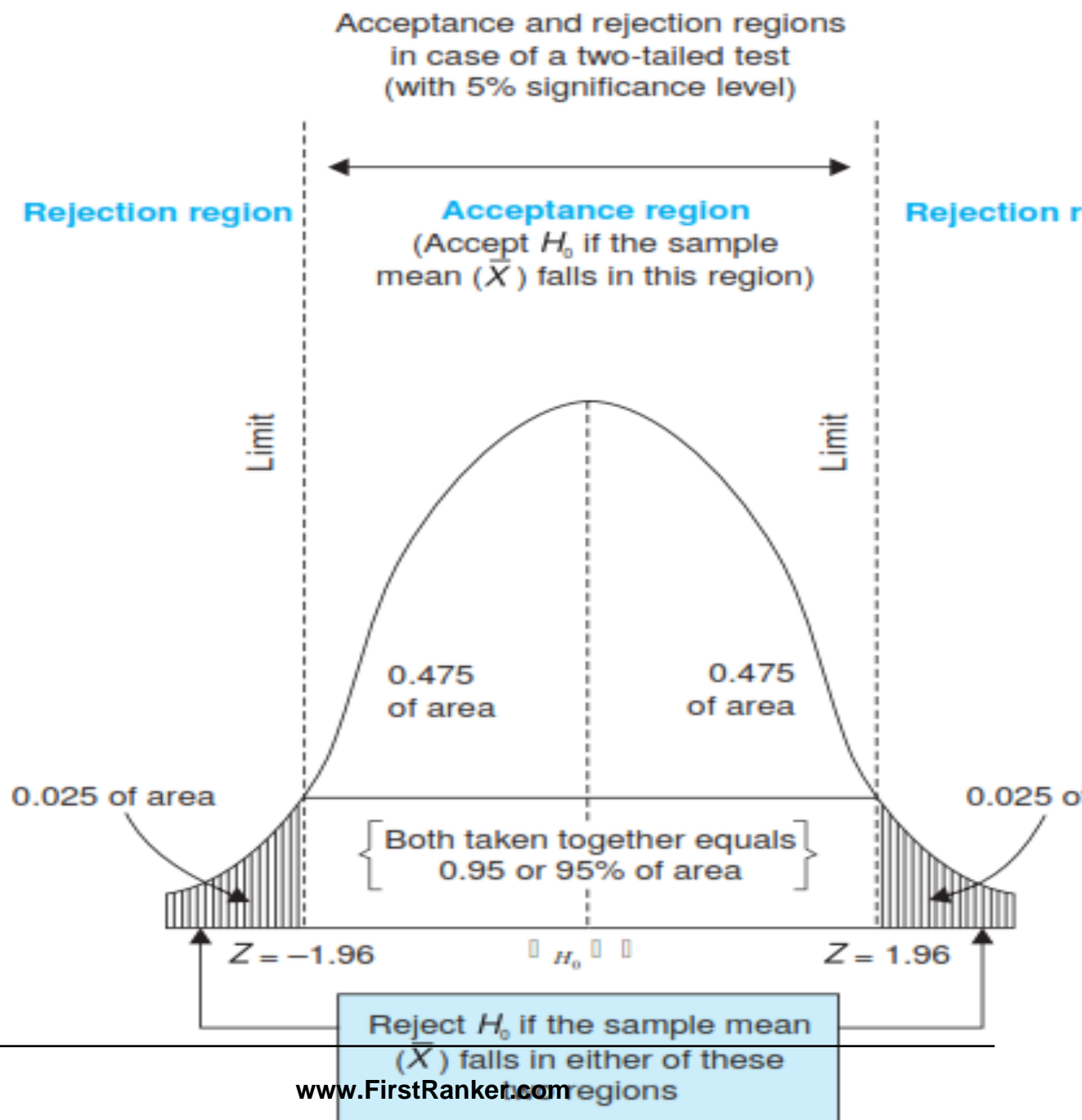
(i) State H_0 and H_1 :

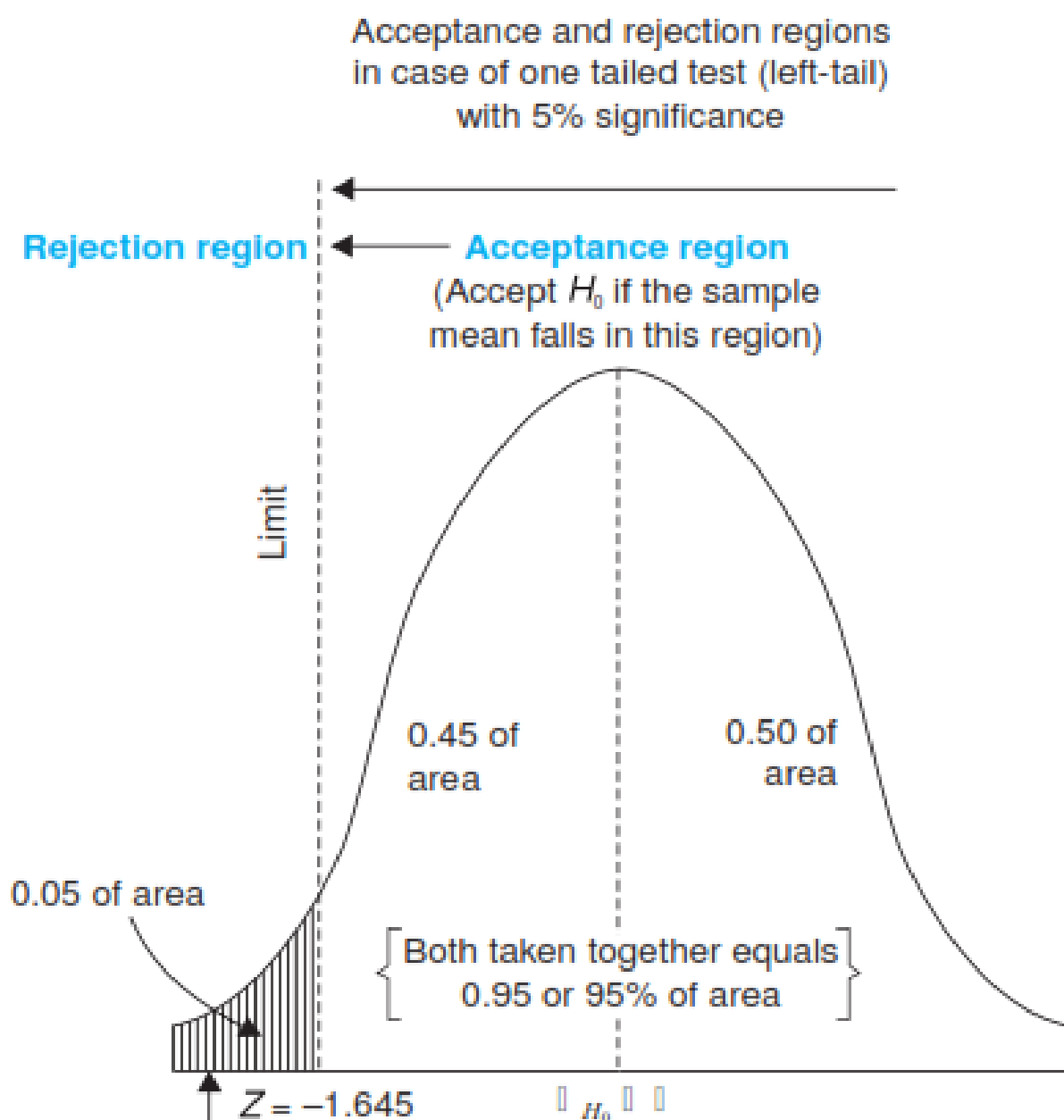
Null hypothesis $H_0 : \mu = 10$ tons

Alternative Hypothesis $H_a : \mu > 10$ tons

(ii) Selecting a Significance level

- The hypotheses are tested on a pre-determined level of significance and as such the same should be specified.
- Generally, in practice, **either 5% level or 1% level** is used for the purpose.
- The 5 per cent level of significance means that we are willing to take as much as a 5 per cent risk of rejecting the null hypothesis (H_0) happens to be true.





Reject H_0 if the sample mean
(\bar{X}) falls in this region

(iii) Deciding the distribution t

After deciding the level of significance, the testing is to determine the appropriate sampling

The choice generally remains between normal t-distribution.

(iv) Selecting a random sample
computing an appropriate value

- Another step is to select a random sample
appropriate value from the sample data computed
utilizing the relevant distribution.
- In other words, draw a sample to furnish empirical

(v) Calculation of the probability

One has then to calculate the probability that
diverge as widely as it has from expectation
were in fact true.

(vi) Comparing the probability

- Yet another step consists in comparing the p with the specified value for α , the significant
- If the calculated probability is equal to or smaller than α (in the case of one-tailed test, then reject the null hypothesis in favor of the alternative hypothesis),
- but if the calculated probability is greater, then do not reject the null hypothesis.

Errors in hypothesis

- Type 1 error
 - Hypothesis is rejected when it is true
- Type 2 error
 - Hypothesis is not rejected when it is false

H_0 (true)		
H_0 (false)		

Types of tests

- Parametric test
- Non-parametric test

PARAMETRIC TESTS

- z- test – for large samples
- t- test- for small samples
- f- test- for significance of difference variance.

z-test

- A type of statistical analysis that considers the

The mean of the variable in a sample set and

The mean of the variable in a larger population

Circumstance where the Z test

- A z-test is used for testing the mean of a population, comparing the means of two populations, with large samples, if you know the population standard deviation or normal distribution.
- It is also used for testing the proportion of a population against a standard proportion, or comparing the proportions of two populations.

Example: Comparing the average engineering salaries of two companies.

Example: Comparing the fraction defectives from two different machines.

T-test

- A **T-test** is a statistical examination of two po
- A two-sample **t-test** examines

whether two samples are different and is com

- when the variances of two normal distributio
- when an experiment uses a small sample size

t-test – when to use

- A t-test is used for testing the mean of one population or comparing the means of two populations if you do not know the standard deviation and when you have a limited sample size.
- If you know the populations' standard deviation, you use a z-test.

Example: Measuring the average diameter of shafts from a factory if you have a small sample.

F-test

- **Definition:** **F-test** is a statistical **test** that is used to compare two populations having normal distribution having equal standard deviation. This is an important part of ANOVA.

WHEN?

- An F-test is used to compare 2 populations' variances of any size. It is the basis of ANOVA.

Example: Comparing the variability of bolt diameters

Non Parametric te

U test

- (also called the **Mann–Whitney–Wilcoxon (Mann–Whitney) sum test**, or **Wilcoxon–Mann–Whitney test**)
- is a nonparametric test of the null hypothesis that two samples are from the same population against an alternative hypothesis, especially that a particular population tends to have larger values than the other.

K-W test

- is a non-parametric method for testing whether same distribution.
- It is used for comparing two or more independent different sample sizes. It extends the Mann–Whitney more than two groups.
- The parametric equivalent of the Kruskal-Wallis test of variance (ANOVA).

Bivariate Analysis

Introduction to bivariate analysis

- When one measurement is made on each observation, **bivariate analysis** is applied.

If more than one measurement is made on each observation, **multivariate analysis** is applied.

In this section, we focus on **bivariate analysis**. In this case, two measurements are made on each observation.

The two measurements will be called X and Y . If two measurements are obtained for each observation, the data is the pair (X, Y) .

- Bivariate data can be stored in a table with

	X	Y
Obs. 1	2	1
Obs. 2	4	4
Obs. 3	3	1
Obs. 4	7	5
Obs. 5	5	6
Obs. 6	2	1
Obs. 7	4	4
Obs. 8	3	1
Obs. 9	7	5
Obs. 10	5	6

- Some examples:
 - Height (X) and weight (Y) are measured for each individual in a sample.
 - Stock market valuation (X) and quarterly earnings (Y) are recorded for each company.
 - A cell culture is treated with varying doses of a drug, and the growth rate (X) and cell count (Y) are recorded for each trial.
 - Temperature (X) and precipitation (Y) are recorded for each day at a set of weather stations.

CHI SQUARE T

INTRODUCTION

- The chi-square test is an important test among tests of significance developed by statisticians.
- It was developed by Karl Pearson in 1900.
- CHI SQUARE TEST is a non parametric test that does not require any assumption or distribution of any variable.
- This statistical test follows a specific distribution called the chi-square distribution.
- In general, the test we use to measure the difference between what is observed and what is expected according to the assumed hypothesis is called the **chi-square** test.

IMPORTANT CHARACTERISTICS OF CHI-SQUARE TEST

- This test (as a non-parametric test) is based on frequencies and not on the parameters standard deviation.
- The test is used for testing the hypothesis and is useful for estimation.
- This test can also be applied to a contingency table with several classes and as such it is a test in research work.
- This test is an important non-parametric test. assumptions are necessary in regard to population, no need of parameter values and less mathematical details are involved.

APPLICATIONS OF A CHI SQ

This test can be used in

- 1) Goodness of fit of distributions**
- 2) test of independence of attributes**
- 3) test of homogeneity.**

1) TEST OF GOODNESS OF FIT OF DISTRIBUTION

➤ This test enables us to see how well does the theoretical distribution (such as Binomial distribution, Poisson distribution or Normal distribution) fit the observed data.

➤ The χ^2 test formula for goodness of fit is:

$$\chi^2 = \sum \frac{(o - e)^2}{e}$$

Where,

o = observed frequency

e = expected frequency

➤ If χ^2 (calculated) > χ^2 (tabulated), with (n-1) degrees of freedom, the null hypothesis is rejected otherwise accepted.

➤ And if null hypothesis is accepted, then it implies that the given distribution follows the theoretical distribution.

2) TEST OF INDEPENDENCE OF ATTRIBUTES

- Test enables us to explain whether or not two attributes are associated.
- For instance, we may be interested in knowing whether new medicine is effective in controlling fever or not. If the new medicine is useful.
- In such a situation, we proceed with the null hypothesis that the two attributes (viz., new medicine and controlling fever) are independent which means that new medicine is not effective in controlling fever.
- $\chi^2(\text{calculated}) > \chi^2(\text{tabulated})$ at a certain level of significance for given degrees of freedom, the null hypothesis is rejected, i.e. two variables are dependent. (i.e., new medicine is effective in controlling the fever) and $\chi^2(\text{calculated}) < \chi^2(\text{tabulated})$, the null hypothesis is not rejected, i.e. 2 variables are independent. (i.e., the new medicine is not effective in controlling the fever).

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- when null hypothesis is rejected, it can be concluded that there is a significant association between two attributes

3) TEST OF HOMOGENITY

- This test can also be used to test whether events follow uniformity or not e.g. the admission of patients in government hospital in all days of the week uniform or not can be tested with the help of χ^2 test.
- χ^2 (calculated) < χ^2 (tabulated), then null hypothesis is accepted, and it can be concluded that there is no difference in the occurrence of the events. (uniformity of admission of patients through out the week)

CONDITIONS FOR THE APPLIED TEST

The following conditions should be satisfied and be applied:

- 1) The data must be in the form of frequencies
- 2) The frequency data must have a precise number of groups
must be organised into categories or groups
- 3) Observations recorded and used are collected on a single
basis.
- 4) All the items in the sample must be independent
- 5) No group should contain very few items, (in the worst case where the frequencies are less than 5, this can be done by combining the frequencies of adjacent groups so that the new frequencies become greater than 5. (Most statisticians take this number as 5, but 10 is often better by most of the statisticians.)
- 6) The overall number of items must also be large enough
It should normally be at least 50.

Multivariate analysis

- Multivariate analysis is essentially the simultaneous analysing multiple independent variables with multiple dependent (outcomes) using matrix algebra (most multivariate analysis)

Purpose.

- Behaviours, emotions, cognitions, and affect are described in terms of one or two variables.
- Furthermore, these traits cannot be measured directly, at speed, but must be inferred from constructs that are measured by multiple factors or variables.

- Importance is usually based upon how much variance can be extracted from the data.
- Variance is a numerical representation of the (behaviour, emotion, cognition, etc.) in the p
- We assume it represents how much of that t individual.
- If two variables are associated or correlated they share some common underlying trait/fac equality in how they vary on the scores in th
- That underlying trait is causing them to co-var

Why the multivariate approach

With univariate analyses we have just one dependent variable

Although any analysis of data involving more than one dependent variable can be called
as 'multivariate', we typically reserve the term for analyses involving
multiple independent variables

So MV analysis is an extension of UV ones, or conversely, UV
analyses are special cases of MV ones

Multivariate Pros and Cons Summary

Advantages of using a multivariate statistic

- Richer realistic design
- Looks at phenomena in an overarching way (provides a more holistic analysis)
- Each method differs in amount or type of Independent Variables (IVs) and Dependent Variables (DVs)
- **Can help control for Type I Error**

Disadvantages

- Larger Ns are often required
- More difficult to interpret
- Less known about the robustness of assumptions

Introduction

- ANOVA is an abbreviation for the method: Analysis Of Variance
Invented by R.A. Fisher
- ANOVA is used to test the significant difference between more than two groups and to make inferences about where the groups are drawn from population having

Theoretical Example: Sections At IMT

Sec: A



Sec: B



GK

CGPA

Indecipline

Continued..

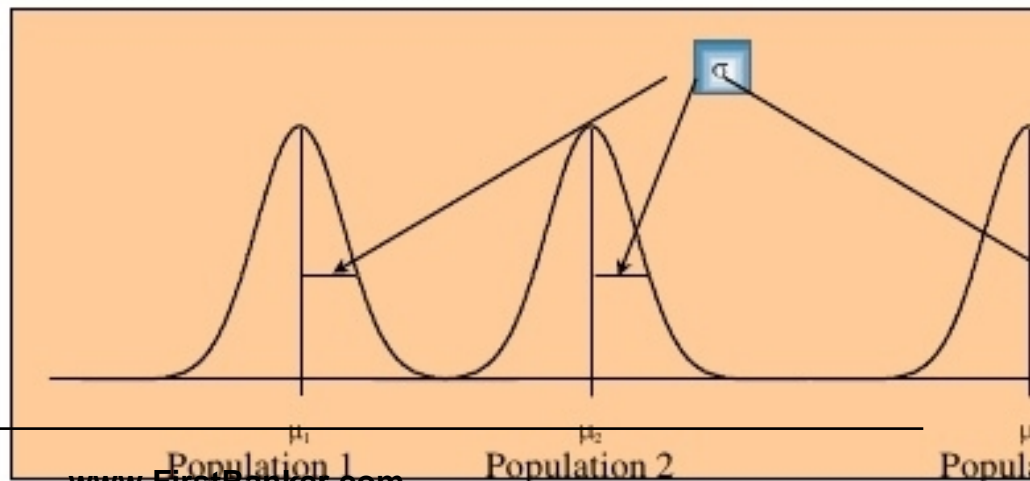
- ANOVA is comparison of means. Each value of a factor or combination of treatment.
- The ANOVA is a powerful and common procedure in the social sciences. It covers a variety of situations.

Why ANOVA instead of t-tests?

- If you are comparing means between two groups, why not just do several t-tests to compare the mean from one group to the mean from each of the other groups?
 - Before ANOVA, this was the only option to compare means between more than two groups.
- The problem with the multiple t-tests is that as the number of groups increases, the number of two sample t-tests also increases.
- As the number of tests increases the probability of making a Type I error also increases.

Analysis of Variance : Assu

- We assume *independent random sampling* from the r populations
- We assume that the r populations under consideration
 - are *normally distributed*,
 - with means μ_i that may or may not be equal,
 - but with *equal variances*, σ_i^2 .



ANOVA Hypoth

- The Null hypothesis for ANOVA means for all groups are equal:

$$H_0 : \mu_1 = \mu_2 = \mu_3 = \dots =$$

- The Alternative hypothesis for *at least two* of the means are n
- The test statistic for ANOVA is F-statistic.

One & N way ANOVA

- One way ANOVA
Analysis of variance, so named because it can consider only one independent variable at a time.
- N way ANOVA
As its name suggests, this is a procedure that allows you to examine the effects of multiple independent variables concurrently.

