

Module 5

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 preliminary guide to some investigation or accepted as highly established facts.
- Quite often a research hypothesis is a prediction of being tested by scientific methods, that an independent 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 a

(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 a

Characteristics of hypothesis:

- (vii) Hypothesis should be amenable to test in reasonable time.

One should not use even an excellent hypothesis if it cannot be tested in reasonable time for one cannot spend too much time and data to test it.

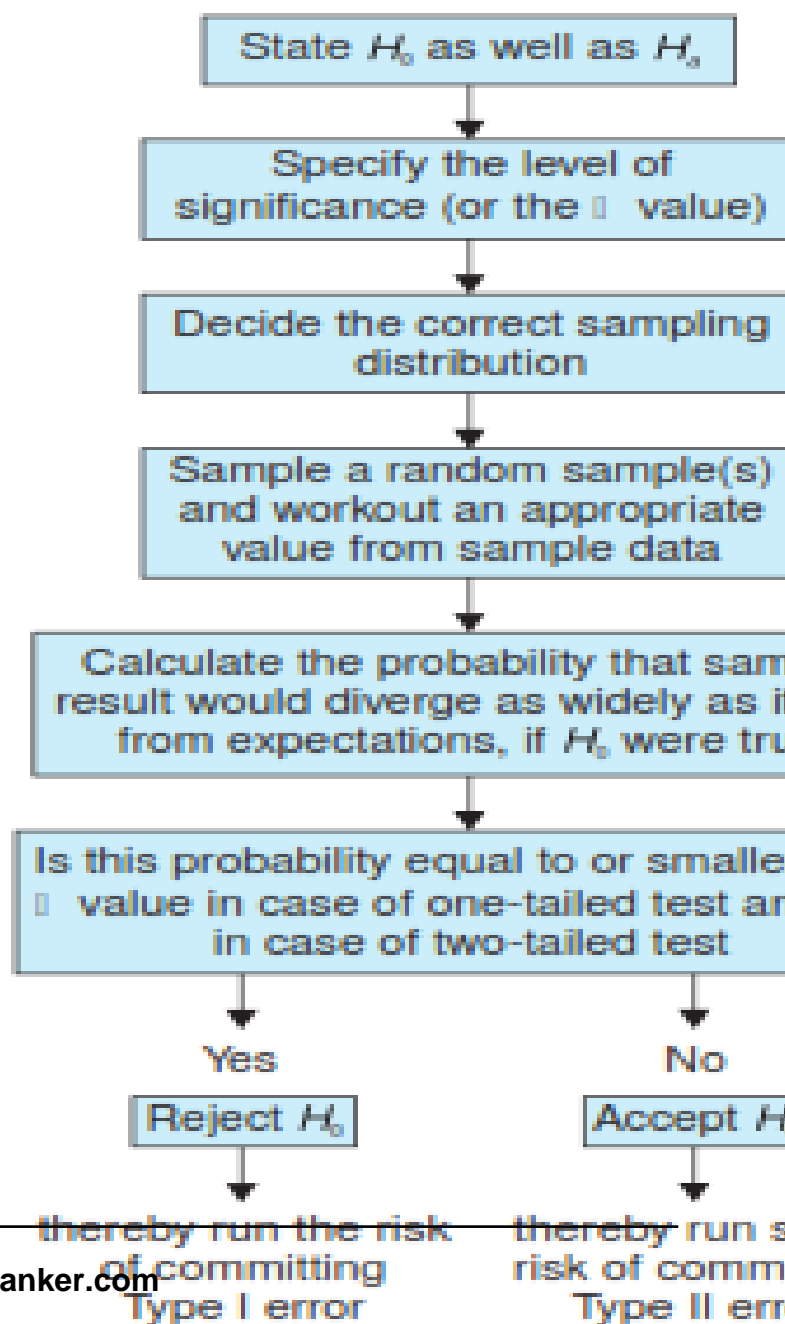
Characteristics of hypothesis:

(viii) Hypothesis must explain the facts that g
explanation.

- Thus hypothesis must actually explain what i
should have empirical reference

PROCEDURE FOR HYPOTHESIS

FLOW DIAGRAM FOR HYPOTHESIS TESTING



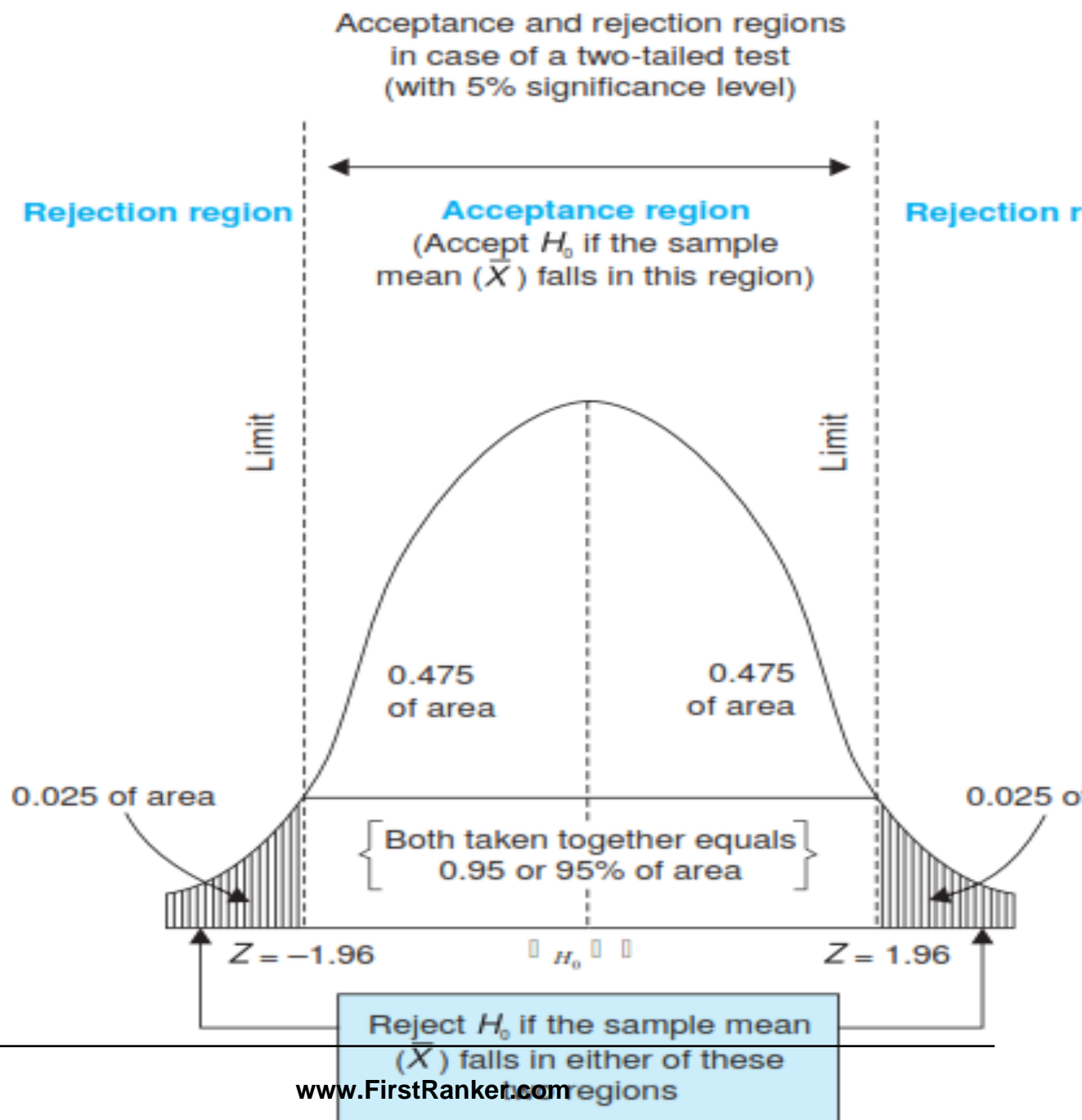
(i) State H_0 and H_1 :

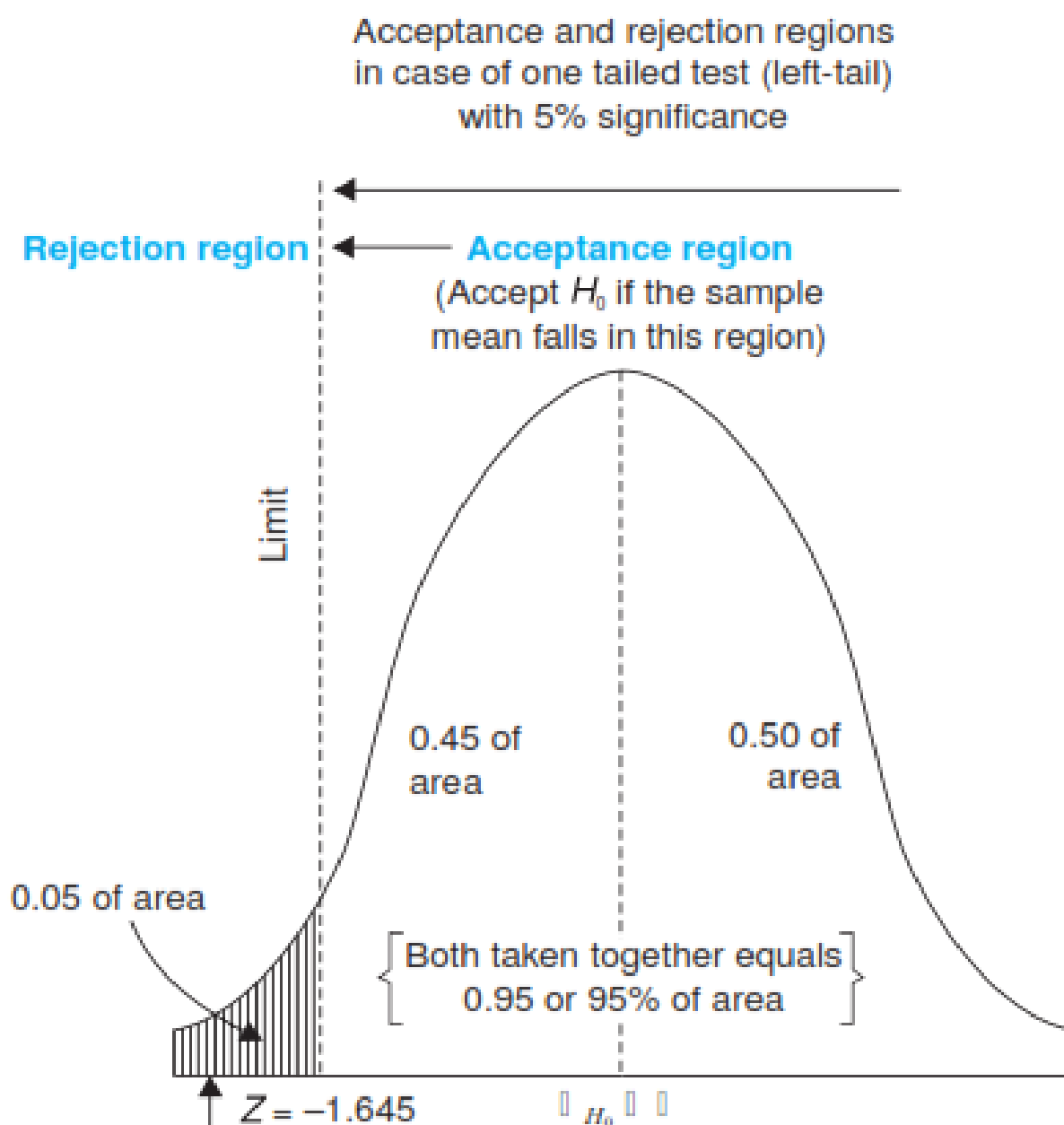
Null hypothesis $H_0 : \mu = 10$ tons

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 sampli

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

- Comparing the probability thus calculated with the significance level.
- If the calculated probability is equal to or smaller than the significance level (for a two-tailed test, or half of one-tailed test, then reject the null hypothesis in favor of the alternative hypothesis),
- but if the calculated probability is greater, then

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, or 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 populations
- A two-sample **t-test** examines

whether two samples are different and is commonly used in two cases:

- when the variances of two normal distributions are unequal
- 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.

1. Testing difference between samples (independent Sample

Alternative Hypothesis: Nourishment programme I programme 'B' increase the children's weight significantly.

Solution:

X	Nourishment programme A		
	$x - \bar{x}$ = (x-46)	$(x - \bar{x})^2$	y
44	-2	4	42
37	-9	81	42
48	2	4	58
60	14	196	64
41	-5	25	64
			67
			62
230	0	310	39

$$t = \frac{\bar{x} - \bar{y}}{\sqrt{s^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

$$\sum x = 230$$

$$\sum y =$$

$$\sum (x - \bar{x})^2 = 310$$

$$\sum (y - \bar{y})^2 =$$

$$\bar{x} = \frac{\sum x}{n_1} = \frac{230}{5} = 46$$

$$\bar{y} = \frac{\sum y}{n_2} = \frac{399}{7} = 57$$

$$s^2 = \frac{1}{n_1 + n_2 - 2} \left\{ \sum (x - \bar{x})^2 \right\}$$

$$\text{D.F} = (n_1 + n_2 - 2) = (5 + 7 - 2)$$

$$s^2 = \frac{1}{10} \{310 + 674\} = 98.4$$

$$t = \frac{\bar{x} - \bar{y}}{\sqrt{s^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

When is a one-tailed test appropriate?

- If you are using a significance level of 0.05, a portion of your alpha to testing the statistical significance of interest.
- This means that 0.05 is in one tail of the distribution statistic.
- When using a one-tailed test, you are testing a relationship in **one direction** and completely ignoring the possibility of a relationship in the other direction.

- For example, imagine that you have developed a
- It is cheaper than the existing drug and, you believe
- In testing this drug, you are only interested in testing whether it is more effective than the existing drug.
- You do not care if it is significantly **more effective**
- You only wish to show that it is not less effective than the existing drug. A one-tailed test would be appropriate.

- Our null hypothesis is that the mean is equal to x
- A one-tailed test will test either if the mean is greater than x or less than x

Or

- if the mean is significantly less than x , then the null hypothesis is rejected

Two-tailed test

- If you are using a significance level of 0.05, a of your alpha to testing the statistical significance half of your alpha to testing statistical significance in both directions.
- This means that .025 is in each tail of the distribution.

- Our null hypothesis is that the mean is equal
- A two-tailed test will test both if the mean is
than x and if the mean significantly less than
- The mean is considered significantly different
statistic is in the top 2.5% or bottom 2.5% of
distribution, resulting in a p-value less than 0

2. To test Significance of the n Random sample

$$t = \frac{(\bar{X} - \mu) \sqrt{n}}{S}$$

\bar{X} = the mean of the sample

μ = the actual or hypothetical mean

n = the sample size

S = the standard deviation of the sample

$$S = \sqrt{\frac{\sum (X - \bar{X})^2}{n - 1}} \quad \text{or} \quad S = \sqrt{\frac{\sum d^2 - \frac{(\sum d)^2}{n}}{n - 1}}$$

$$= \sqrt{\frac{1}{(n - 1)} \left[\sum d^2 - \frac{(\sum d)^2}{n} \right]}$$

d = deviation from the assumed mean

Fiducial limits of Population Mean

- Assuming that the sample is a random sample from a population of unknown mean the 95% fiducial limits of population mean (μ) are

Handwritten text in red ink on a white background:

$$\bar{X} \pm \frac{S}{\sqrt{n}} \text{ to } 0.01$$

99% limits are.

$$\bar{x} \pm \frac{s}{\sqrt{n}} \text{ to } 0.01$$

Example

The manufacturer of a certain make of electric bulbs have a **mean life of 25** months with a standard deviation of 5 months. A random sample of 6 such bulbs gave the following life in months:

Life of months : 24 26 30 20 20 18

Can you regard the producer's claim to be valid at 5% significance?

CALCULATION OF \bar{X} and

X	$(X - \bar{X})$ X
24	+1
26	+3
30	+7
20	-3
20	-3
18	-5
$\Sigma X = 138$	

$$\bar{X} = \frac{\sum X}{n} = \frac{138}{6} = 23$$

$$S = \sqrt{\frac{\sum x^2}{n-1}} = \sqrt{\frac{102}{5}} = \sqrt{20.4}$$

$$= \frac{|23 - 25|}{4.517} \sqrt{6} = \frac{2 \times 2.449}{4.517}$$

$$v = n - 1 = 6 - 1 = 5. \text{ For } v =$$

3. Testing difference between two samples (Dependent Samples or observations)

Two samples are said to be dependent when the observations in one sample are related to those in the other in any pair on the same subject.

Two samples may consist of pairs of observations from the same selected population.

Say, for effectiveness of coaching class or training

3. Testing difference between me (Dependent Samples or matched

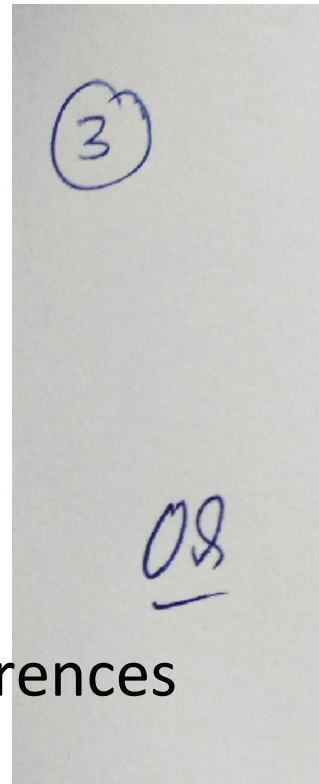
It is defined by

\bar{d} = the mean of the differences

S = the standard deviation of the differences

If $t >$ table value then H_0 is rejected

If $t <$ table value then H_0 is accepted



- The value of the **S** is calculated by,

$$S =$$

or

$$S =$$

- It should be noted that **t** is based on ***n-1*** deg

Example

- To verify whether a course in accounting improved the performance of students, a similar test was given to 12 participants both before and after the course. The original marks recorded in alphabetical order of participants were- 44,40,61,52,32,44,70,41,60,55,58,65 and 72. After the course, the marks were in the same order as before but the marks were 53,38,69,57,46,39,73,48,73,74,60 and 78. was the course effective?

Hypothesis :

there is no difference in the marks obtained in the course, i.e. the course has not been useful

<i>Participants</i>	<i>Before (1st Test)</i>	<i>After (2nd Test)</i>
A	44	53
B	40	38
C	61	69
D	52	57
E	32	46
F	44	39
G	70	73
H	41	48
I	67	73
J	72	74
K	53	60
L	72	78

$$\bar{d} = \frac{\sum d}{n} = \frac{60}{12} = 5$$

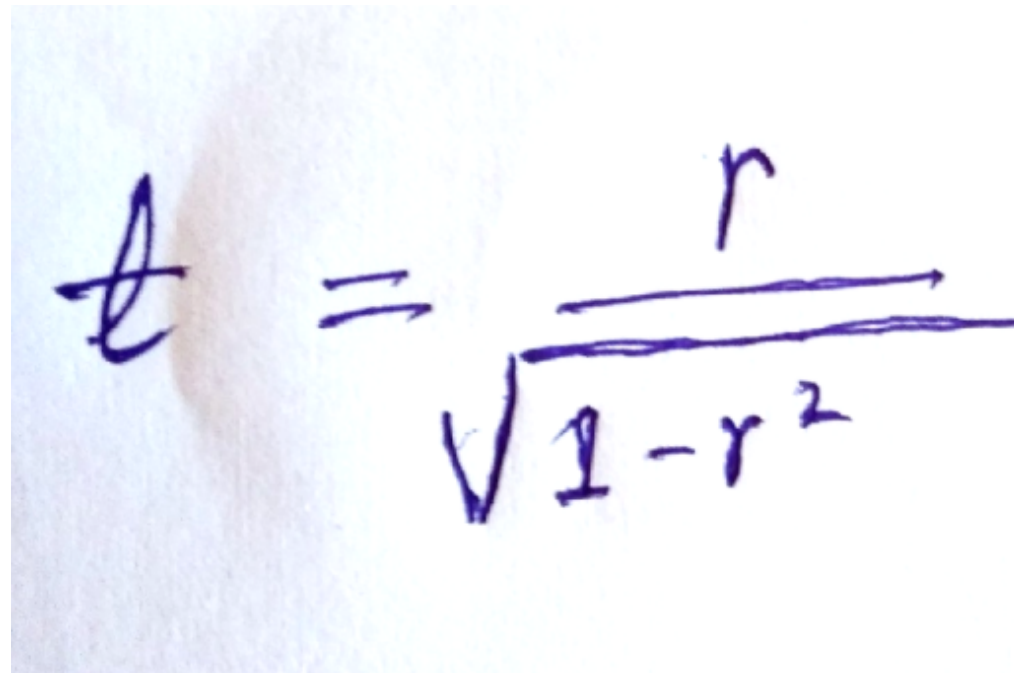
$$S = \sqrt{\frac{\sum d^2 - n(\bar{d})^2}{n-1}} = \sqrt{\frac{578 - 12 \times 5^2}{12-1}}$$

$$t = \frac{5 \times \sqrt{12}}{5.03} = \frac{5 \times 3.464}{5.03} = 3.443$$

$$v = n - 1 = 12 - 1 = 11; \text{ For } v = 1$$

4. Testing the Significance of a Correlation Coefficient

- Given random sample from bivariate population
- If we are to test the hypothesis that the correlation coefficient in the population is zero.
- i.e. the variables in the population are uncorrelated.



A photograph of a piece of white paper with a handwritten formula in blue ink. The formula is $t = \frac{r}{\sqrt{1-r^2}}$. The letter 't' is on the left, followed by an equals sign. To the right of the equals sign is a fraction where the numerator is 'r' and the denominator is the square root of '1-r^2'.

- Here t is based on $n-2$ degree of freedom.
- We say that the value of r is significant at 5% level.
- If $t < t_{0.05}$ the data are considered with the hypothesis of being uncorrelated.
- r = correlation coefficient

Example 1

- A random sample of 27 pairs of observations from a normal population gives a correlation coefficient of 0.5. Is it likely that the variables in the population are uncorrelated?
- **Solution**
 - H_0 : there is no significant difference in the sample correlation and correlation in the population

Example 2

- How many pairs of observations must be included in order that an observed correlation coefficient has a calculated value of r greater than 2.72?

Solution-

Here, r value is 0.42, we have to find out n

Example 3

- The following table gives the ages in years of husbands and wives at marriage. Compute the correlation coefficient and test its significance.
- Husband's Age : 23 27 28 29 30
 36 39
- Wife's Age : 18 22 23 24 25 26
 32

- Solution
- H_0 : there is no correlation in the population
- Here you need to find out

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

- The F-Test is named in honor of the great statistician Sir Ronald A. Fisher.
- The object of the ***F-test*** is to find out whether the two estimates of population variance differ significantly. If the two samples may be regarded as drawn from a normal distribution having the same variance. For carrying out the test, we calculate the ratio F .

$$F = \frac{S_1^2}{S_2^2}, \text{ where } S_1^2 = \frac{\Sigma (X_1 - \bar{X}_1)^2}{n_1}$$

and

$$S_2^2 = \frac{\Sigma (X_2 - \bar{X}_2)^2}{n_2}$$

$S_1^2 > S_2^2$ means S_1 is always larger estimate of

V_1 = Degree of Freedom for samples having larger variance

V_2 = Degree of Freedom for samples having Smaller variance

$$V_1 = n_1 - 1$$

If $F > \text{table value}$ then F ratio is considered as significant
 H_1 is accepted

If $F < \text{table value}$ accept H_0 , means both the sample
population having same variance.

A X_1	$(X_1 - \bar{X}_1)$ x_1	x_1^2	B X_2
66	-14	196	64
67	-13	169	66
75	-5	25	74
76	-4	16	78
82	+2	4	82
84	+4	16	85
88	+8	64	87
90	+10	100	92
92	+12	144	93
			95
			97
$\Sigma X_1 = 720$	$\Sigma x_1 = 0$	$\Sigma x_1^2 = 734$	$\Sigma X_2 = 913$

H_0 = Two Populations Have Same Variance

$$V1 = 10$$

$$V2 = 8$$

$$F_{0.05} = 3.36$$

CALCULATED VALUE OF $F = 0.707$

The Calculated F Value Is Less Than Table Value HENCE THE HYPOTHE

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.

Mann–Whitney- U test

- This test is to determine whether two independent samples have been drawn from the same population.
- This test applies in very general conditions and the populations sampled are continuous.

Mann–Whitney-*U* test Steps t

- We first of all rank the data jointly, taking the single sample in either an increasing or decreasing magnitude.
- We usually adopt low to high ranking process rank 1 to an item with lowest value, rank 2 to so on.
- In case there are ties, then we would assign observation the mean of the ranks which the
 - (for 11 11 11 rank may be $(6+7+8)/3=7$)

Mann–Whitney-*U* test Steps t

- After this we find the sum of the ranks assigned to the first sample (R_1) and also the sum of the ranks assigned to the second sample (R_2)
- Then we work out the test statistic i. e. U which is the difference between the ranked observations under

$$U = n_1 \times n_2 + \frac{n_1 (n_1 + 1)}{2} - R_1$$

n_1 and n_2 are the sample sizes and R_1 is the sum of the values of the first Sample

Mann–Whitney- U test Steps t

- H_0 : two samples are from identical population
- If the null hypothesis that the $n_1 + n_2$ observations are from an identical population is true, the said U statistic follows a distribution with

$$\text{Mean} = \mu = \frac{n_1 n_2}{2}$$

$$\text{\& SD } \sigma = \frac{\sqrt{n_1 n_2 (n_1 + n_2 + 1)}}{12}$$

- If n_1 and n_2 are sufficiently large (i.e. both n_1 and n_2 are sufficiently large) the sampling distribution of U can be approximated by a normal distribution and the limits of the acceptance determined in the usual way at a given level

Example 1

- The value in one sample are 53 38 69 57 46 3
In another sample they are 44 40 61 52 32 4
test at the 10% level the hypothesis that the
with the same mean. Apply *U-test*

Solution

size of sample item in ascending order	Rank	Re
32	1	B
38	2	A
39	3	A
40	4	B
41	5	B
44	6.5	B
44	6.5	B

- $R1 = 2+3+8+9+11.5+13+14+17+21.5+21.5+23$
- $R2 = 1+4+5+6.5+6.5+10+11.5+15+16+18+19.5$
- $N1 = 12$
- $N2 = 12$
- $U = 54.5$

- Since in the given problem n_1 and n_2 both are sampling distribution of U approximately clo
- Keeping this in view, we work out the mean hypothesis that the two samples come from under

$$\mu = 72$$

$$\sigma = 17.32$$

- As the alternative hypothesis is that the mean of two populations are not equal, a two-tailed test is used. The limits of acceptance region, keeping in view the level of significance as given, can be worked out as under
- As Z value for 0.45 of the area under the normal curve is 1.64, the following limits of acceptance region
- Upper limit = $\mu + 1.64 \sigma_U = 100.40$
- Lower limit = $\mu - 1.64 \sigma_U = 43.60$

The Kruskal-Wallis test (or H test)

- This test is conducted in a way similar to the
- This test is used to test the null hypothesis that random samples come from identical universes. The alternative hypothesis is that the means of the groups are not equal.
- This test is analogous to the one-way analysis of variance. Unlike the latter it does not require the assumption of normality, that the data are from approximately normal populations or that they have the same standard deviation.

The Kruskal-Wallis test (or H test)

- In this test, like the U test, the data are ranked high or high to low as if they constituted a single population.
- The test statistic is H for this test which is used to test the null hypothesis that the populations are identical.

$$H = \frac{12}{n(n+1)} \sum_{i=1}^k \frac{R_i^2}{n_i} - 3(n+1)$$

where $n = n_1 + n_2 + \dots + n_k$ and R_i being the sum of observations in the i th sample.

The Kruskal-Wallis test (or H test)

- If the null hypothesis is true that there is no difference between sample means and each sample has at least five observations, the sampling distribution of H can be approximated by a chi-square distribution with $(k - 1)$ degrees of freedom.
- As such we can reject the null hypothesis at a significance level α if H value calculated, as stated above, exceeds the critical value of chi-square.
- * If any of the given samples has less than five observations, the chi-square distribution approximation can not be used and the test is based on table

The Kruskal-Wallis test (or H test)

Illustration

- Use the Kruskal-Wallis test at 5% level of significance to test the hypothesis that a professional bowler performs equally well with four bowling balls, given the following results

Bowling Results in Five Games

With Ball No. <i>A</i>	271	282
With Ball No. <i>B</i>	252	275
With Ball No. <i>C</i>	260	255
With Ball No. <i>D</i>	279	242

<i>Bowling results</i>	<i>Rank</i>
302	1
297	2
282	3
279	4
276	5
275	6
271	7
270	8
268	9
266	10
262	11
260	12
258	13
257	14
255	15
252	16
248	17
246	18
242	19
239	20

Table 12.7 (a): Bowling Results with Different Balls a

<i>Ball A</i>	<i>Rank</i>	<i>Ball B</i>	<i>Rank</i>	<i>Ball C</i>	<i>Rank</i>
271	7	252	16	260	15
282	3	275	6	255	12
257	14	302	1	239	18
248	17	268	9	246	19
262	11	276	5	266	14
$n_1 = 5$	$R_1 = 52$	$n_2 = 5$	$R_2 = 37$	$n_3 = 5$	$R_3 = 78$

Now we calculate H statistic as under:

$$H = \frac{12}{n(n+1)} \sum_{i=1}^k \frac{R_i^2}{n_i} - 3(n+1)$$

$$= \frac{12}{20(20+1)} \left\{ \frac{52^2}{5} + \frac{37^2}{5} + \frac{75^2}{5} \right\}$$

$$= (0.02857) (2362.8) - 63 = 67.51$$

- As the four samples have five items* each, the value of H approximates closely with chi-square distribution.
- Now taking the null hypothesis that the bowler performs equally well with the four balls, we have the value of chi-square or $4 - 1 = 3$ degrees of freedom at 5% level of significance.
- Since the calculated value of H is only 4.51, which is less than the critical value of 7.815, so we accept the null hypothesis. This means the bowler performs equally well with the four balls.

Illustration:

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 observed data fit the theoretical distribution (such as Binomial distribution, Poisson distribution or Normal distribution).

➤ 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 is concluded 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).

-
- 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.
- $\chi^2(\text{calculated}) < \chi^2(\text{tabulated})$, 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 applied:

- 1) The data must be in the form of frequency
- 2) The frequency data must have a precise number 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 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. (statisticians take this number as 5, but 10 is better by most of the statisticians.)
- 6) The overall number of items must also be large. 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.

- Behaviors, emotions, cognitions, and attitudes are described in terms of one or two variables.
- Furthermore, these traits cannot be measured directly and 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 (behavior, emotion, cognition, etc.) in the po
- 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-va

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 more than two 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 statistical method: Analysis Of Variance
Invented by R.A. Fisher
- ANOVA is used to test the significance of the 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..

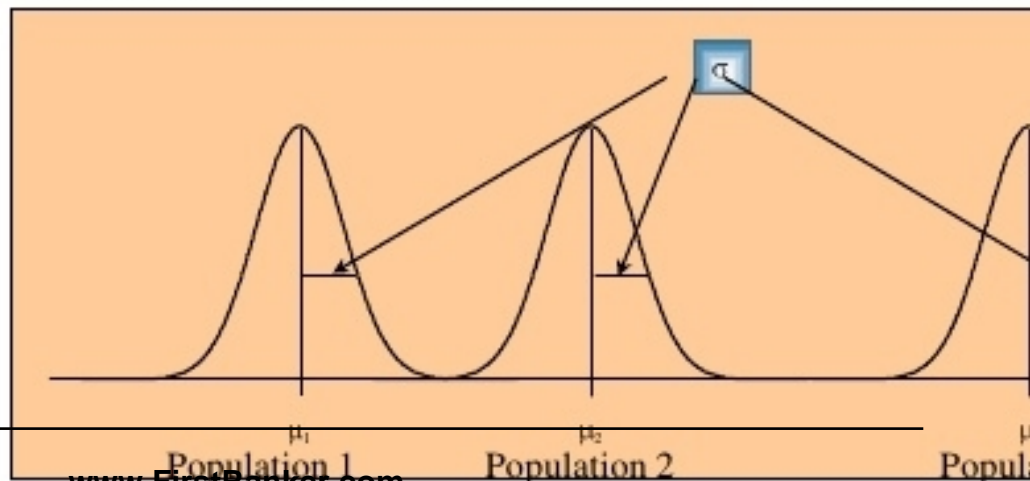
- ANOVA is comparison of means. Each value of a factor or combination of treatment.
- The ANOVA is a powerful and common procedure in the social sciences. It covers a variety of situations.

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 chance 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.

