

INFERENTIAL STATISTICS

Hypothesis Testing

Meaning of Hypothesis

A hypothesis is a tentative explanation for certain events, phenomena or behaviors. In statistical language, a hypothesis is a statement of prediction or relationship between or among variables. Plainly stated, a hypothesis is the most specific statement of a problem. It is a requirement that these variables are related. Furthermore, the hypothesis is testable which means that the relationship between the variables can be put into test on the data gathered about the variables.

Null and Alternative Hypotheses

There are two ways of stating a hypothesis. A hypothesis that is intended for statistical test is generally stated in the null form. Being the starting point of the testing process, it serves as our working hypothesis. A *null hypothesis (H_o)* expresses the idea of non-significance of difference or non- significance of relationship between the variables under study. It is so stated for the purpose of being accepted or rejected.

If the null hypothesis is rejected, the *alternative hypothesis (H_a)* is accepted. This is the researcher’s way of stating his *research hypothesis* in an operational manner. The research hypothesis is a statement of the expectation derived from the theory under study. If the related literature points to the findings that a certain technique of teaching for example, is effective, we have to assume the same prediction. This is our alternative hypothesis. We cannot do otherwise since there is no scientific basis for such prediction.

Chain of reasoning for inferential statistics

- 1. Sample(s) must be randomly selected
- 2. Sample estimate is compared to underlying distribution of the same size sampling distribution
- 3. Determine the probability that a sample estimate reflects the population parameter

The four possible outcomes in hypothesis testing		
	Actual Population Comparison	
	Null Hyp. True	Null Hyp. False
DECISION	(there is no difference)	(there is a difference)
Rejected Null Hyp	Type I error (alpha)	Correct Decision
Did not Reject Null	Correct Decision	Type II Error
	(Alpha = probability of making a Type I error)	

Regardless of whether statistical tests are conducted by hand or through statistical software, there is an implicit understanding that systematic steps are being followed to determine statistical significance. These general steps are described on the following page and include 1) assumptions, 2) stated hypothesis, 3) rejection criteria, 4)

computation of statistics, and 5) decision regarding the null hypothesis. The underlying logic is based on rejecting a statement of no difference or no association, called the null hypothesis. The null hypothesis is only rejected when we have evidence beyond a reasonable doubt that a true difference or association exists in the population(s) from which we drew our random sample(s).

Reasonable doubt is based on probability sampling distributions and can vary at the researcher's discretion. Alpha .05 is a common benchmark for reasonable doubt. At alpha .05 we know from the sampling distribution that a test statistic will only occur by random chance five times out of 100 (5% probability). Since a test statistic that results in an alpha of .05 could only occur by random chance 5% of the time, we assume that the test statistic resulted because there are true differences between the population parameters, not because we drew an extremely biased random sample.

When learning statistics we generally conduct statistical tests by hand. In these situations, we establish before the test is conducted what test statistic is needed (called the **critical value**) to claim statistical significance. So, if we know for a given sampling distribution that a test statistic of plus or minus 1.96 would only occur 5% of the time randomly, any test statistic that is 1.96 or greater in absolute value would be statistically significant. In an analysis where a test statistic was exactly 1.96, you would have a 5% chance of being wrong if you claimed statistical significance. If the test statistic was 3.00, statistical significance could also be claimed but the probability of being wrong would be much less (about .002 if using a 2-tailed test or two-tenths of one percent; 0.2%). Both .05 and .002 are known as alpha; the probability of a Type I error.

When conducting statistical tests with computer software, the exact probability of a Type I error is calculated. It is presented in several formats but is most commonly reported as "**p <**" or "**Sig.**" or "**Signif.**" or "**Significance.**" Using "p <" as an example, if a priori you established a threshold for statistical significance at alpha .05, any test statistic with significance at or less than .05 would be considered statistically significant and you would be required to reject the null hypothesis of no difference. The following table links p values with a benchmark alpha of .05:

P <	Alpha	Probability of Type I Error	Final Decision
.05	.05	5% chance difference is not significant	Statistically significant
.10	.05	10% chance difference is not significant	Not statistically significant
.01	.05	1% chance difference is not significant	Statistically significant
.96	.05	96% chance difference is not significant	Not statistically significant

Steps to Hypothesis Testing

Hypothesis testing is used to establish whether the differences exhibited by random samples can be inferred to the populations from which the samples originated.

General Assumptions

- Population is normally distributed
- Random sampling
- Mutually exclusive comparison samples
- Data characteristics match statistical technique

For interval / ratio data use ∞

t-tests, Pearson correlation, ANOVA, regression

For nominal / ordinal data use ∞

Difference of proportions, chi square and related measures of association

State the Hypothesis

Null Hypothesis (H_0): There is no difference between ___ and ___.

Alternative Hypothesis (H_a): There is a difference between ___ and ___.

Note: The alternative hypothesis will indicate whether a 1-tailed or a 2-tailed test is utilized to reject the null hypothesis.

H_a for 1-tail tested: The ___ of ___ is greater (or less) than the ___ of ___.

Set the Rejection Criteria

This determines how different the parameters and/or statistics must be before the null hypothesis can be rejected. This "region of rejection" is based on alpha (α) -- the error associated with the confidence level. The point of rejection is known as the critical value.

Compute the Test Statistic

The collected data are converted into standardized scores for comparison with the critical value.

Decide Results of Null Hypothesis

If the test statistic equals or exceeds the region of rejection bracketed by the critical value(s), the null hypothesis is rejected. In other words, the chance that the difference exhibited between the sample statistics is due to sampling error is remote--there is an actual difference in the population.

Let us consider an experiment involving two groups, an experimental group and a control group. The experimenter likes to test whether the treatment (values clarification lessons) will improve the self-concept of the experimental group. The same treatment is not given to the control group. It is presumed that any difference between the two groups after the treatment can be attributed to the experimental treatment with a certain degree of confidence.

The hypothesis for this experiment can be stated in various ways:

a) No existence or existence of a difference between groups

H_0 : There is no significant difference in self-concept between the group exposed to values clarification lessons and the group not exposed to the same.

H_a : The self-concept of the group exposed to values clarification lessons differ significantly that of the other group.

b) No existence or existence of an effect of the treatment

H_0 : There is no significant effect of the values clarification lessons on the self-concept of the students.

H_a : Values clarification lessons have no significant effect on the self-concept of students.

c) No existence of relationship between the variables

H₀: The self-concept of the students is not significantly related to the values clarification lessons conducted on them.

H_a: The self-concept of the students is not significantly related to the values clarification lessons they were exposed to.

Parametric Test

What are the parametric tests?

The parametric tests are tests that require normal distribution, the levels of measurement of which are expressed in an interval or ratio data. The following parametric tests are:

t-test for one population mean compared to a sample mean

t- test for Independent Samples

t- test for Correlated Samples

z- test, for Two Sample Means

z- test for One Sample Group

F- test (ANOVA)

R (Pearson Product Moment Coefficient of Correlation)

Y=a+bx (Simple Linear Regression Analysis)

$$Y = b_0 + b_1 x_1 + b_2 x_2 + \dots + b_n x_n$$

(Multiple Regression Analysis)

Comparing a Population Mean to a Sample Mean (T-test)

Example 1: Compare the mean age of incoming students to the known mean age for all previous incoming students. A random sample of 30 incoming college freshmen revealed the following statistics: mean age 19.5 years, standard deviation 1 year. The college database shows the mean age for previous incoming students was 18.

Solving by the Stepwise Method

I. Problem: Is there a significant difference between the mean age of past college students and the mean age of current incoming college students?

II. Hypothesis

H₀: There is no significant difference between the mean age of past college students and the mean age of current incoming college students.

H₀: $\bar{x}_1 = \bar{x}_2$

H_a: There is a significant difference between the mean age of past college students and the mean age of current incoming college students.

III. Level of Significance

Set the Rejection Criteria

Significance level .05 alpha, 2-tailed test
Degrees of Freedom = n-1 or 29
Critical value from t-distribution = 2.045

IV. Statistics

Compute the Test Statistic

$$t = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n-1}}}$$

$$t = \frac{19.5 - 18}{\frac{1}{\sqrt{30-1}}}$$

$$t = \frac{1.5}{\frac{1}{\sqrt{29}}}$$

$$t = \frac{1.5}{.186}$$

$$t = 8.065$$

V. Decision Rule: If the t- computed value is greater than or beyond the t-tabular / critical value, reject H_0 .

VI Conclusion: Given that the test statistic (8.065) exceeds the critical value (2.045), the null hypothesis is rejected in favor of the alternative. There is a statistically significant difference between the mean age of the current class of incoming students and the mean age of freshman students from past years. In other words, this year's freshman class is on average older than freshmen from prior years.

If the results had not been significant, the null hypothesis would not have been rejected. This would be interpreted as the following: There is insufficient evidence to conclude there is a statistically significant difference in the ages of current and past freshman students.

What is the t-test for independent samples?

The t-test is a test of difference between two independent groups. The means are being compared \bar{x}_1 against \bar{x}_2 .

When do we use the t-test for independent samples?

The t-test for independent samples is used when we compare means of two independent groups.

When the distribution is normally distributed,
Sk = 0 and Ku = .265.
we use interval or ratio data.

the sample is less than 30.

Why do we use the t-test for independent sample?

The t-test is used for independent sample because it is more powerful test compared with other tests of difference of two independent groups.

How do we use the t-test for independent samples?

$$t = \frac{\frac{\frac{SS_1 + SS_2}{n_1 + n_2 - 2}}{\left[\frac{1}{n_1} + \frac{1}{n_2} \right]} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}{\bar{x}_1 - \bar{x}_2}$$

Where:

- t = the t test
- \bar{x}_1 = the mean of group 1
- \bar{x}_2 = the mean of group 2
- SS₁ = the sum of squares of group 1
- SS₂ = the sum of squares of group 2
- n₁ = the number of observations in group 1
- n₂ = the number of observations in group 2

or

$$t = \frac{\frac{(n_1 - 1)(s_1)^2 + (n_2 - 1)(s_2)^2}{n_1 + n_2 - 2}}{\left[\frac{1}{n_1} + \frac{1}{n_2} \right]} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}{\bar{x}_1 - \bar{x}_2}$$

Where: *Type equation here.*

t = the t test

\bar{x}_1 = the mean of group 1 or sample 1

\bar{x}_2 = the mean of group 2 or sample 2

S_1 = the standard deviation of group 1 or sample 1

S_2 = the standard deviation of group 2 or sample 2

n_1 = the number of observations in group 1

n_2 = the number of observations in group 2

Example 1. The following are the scores of 10 male and 10 female AB students in spelling. Test the null hypothesis that there is no significant difference between the performance of male and female AB students in the rest. Use the t-test .05 level of significance.

Male (x_1)	Female (x_2)
14	12
18	9
17	11
16	5
4	10
14	3
12	7
10	2
9	6
17	13

Solution:

Male		Female	
x_1	x_1^2	x_2	x_2^2
14	196	12	144
18	324	9	81
17	289	11	121
16	256	5	25
4	16	10	100
14	196	3	9
12	144	7	49
10	100	2	4
9	81	6	36
17	289	13	169

$\sum x_1 = 131$ $\sum x_1^2 = 1891$ $\sum x_2 = 78$ $\sum x_2^2 = 738$

$n_1 = 10$ $n_2 = 10$

$$\bar{x}_1 = 13.1$$

$$\bar{x}_2 = 7.8$$

$$t = \frac{\frac{n_1 + n_2 - 2}{(n_1 - 1)(s_1)^2 + (n_2 - 1)(s_2)^2} \left[\frac{1}{n_1} + \frac{1}{n_2} \right] (\bar{x}_1 - \bar{x}_2)}{\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

$$t = \frac{13.1 - 7.8}{\sqrt{\left[\frac{174.9 + 129.6}{10 + 10 - 2} \right] \left[\frac{1}{10} + \frac{1}{10} \right]}}$$

$$t = \frac{5.3}{\sqrt{\left[\frac{304.5}{18} \right] (.1 + .1)}}$$

$$t = \frac{5.3}{\sqrt{(16.92)(.2)}}$$

$$t = \frac{5.3}{\sqrt{3.384}}$$

$$t = \frac{5.3}{1.8395}$$

$$t = 2.88$$

Solving by the Stepwise Method

I. Problem:

Is there a significant difference between the performance of the male and female students in spelling?

II. Hypothesis:

H_0 : There is no significant difference between the performance of the male and the female AB students in spelling.

$$H_0: \bar{x}_1 = \bar{x}_2$$

H_a : There is a significant difference between the performance of the male and the female AB students in spelling.

$$H_a: \bar{x}_1 \neq \bar{x}_2$$

/

III. Level of Significance:

$\alpha = .05$

$$df = n_1 + n_2 - 2$$
$$= 10 + 10 - 2$$
$$= 18$$

$t_{.05} = 2.101$ t-tabular value at .05

IV. Statistics: t-test for two independent Samples

V. Decision Rule: If the t- computed value is greater than or beyond the t- tabular / critical value, reject H_0 .

VI. Conclusion:

Since the t-computed value of 2.88 is greater than t-tabular value of 2.101 at .05 level of significance with 18 degrees of freedom, the null hypothesis is rejected in favor of the research hypothesis. This means that there is a significant difference between the performance of the male and female AB students in spelling, implying that the male students performed better than the female students considering that the mean/average score of the male students of 13.1 is greater compared to the average score of female students of only 7.8.

Example 2.

Two groups of experimental rats were injected with tranquilizer at 1.0 mg. And 1.5 mg dose respectively. The time given in seconds that took them to fall asleep in hereby given. Use the t-test for independent samples at .01 to test null hypothesis that the difference in dosage has no effect on the length of time it took them to fall asleep.

1.0 mg dose:	9.8, 13.2, 11.2, 9.5, 13.0, 12.1, 9.8, 12.3, 7.9, 10.2, 9.7.
1.5 mg dose:	12.0, 7.4, 9.8, 11.5, 13.0, 12.5, 9.8, 10.5, 13.5.

Solution:

(1.0 mg dose)		(1.5 mg dose)	
X_1	X_1^2	X_2	X_2^2
9.8	96.04	12.0	144.00
13.2	174.24	7.4	54.76
11.2	125.44	9.8	96.04
9.5	90.25	11.5	132.25
13.0	169.00	13.0	169.00
12.1	146.41	12.5	156.25
9.8	96.04	9.8	96.04
12.3	151.29	10.5	110.25
7.9	62.41	13.5	182.25
10.2	104.04	$\Sigma X_2 = 100$	$\Sigma X_2^2 = 1140.84$
9.7	94.09	$n_2 = 9$	

$$\begin{aligned}\Sigma X_1 &= 118.7 \\ \bar{x}_1 &= 11.11 \\ \Sigma X_1^2 &= 1309.25\end{aligned}$$

$$N_1 = 11$$

$$\bar{X}_1 = 10.79$$

$$SS_1 = \Sigma \frac{1 - \bar{x}_1^2}{X_1} - \frac{(\Sigma x_1 \bar{x}_1^2)}{n_1} \qquad SS_2 = \Sigma X_2^2 - \frac{(\Sigma x_2 \bar{x}_2^2)}{n_2}$$

$$= 1309.25 - \frac{118.7 \bar{x}_1^2}{\bar{x}_1}$$

$$= 1140.84 - \frac{100 \bar{x}_2^2}{\bar{x}_2}$$

$$SS_1 = \boxed{28.37}$$

$$\boxed{29.73}$$

$$t = \frac{\frac{n_1 + n_2 - 2}{SS_1 + SS_2} \left[\frac{1}{n_1} + \frac{1}{n_2} \right] \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\bar{x}_1}}}{\bar{x}_1}$$

$$= \frac{10.79 - 11.11}{\sqrt{\left[\frac{28.37 + 29.73}{11 + 9 - 2} \right] \left[\frac{1}{11} + \frac{1}{9} \right]}}$$

$$= \frac{-0.32}{\sqrt{\left[\frac{58.10}{18} \right] (.09 + .11)}}$$

$$= \frac{-0.32}{\sqrt{(3.23)(.2)}}$$

$$= \frac{-0.32}{\sqrt{.646}}$$

$$= \frac{-0.32}{.8037}$$

$$t = -0.40$$

Solving by the Stepwise Method

- I. **Problem** : Is there a significant difference brought about by the dosages on the length of time it took for the rats to fall asleep
- II. **Hypothesis:** H_0 : There is no significant differences brought about by the dosages on the length of time it took the rats to fall asleep.
 H_a : There is a significant difference brought about by the dosages on the length of time it took for the rats to fall asleep.

III. **Level of Significance:**
 $\alpha = .01$

$$\begin{aligned} df &= n_1 + n_2 - 2 \\ &= 11 + 9 - 2 \\ &= 18 \\ t_{.05} &= -2.88 \end{aligned}$$

- IV. **Statistics:** t-test for two independent samples
- V. **Decision Rule:** If the t- computed value is greater than or beyond the t- tabular / critical value, reject H_0 .
- VI. **Conclusion:** Since the t-computed value of $\pm .40$ is within the critical value of ± 2.88 at .01 level of significance with 18 degrees of freedom, the null hypothesis is not rejected. This means that no significant difference was brought about by the dosages on the length of time it took for the rats to fall asleep.

Example 3. To find out whether a new serum would arrest leukemia, 16 patients on the advanced stage of the disease were selected. Eight patients received the treatment and eight did not. The survival was taken from the time the experiment was conducted.

No treatment		With Treatment	
X_1	X_1^2	X_2	X_2^2
2.1	4.41	4.2	17.64
3.2	10.24	5.1	26.01
3.0	9.00	5.0	25.00
2.8	7.84	4.6	21.16
2.1	4.41	3.9	15.21
1.2	1.44	4.3	18.49
1.8	3.24	5.2	27.04
1.9	3.61	3.9	15.21
$\sum x_1 = 18.1$	$\sum X_1^2 = 44.19$	$\sum x_2 = 36.2$	$\sum X_2^2 = 165.76$
$n_1 = 8$		$n_2 = 8$	
$\bar{x}_1 = 2.2$		$\bar{x}_2 = 4.52$	

Solution: Solve for $\sum x_1$, $\sum X_1^2$ and \bar{x}_1 likewise for $\sum x_2$, $\sum X_2^2$ and \bar{x}_2 . Then solve for the sum of the squares of group 1, and sum of squares of group 2 SS_1 and SS_2 .

$$SS_1 = \sum X_1^2 - \frac{(\sum x_1)^2}{n_1}$$

$$= 44.19 - \frac{18.7^2}{8}$$

$$= 44.19 - 40.95$$

$$SS_1 = 3.24$$

$$SS_2 = \sum X_2^2 - \frac{(\sum x_2)^2}{n_2}$$

$$= 165.76 - \frac{36.2^2}{8}$$

$$= 165.76 - 163.80$$

$$SS_2 = 1.96$$

$$t = \frac{\frac{\sum x_1 - \sum x_2}{\sqrt{\frac{SS_1 + SS_2}{n_1 + n_2 - 2}}}}{\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

$$= \frac{2.26 - 4.52}{\sqrt{\left[\frac{3.24 + 1.96}{8 + 8 - 2} \right] \left[\frac{1}{8} + \frac{1}{8} \right]}}$$

$$= \frac{-2.26}{\sqrt{\left[\frac{5.2}{14} \right] \left[\frac{2}{8} \right]}}$$

$$= \frac{-2.26}{\sqrt{(.37)(.25)}}$$

$$= \frac{-2.26}{\sqrt{.0925}}$$

$$= \frac{-2.26}{.3041}$$

$$t = -7.43$$

Solving by the Stepwise Method

- I. **Problem:** Will the new serum arrest leukemia for the 8 patients who had reached the advanced stage?
- II. **Hypothesis:**

H_0 : The new serum will not arrest the leukemia of the 8 patients who had reached the advanced stage of the disease.

H_1 : The new serum will arrest the leukemia of the 8 patients who had reached an advanced stage of the disease.

III. Level of Significance:

$$\alpha = .05$$

$$\begin{aligned} df &= n_1 + n_2 - 2 \\ &= 8 + 8 - 2 \\ &= 14 \\ t_{.05} &= -2.145 \end{aligned}$$

IV. Statistics: t-test for independent samples

V. Decision Rule: If the t-computed value is greater than or beyond the critical or tabular value, reject H_0 .

VI. Conclusions: The t-computed value is -7.43 which is beyond the critical value of -2.145 at .05 level of significance with 14 degrees of freedom. The null hypothesis therefore is disconfirmed in favor of the research hypothesis. This means that the new serum will arrest the advanced stage of leukemia.

What is the t-test correlated samples?

The t-test for correlated samples is another parametric test applied to one group of samples. It can be used in the evaluation of a certain program or treatment. Since this is another parametric test, conditions must be met like the normal distribution and the use of interval or ratio data.

When do we use the t-test for correlated samples?

The test for correlated samples is applied when the mean before and the mean after are being compared. The pretest (mean before) is measured, the treatment of the intervention is applied and then the posttest (mean after) is likewise measured. Then the two means (pretest vs. the posttest) are compared.

Why do we use the t-test for correlated samples?

The t-test for correlated samples is used to find out if a difference exists between the before and after means. If there is a difference in favor of the posttest then the treatment or intervention is effective. However, if there is no significant difference then the treatment is not effective.

This is the appropriate test for evaluation of government programs. This is used in an experimental design to test the effectiveness of a certain technique or method or program that had been developed.

How do we use the t-test for correlated samples?

The formula is

$$t = \frac{\frac{\sum D^2}{n} - \frac{(\sum D)^2}{n^2}}{\frac{\sum D^2}{n} - \frac{(\sum D)^2}{n^2}}$$

Where: \bar{D} = the mean difference between the pretest and the posttest.

$\sum D^2$ = the sum of the squares of the difference between the pretest and the posttest.

$\sum D$ = the summation of the difference between the pretest and the posttest.

n = the sample size

Example 1. An experiment study was conducted on the effect of programmed materials in English on the performance of 20 selected college students. Before the program was implemented the pretest was administered and after 5 months the same instrument was used to get the posttest result. The following is the result of the experiment. Use α at .05 level.

Pretest	Posttest		
X_1	X_2	D	D^2
20	25	-5	25
30	35	-5	25
10	25	-15	225
15	25	-10	100
20	20	0	0
10	20	-10	100
18	22	-4	16
14	20	-6	36
15	20	-5	25
20	15	5	25
18	30	-12	144
15	10	5	25
20	25	-5	1
18	10	8	64
40	45	-5	25
10	15	-5	25
10	10	0	0
12	18	-6	36
20	25	-5	25
		$\sum D = -81$	$\sum D^2 = 94$
		$\bar{D} = \frac{-81}{20}$	

$$\bar{D} = -4.05$$

What are the steps in using the t-test for correlated samples?

Subtract the difference between the two observations before and after the treatment.

Find the summation of the difference algebraically $\sum D$ (consider the + and – sign)

Compute \bar{D} the mean difference, $\frac{\sum D}{n}$

Square the difference between the before and after observation and get the summation that is $\sum D^2$

Determine the number of observation that is small letter n.

If the t-compound value is greater than or beyond the critical value disconfirm the null hypothesis and confirm the research or alternative hypothesis.

$$\begin{aligned}
 t &= \frac{\frac{\sum D^2 - \frac{(\sum D)^2}{n}}{n(n-1)}}{\frac{\bar{D}}{\sqrt{\frac{1}{n(n-1)}}}} \\
 &= \frac{\frac{-81 - \frac{20}{20(20-1)}}{20(20-1)}}{\frac{947 - \frac{328.05}{20(19)}}{\frac{-4.05}{\sqrt{\frac{1}{20(19)}}}}} \\
 &= \frac{-4.05}{\sqrt{\frac{947 - 328.05}{20(19)}}} \\
 &= \frac{-4.05}{\sqrt{\frac{618.95}{380}}}
 \end{aligned}$$

$$= \frac{-4.05}{\sqrt{1.6288}}$$

$$= \frac{-4.05}{1.2762}$$

$$t = -3.17$$

Solving by the Stepwise Method

I. **Problem:** Is there a significant difference between the pretest and the posttest on the use of programmed materials in English?

II. **Hypotheses:**

H_0 : There is no significant difference between the pretest and the posttest, or the use of the programmed materials did not affect the students' performance in English.

H_1 : The posttest result is higher than the pretest result.

III. **Level of Significance:**

$$\alpha = .05$$

$$\begin{aligned} df &= n-1 \\ &= 20-1 \\ &= 14 \\ t_{.05} &= -1.729 \end{aligned}$$

IV. **Statistics:** t-test for correlated samples

V. **Decision Rule:** If the t-computed value is greater than or beyond the critical value, reject H_0 .

VI. **Conclusion:** The t-computed value of 3.17 is beyond the t-critical value of /-1.73/ at .05 level of significance with 19 degrees of freedom. The null hypothesis is therefore rejected in favor of the research hypothesis. This means that the posttest result is higher than the pretest result. It implies that the use of the programmed materials in English is effective.

What is the z-test?

The z-test is another test under parametric statistics which requires normality of distribution. It uses the two population parameters μ and σ .

It is used to compare two means, the sample mean, and the perceived population mean.

It is also used to compare the two sample means taken from the same population. It is used when the samples are equal to or greater than 30. The z-test can be applied in two ways: the One-Sample Mean Test and the Two-Sample Mean Test.

The tabular value of the z-test at .01 and .05 level of significance is shown below.

Test	Level of Significance	
	.01	.05
One-tailed	± 2.33	± 1.645
Two-tailed	± 2.575	± 1.96

What is the z-test for one sample group?

The z-test for one sample group is used to compare the perceived population mean μ against the sample mean, \bar{X}

When is the z-test for a one-sample group?

The one-sample group test is used when the sample is being compared to the perceived population mean. However if the population standard deviation is not known the sample standard deviation can be used as a substitute.

Why is the z-test used for a one-sample group?

The z-test is used for a one-sample group because this is appropriate for comparing the perceived population mean μ against the sample mean \bar{X} . We are interested if significant difference exists between the population μ against the sample mean. For instance a certain tire company would claim that the life span of its product will last 25,000 kilometers. To check the claim, sample tires will be tested by getting sample mean \bar{X} .

How do we use the z-test for a one-sample group?

The formula is

$$z = \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}}$$

Where:

- \bar{X} = sample mean
- μ = hypothesized value of the population mean
- σ = population standard deviation
- N = sample size

Example 1. The ABC company claims that the average lifetime of a certain tire as at least 28,000 km. To check the claim, a taxi company puts 40 of these tires on its taxis and gets a mean lifetime of 25,560 km. With a standard deviation of 1,350 km, is the claim true? Use the z-test at .05.

What are the steps in using the z-test for a one sample group?

- Solve for the mean of the sample and also the standard deviation if the population σ is not known.
- Subtract the population from the sample mean and multiply it by the square root of n sample.
- Divide the result from step 2 by the population standard deviation on the sample standard deviation if the population σ is not known.
- Compare the result in the table of the tabular value of the z-test at .01 and .05 level of significance.

Test	Level of Significance	
	.01	.05
One tailed	± 2.33	± 1.645
Two tailed	± 2.575	$\pm 1.96\alpha\alpha$

If the z-compound value is greater than or beyond the critical value, reject the null hypothesis and confirm the research hypothesis.

I. **Problem:** Is the claim true that the average lifetime of a certain tire is at least 28,000 km?

II. **Hypotheses:**

H_0 : The average lifetime of a certain tire is 28,000 km.

H_1 : The average lifetime of a certain tire is not 28,000 km.

III. **Level of Significance:**

$\alpha = .05$
 $z = \pm 1.645$

IV. Statistics:

z-test for a one-tailed test

Computation:

$$\begin{aligned} Z &= \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} \\ &= \frac{(25,560 - 28,000) \sqrt{40}}{1,350} \\ &= \frac{(-2,440)(6.32)}{1,350} \\ &= \frac{-15,420.8}{1,350} \\ z &= -11.42 \end{aligned}$$

V. Decision Rule: if the z computed is greater than or beyond the z tabular value, reject the H_0 .

VI. Conclusion: Since the z computed of -11.42 is beyond the critical value of -1.645 at .05 level of significance the research hypothesis is rejected which means that the average lifetime of a certain tire is not 28,000 km.

What is the z-test for a two-sample mean test?

The z-test for a two-sample mean test is another parametric test used to compare the means of two independent groups of samples drawn from a normal population if there are more than 30 samples for every group.

When do we use the z-test for two sample mean?

The z-test for two-sample mean is used when we compare the means of samples of independent groups taken from a normal population.

Why do we use the z-test?

The z-test is used to find out if there is a significant difference between the two populations by only comparing the sample mean of the population.

How do we use the z-test for a two-sample mean test?

The formula is

$$Z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

where:

\bar{x}_1 = the mean of sample 1

\bar{x}_2 = the mean of sample 2

s_1^2 = the variance of sample 1

s_2^2 = the variance of sample 2

n_1 = size of sample 1

n_2 = size of sample 2

Example 1. An admission test was administered to incoming freshmen in the colleges of Nursing and Veterinary Medicine with 100 students each college randomly selected. The mean scores of the given samples where $\bar{x}_1 = 90$ and $\bar{x}_2 = 85$ and the variances of the test scores were 40 and 35 respectively. Is there a significant difference between the two groups? Use .01 level of significance.

What are the steps in solving for the z-test for two-sample mean?

Compute the sample mean of group 1, \bar{x}_1 and also the sample mean of group 2, \bar{x}_2

Compute the standard deviation of group 1 SD_1 and standard deviation of group 2 SD_2 .

Square the SD_1 of group 1 to get the variance of group 1 s_1^2 and also square the SD_2 of the group 2 to get the variance of group 2 s_2^2 .

Determine the number of observation in group 1 n_1 and also the number of observation in group 2 n_2 .

Compare the z computed value from the z- tabular value at certain level of significance.

Level of Significance		
Test	.01	.05
One-tailed	± 2.33	± 1.645
Two-tailed	± 2.575	± 1.96

If the computed z-value is greater than or beyond the critical value reject the null hypothesis and do not reject the research hypothesis.

I. Problem: Is there a significant difference between the two groups?

II. Hypotheses:

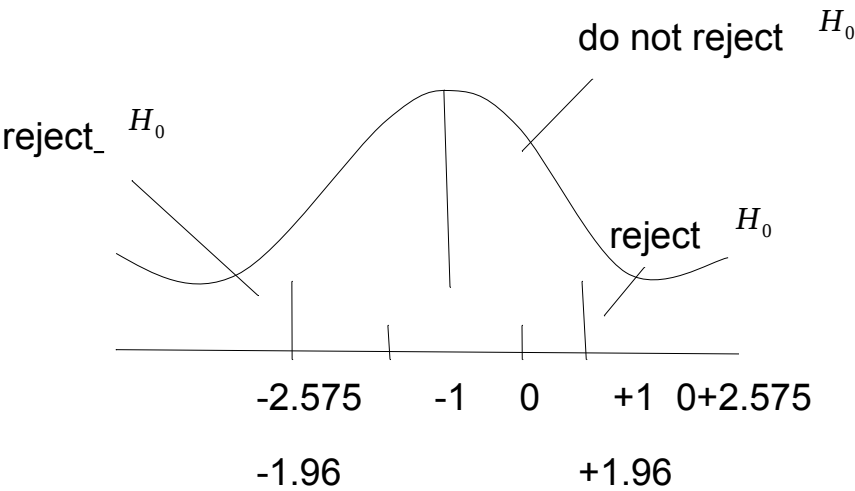
$$H_0: \quad \acute{x}_1=\acute{x}_2$$

$$H_1: \quad \acute{x}_1\neq \acute{x}_2$$

III. Level of Significance

$$\alpha \quad = \text{.01}$$

$$z = \pm 2.575$$



IV. **Statistics:** z-test for two-tailed test

$$Z = \frac{\acute{x}_1-\acute{x}_2}{\sqrt{\frac{s_1^2}{n_1}+\frac{s_2^2}{n_2}}}$$

$$= \frac{90-85}{\sqrt{\frac{40}{100}+\frac{35}{100}}}$$

$$= \frac{5}{\sqrt{\frac{75}{100}}}$$

$$= \frac{5}{\sqrt{.75}}$$

$$= \frac{5}{.866}$$

$$z = 5.774$$

V. Decision Rule: If the z-computed value is greater than or beyond the z-tabular value, reject the null hypothesis..

VI. Conclusion: Since the z-computed value of 5.774 is greater than the z-tabular value of 2.575 at .01 level of significance, the research hypothesis is rejected which means that there is a significant difference between the two groups. It implies that the incoming freshmen of the College of Nursing are better than the incoming freshmen of the College of Veterinary Medicine.

What is the F-test?

The F-test is another parametric test used to compare the means of two or more groups of independent samples. It is also known as the analysis of variance, (ANOVA).

The three kinds of analysis of variance are:

one-way analysis of variance

two-way analysis of variance

three-way analysis of variance

The F-test is the analysis of variance (ANOVA). This is used in comparing the means of two or more independent groups. One-way ANOVA is used when there is only one variable involved. The two-way ANOVA is used when two variables are involved: the column and the row variables. The researcher is interested to know if there are significant differences between and among columns and rows. This is also used in looking at the interaction effect between the variables being analyzed.

Like the t-test, the F-test is also a parametric test which has to meet some conditions, and the data to be analyzed if they are normal are expressed in interval or ratio data. This test is more efficient than other tests of difference.

Why do we use the F-test?

The F-test is used to find out if there is a significant difference between and among the means of the two or more independent groups.

When do we use F-test?

The F-test is used when there is normal distribution and when the level of measurement is expressed in interval or ratio data just like the t-test and the z-test.

How do we use the F-test?

To get the F computed value, the following computations should be done.

$$CF = \frac{(\sum i)^2}{N}$$

TSS is the total sum of squares minus the CF, the correction factor.

BSS is the between sum of squares minus the CF correction factor.

WSS is the sum of squares or it is the difference between the TSS minus BSS.

After getting the TSS, BSS and WSS, the ANOVA table should be constructed.

ANOVA Table					
Sources of	Df	SS	MS	F-Value Computed	Tabular
Between	K-1	BSS	$\frac{BSS}{df}$	$\frac{MSB}{MSW} = F$	see the
Within	(N-1)-(K-1)	WSS	$\frac{WSS}{df}$		desired level of significance
Group					w/ df between
Total	N-1	TSS			and w/ group

What are the steps in solving for the F-value?

The ANOVA table has five columns. These are:
sources of variations, degrees of freedom, sum of squares, mean squares and the F-value, both the computed and the tabular values.

The *sources of variations* are between the groups, within the group itself and the total variations.

The degrees of freedom for the total is the total number of observation minus 1.

The degrees of freedom from the between group is the total number of groups minus 1.

The degrees of freedom for the within group is the total df minus the between groups df.

- The MSB mean squares between is equal to the BSS/df.
- The MSW mean square within is equal to the WSS/df.
- To get the F-computed value, divide MSB/MSW.
- If the F-computed value at a given level of significance with the corresponding df's of BSS and WSS.

- Example 1: A sari-sari store is selling 4 brands of shampoo. The owner is interested if there is a significant difference in the average sales of the four brands of shampoo for one week. The following data are recorded.

Brand			
A	B	C	D
7	9	2	4
3	8	3	5
5	8	4	7
6	7	5	8
9	6	6	3
4	9	4	4
3	10	2	5

Solving by the Stepwise Method

- ## II. Hypotheses:

H_1 : There is a significant difference in the average sales of the four brands of shampoo.

$$CF = \frac{37 + 57 + 26 + 36 \text{ } \color{red}{\mathfrak{I}}^2}{n_1 n_2 n_3 n_4} = \color{red}{\mathfrak{I}}$$

$$TSS = x_1^2 + x_2^2 + x_3^2 + x_4^2 - CF$$

$$= 1014 - 869.14$$

$$\text{BSS} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix}^2$$

$$= \begin{array}{c} 37 \text{ } \textcolor{brown}{\text{!}}^2 \\ \textcolor{brown}{\text{!}} \\ 57 \text{ } \textcolor{brown}{\text{!}}^2 \\ \textcolor{brown}{\text{!}} \\ 26 \text{ } \textcolor{brown}{\text{!}}^2 \\ \textcolor{brown}{\text{!}} \\ 36 \text{ } \textcolor{brown}{\text{!}}^2 \\ \textcolor{brown}{\text{!}} \\ \textcolor{brown}{\text{!}} \\ \textcolor{brown}{\text{!}} \\ \textcolor{brown}{\text{!}} \end{array}$$

$$= 941.42 - 869.14$$

WSS = 72.58

III. Decision Rule: If the F computed value is greater than the F-tabular value, reject H_0 .

IV. **Conclusion:** Since the F-computed value of 7.98 is greater than the F –tabular value of 3.01 at .05 level of significance with 3 and 24 degrees of freedom, the null hypothesis is rejected in favor of the research hypothesis which means that there is a significant difference in the average sales of the 4 brands of shampoo.

What is the Scheffé’s Test?

To find out where the differences lies, another test must be used.

The F-test tells us that there is a significant difference in the average sales of the 4 brands of shampoo but as to where the difference lies, it has to be tested further by another test, the Scheffé’s test formula.

$$F' = \frac{\frac{\bar{x}_1 - \bar{x}_2}{n_1 + n_2}^2}{\frac{SW^2}{df}}$$

Where:

F' = Scheffé’s test

\bar{x}_1 = mean of group 1

\bar{x}_2 = mean of group 2

n_1 = number of samples in group 1

n_2 = number of samples in group 2

SW^2 = within mean squares

A vs. B

$$F' = \frac{5.28 - 8.14}{49}$$

$$\frac{8.1796}{42.28}$$

$$\frac{8.1796}{.86}$$

F' 9.5

A vs C

$F' = \frac{5.28 - 3.71}{.86}$

$= \frac{2.4649}{.86}$

$F' = 2.87$

A vs D

$F' = \frac{5.28 - 5.14}{.86}$

$= \frac{.0196}{.86}$

$F' = .02$

B vs C

$F' = \frac{8.14 - 3.71}{.86}$

$= \frac{19.6249}{.86}$

$F' = 22.82$

B vs D

$F' = \frac{8.14 - 5.14}{.86}$

$= \frac{9}{.86}$

$F' = 10.46$

C vs D

$F' = \frac{-5.14 - 3.71}{.86}$

$= \frac{2.0449}{.86}$

$F' = 2.38$

Comparison of the Average Sales of the Four Brands of Shampoo

Between		F' ($F_{.05}$)	
Brand		$(K-1)$ $(3.01)(3)$	Interpretation
A vs B	9.51	9.03	significant
A vs C	2.87	9.03	not significant
A vs D	.02	9.03	not significant
B vs C	22.82	9.03	significant
B vs D	10.46	9.03	significant
C vs D	2.38	9.03	not significant

The above table shows that there is a significant difference in the sales between brand A and brand B, brand B and brand C and also brand B and D. However, brands A and C, A and D and C and D not significantly differ in their average sales.

This implies that brand B is more saleable than brands A, C and D.

Example 2. The following data represent the operating time in hours of the 3 types of scientific pocket calculators before a recharge is required. Determine the difference in the operating time of the three calculators. Do the analysis of variance at .05 level of significance.

Brand					
F_{x_1}	X_1^2	F_{x_2}	X_2^2	F_{x_3}	X_3^2
4.9	24.01	6.4	40.96	4.8	23.04
5.3	28.09	6.8	46.24	5.4	29.16
4.6	21.16	5.6	31.36	6.7	44.89
6.1	37.21	6.5	42.25	7.9	62.41
4.3	18.49	6.3	39.69	6.2	38.44
6.9	47.61	6.7	44.89	5.3	28.09
		5.3	28.09	5.9	34.81
		4.1	16.81		
		4.3	18.49		
$\Sigma x_1 = 32.1 \quad \Sigma x_1^2 = 176.57 \quad \Sigma x_2 = 52 \quad \Sigma x_2^2 = 308.78 \quad \Sigma x_3 = 42.2 \quad \Sigma x_3^2 = 260.84$					
$n_1 = 6$		$n_2 = 9$		$n_3 = 7$	
$\bar{x}_1 = 5.35$		$\bar{x}_2 = 5.78$		$\bar{x}_3 = 6.03$	

I. **Problem:** Is there a significant difference in the average operating time in hours of the 3 types of pocket scientific calculators before a recharge is required?

II. **Hypotheses:**

H_0 : There is no significant difference in the average operating time in hours among the 3 types of pocket calculators before a recharge is required.

H_1 : There is a significant difference in the average operating time in hours of the 3 types of pocket scientific calculators before a recharge is required.

III. **Level of Significance:**

$\alpha = .05$
 $df = 2 \text{ and } 19$

IV. **Statistics:**
F-test one-way-analysis of variance

Computation:

CF = $\frac{x_1^2 + x_2^2 + x_3^2}{n}$ = 725.08

TSS = $\sum x_i^2 - CF$
= 176.57+308.78+260.78+260.84-725.08
= 746.19 – 725.08
TSS = 21.11

BSS = $\sum \frac{x_i^2}{n_i} - CF$
= $\frac{32.1^2}{2} + \frac{52^2}{2} + \frac{42.2^2}{2} - 725.08$
=171.74+300.44+254.40-725.08
= 726.58-725.08

BSS = 1.50

WSS = TSS-BSS
= 21.11 – 1.50
WSS = 19.61

ANOVA TABLE

Sources of Variation	Degrees of Freedom	Sum of Squares	Mean Squares	Computed F-Value	Tabular
Between Groups K-1	2	1.50	.75	.73	3.52
Within Group (N-1)-(K-1)	19	19.61	1.03		
Total N-1	21	21.11			

- V. **Decision Rule:** If the F-computed value is greater than the F-tabular value, reject the H_0 .
- VI. **Conclusion:** Since the F-computed value of 0.73 is lesser than the F-tabular value of 3.52 at .05 level of significance, the null hypothesis is not rejected. This means that there is no significant difference in the average operating time in hours of the 3 types of pocket scientific calculators before a recharge is required.

What is the F-test two-way-ANOVA with interaction effect?

The F-test, two-way-ANOVA involves two variables, the column and the row variables.

It is used to find out if there is an interaction effect between two variables.

How do you use the F-test two-way-ANOVA with interaction effect?

Consider this example

Example 1. Forty-five language students were randomly assigned to one of three instructors and to one of the three methods of teaching then achievement was measured on a test administered at the end of the term. Use the two-way ANOVA with interaction effect at .05 level of significance to test the following hypotheses:

- H_0 : There is no significant difference in the performance of the three groups of students under three different instructors.

 H_1 : There is a significant difference in the performance of the three groups of students under three different instructors.
- H_0 : There is no significant difference in the performance of the three groups of students under three different methods of teaching.

 H_1 : There is a significant difference in the performance of the three groups of students under three different methods of teaching.
- H_0 : Interaction effects are not present.
 H_1 : Interaction effects are present.

TWO-FACTOR ANOVA with a significant Interaction
TEACHER FACTOR

	A	B	C
Methods of Teaching	40	50	40
	41	50	41
	40	48	40
	39	48	38
	38	45	38

1			
Total			
Method of teaching 2	40	45	50
	41	42	46
	39	42	43
	38	41	43
	38	40	42
Total			
Method of Teaching 3	40	40	40
	43	45	41
	41	44	41
	39	44	39
	38	43	38
Total			
Total			

Solving by the Stepwise Method

- I. **Problem:** 1. Is there a significant difference in the performance of students under the three different teachers?
2. Is there a significant difference in the performance of students under the three different methods of teaching?
3. Is there an interaction effect between teacher and method of teaching factors?

II. **Hypotheses:**

1. H_0 : There is no significant difference in the performance of the three groups of students under three different instructors.
- H_1 : There is a significant difference in the performance of the three groups of students under three different instructors.
2. H_0 : There is no significant difference in the performance of the three groups of students under three different methods of teaching.
- H_1 : There is a significant difference in the performance of the three groups of students under three different methods of teaching.
3. H_0 : Interaction effects are not present.
- H_1 : Interaction effects are present.

III. **Level of Significance**

$$\begin{aligned}
 a &= .05 \\
 df \text{ total} &= N-1 \\
 df \text{ within} &= k(n-1) \\
 df \text{ column} &= c-1 \\
 df \text{ row} &= r-1 \\
 df \text{ c} \times \text{r} &= (c-1)(r-1)
 \end{aligned}$$

IV. **Statistics:** F-test Two-Way-ANOVA with

interaction effect

TWO-FACTOR ANOVA with Significant Interaction
TEACHER FACTOR (Column)

	A	B	C	
Method of Teaching	40	50	40	
FACTOR 1 (row)	41	50	41	
	40	48	40	
	39	48	38	
	38	45	38	
Total	198	241	197	Σ= 636
Method of Teaching	40	45	50	
FACTOR 2 (row)	41	42	46	
	39	42	43	
	38	41	43	
	38	40	42	
Total	196	210	224	Σ= 630
Method of Teaching	40	40	40	
FACTOR 3 (row)	43	45	41	
	31	44	41	
	39	44	39	
	38	43	38	
Total	201	216	199	Σ=616
Total	595	667	620	= 1882

CF = $\frac{1882^2}{636}$ = 78709.42

$SS_T = 40^2 + 41^2 + \dots + 39^2 + 38^2 - CF$

= 79218-78709.42

= 508.58

$$SS_w = 79218 - \frac{198^2}{6} - \frac{196^2}{6} - \frac{201^2}{6} - \frac{241^2}{6} - \frac{210^2}{6} - \frac{216^2}{6} - \frac{197^2}{6} - \frac{224^2}{6} - \frac{199^2}{6}$$

$$= 79218 - 79088.8$$

$$= 129.2$$

$$SS_c = \frac{620^2}{6} + \frac{667^2}{6} + \frac{595^2}{6} - CF$$

$$= \frac{1183314}{15} - 78709.42$$

$$= 178.18$$

$$SS_r = \frac{616^2}{6} + \frac{630^2}{6} + \frac{636^2}{6} - CF$$

$$= \frac{1180852}{15} - 78709.42$$

$$= 787.723.47 - 78709.42$$

$$= 14.05$$

$$SS_{cr} = SS_t - SS_w - SS_c - SS_r$$

$$= 508.58 - 129.2 - 178.18 - 14.05$$

$$= 187.15$$

The degrees of freedom for the different parts of the problem are:

$$\begin{aligned}
 df_t &= N-1 = 45-1 = 44 \\
 df_w &= k(n-1) = 9(5-1) = 9(4) = 36 \\
 df_c &= (c-1) = (3-1) = 2 \\
 df_r &= (r-1) = (3-1) = 2 \\
 df_{c*r} &= (c-1)(r-1) = (3-1)(3-1) = (2)(2) = 4
 \end{aligned}$$

ANOVA TABLE

Sources of Variation	SS	d f	MS	F-value		
				Comput ed	Tabul ar	Interpretati on
Between Columns	178.18	2	89.9	24.82	3.26	S
Rows	14.05	4	7.02	1.95	3.26	NS
Interacti on	187.15	4	46.79	13.03	2.63	S
Within	129.20	36	3.59			
Total	508.58	40				

F-Value Computed:

$$\text{Columns} = \frac{MS_c}{MS_w} = \frac{89.09}{3.59} = \underline{24.82}$$

$$\text{Row} = \frac{MS_R}{MS_w} = \frac{7.02}{3.59} = \underline{1.95}$$

$$\text{Interaction} = \frac{MS_I}{MS_w} = \frac{46.79}{3.59} = 13.03$$

F-Value Tabular at .05

Columns $df = 2/36 = 3.26$

Row $df = 2/36 = 3.26$
 Interaction $df = 4/36 = 2.63$

- V. **Decision Rule:** If the computed F value is greater than the F critical/tabular value, reject H_0 .
- VI. **Conclusion:** With the computed F-value (column) of 24.82 compared to the F-tabular value of 3.26 at .05 level of significance with 2 and 36 degrees of freedom, the null hypothesis is rejected in favor of the of the research hypothesis which means that there is a significant difference in the performance of the three

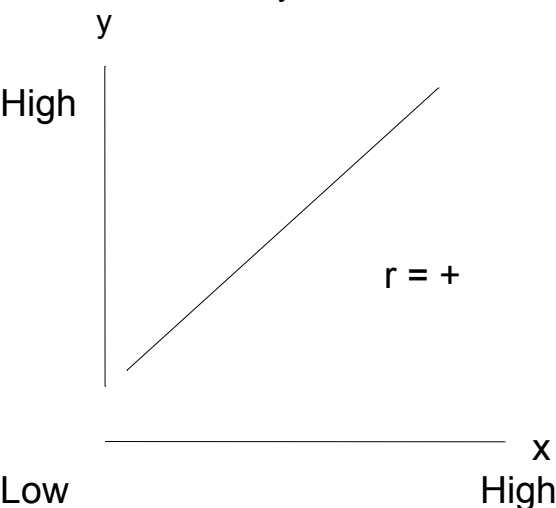
groups of students under three different instructors. It implies that instructor B is better than instructor A. With regard to the F-value (row) of 1.95, it is lesser than the F-tabular value of 3.26 at .05 level of significance with 2 and 36 degrees of freedom. Hence, the null hypothesis of no significant differences in the performance of the students under the three different methods of teaching is not rejected.

However, the F-value (interaction) of 13.03 is greater than the F-tabular value of 2.63 at .05 level of significance with 4 and 36 degrees of freedom. Thus, the research hypothesis is rejected which means that interaction effect is present between the instructors and their methods of teaching. Students under instructor B have better performance under methods of teaching 1 and 3 while students under instructor C have better performance under method 2.

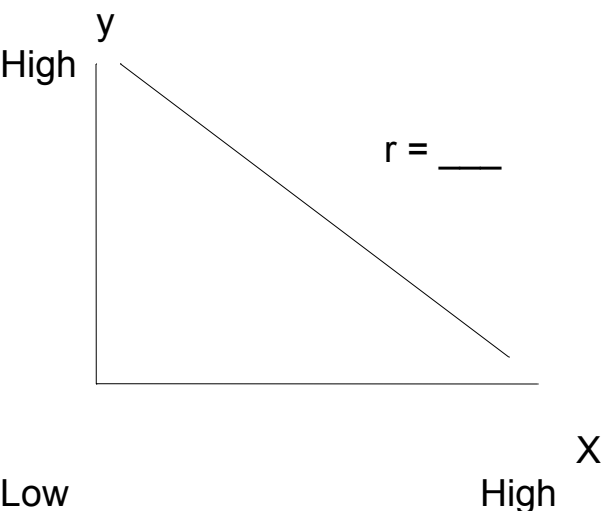
What is the Pearson Product Moment Coefficient of Correlation r ?

The Pearson Product Moment Coefficient of Correlation r is an index of relationship between two variables. The independent variable can be represented by x while the dependent variable can also be represented by y . The value of r is +1, zero to -1. If the value of r is +1 or -1, there is a perfect correlation between x and y . However, if r equals zero then x and y are independent of each other.

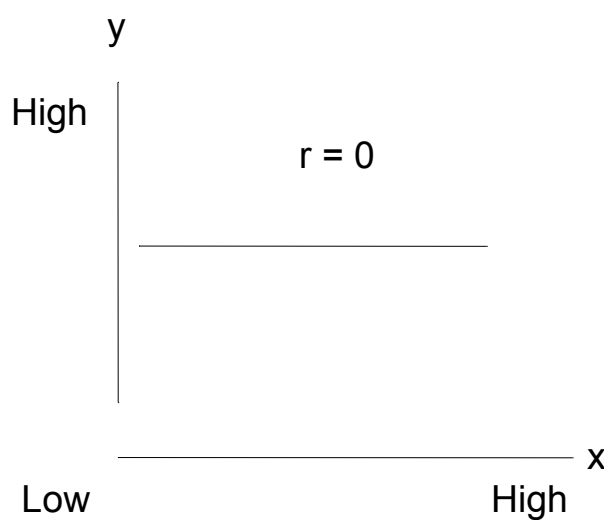
Consider the x and y coordinates in the graph below.



If the trend of the line graph is going upward, the value of r is positive. This indicates that as the value of x increase the value of y also increase. Likewise, if the value of x decreases, the value of y also decreases, the x and y being positively correlated.



If the trend of the line graph is going downward, the value of r is negative. It indicates that as the value of x increases the corresponding value of y decreases, x and y being negatively correlated.



If the trend of the line graph cannot be established either upward or downward, then $r = 0$, indicating that there is no correlation between the x and y variables.

Why do we use r ?

We use r because we want to analyze if a relationship exists between two variables. If there is a relationship that exists between the x and y , then we can determine the extent by which x influences using the coefficient of determination which is equal to the square of r and multiplied by 100%. This can answer or explain how much the independent variable influences the dependent variables or how much y depends on x . This is now the degree of relationship between x and y which can not be seen in other statistical tests of relationship.

We also use r because it is a more powerful test of relationship compared with other nonparametric tests.

When do we use r , the Pearson Product Moment Coefficient of Correlation?

We use the r to determine the index of relationship between two variables, the independent and the dependent variables. If there is relationship between the independent and the dependent variables, we can say that x influences y or y depends on x . however if there is no relationship that exists between x and y then x and y are independent of each other.

The value of r ranges from $+1$ through zero -1 . There is a perfect positive correlation of $r = +1$, likewise there is a negative perfect correlation if the value of $r = -1$. However if $r = 0$ then there is no correlation between the two variables x and y . If there is a positive correlation it indicates that for every increase of y or for every decrease of x there is a corresponding decrease of y . If there is a negative correlation it indicates that for every increase of x there is a corresponding decrease of y , likewise for every decrease of x there is a corresponding increase of y . The relationship is inverse.

How do we use r Pearson Product Moment Coefficient Correlation?

The Pearson Product Moment Coefficient of Correlation represented by small letter r is determined by using the formula

$$r = \frac{\frac{\sum y^2}{2} - \frac{(\sum y)^2}{n}}{\sqrt{\frac{\sum x^2}{2} - \frac{(\sum x)^2}{n}} \sqrt{\frac{\sum y^2}{2} - \frac{(\sum y)^2}{n}}}$$

Where:

- r = the Pearson Product Moment Coefficient of Correlation
- n = sample size
- Σxy = the sum of the product of x and y
- Σx Σy = the product of the sum of Σ x and the sum of Σ y
- Σ x² = sum of squares of x
- Σ y² = sum of squares of y

Example 1. Below are the midterm (x) and final (y) grades.

x	75	70	65	90	85	85	80	70	65	90
y	80	75	65	95	90	85	90	75	70	90

What are the steps in solving for the value of r?

- Determine the number of observation n.
- Get the sum of x that is Σx, the independent variable. Square every x observation and get the sum of x² .
- Get the sum of y the dependent variable Σy.
- Square every y observation and get the sum of y² , that is Σ y² .
- Multiply the x and y. Place it in column xy and get the sum of xy, that is Σxy.
- Apply the formula indicated above and compare the computed r with the tabular value at a certain level of significance with n-2 degrees of freedom. If the computed r value is greater than r tabular value, disconfirm the null hypothesis and confirm the research hypothesis which means that there is a significant relationship between the two variables, x and y.

Solving by Stepwise Method

I. **Problem:** Is there a significant relationship between the midterm and the final examination of 10 students in Mathematics?

II. **Hypotheses:**

H_0 : There is no significant relationship between the midterm grades and the final examination/grades of 10 students in Mathematics.

H_1 : There is a significant relationship between the midterm grades and the final examination grades of 10 students in Mathematics.

III. **Level of Significance:**

$$\begin{aligned} \alpha &= .05 \\ df &= n-2 \\ &= 10-2 \\ &= 8 \\ r_{.05} &= .320 \end{aligned}$$

IV. **Statistics:**

Pearson Product Moment Coefficient of Correlation				
x	y	x^2	y^2	xy
75	80	5625	6400	6000
70	75	4900	5625	5250
65	65	4225	4225	4225
90	95	8100	9025	8550
85	90	7225	8100	7650
85	85	7225	7225	7225
80	90	6400	8100	7200
70	75	4900	5625	5250
65	70	4225	4900	4550
90	90	8100	8100	8100
$\Sigma x =$ 775	$\Sigma y =$ 815	$\Sigma x^2 = 60,$ 925	$\Sigma y^2 = 67,$ 325	Σxy 64,000
$\bar{x} = 7$ 7.5	$\bar{y} = 81$.5			

$$\begin{aligned} r &= \frac{\frac{\Sigma y^2}{n} - \frac{(\Sigma y)^2}{n^2}}{\sqrt{\left(\frac{\Sigma x^2}{n} - \frac{(\Sigma x)^2}{n^2} \right) \left(\frac{\Sigma y^2}{n} - \frac{(\Sigma y)^2}{n^2} \right)}} \\ &= \frac{815^2 - \frac{(\Sigma y)^2}{n}}{\sqrt{10(60,925) - \frac{(\Sigma x)^2}{n}} \sqrt{10(67,325) - \frac{(\Sigma y)^2}{n}}} \\ &= \frac{10(64,000) - (775)(815)}{\sqrt{10(60,925) - \frac{(\Sigma x)^2}{n}} \sqrt{10(67,325) - \frac{(\Sigma y)^2}{n}}} \end{aligned}$$

$$= \frac{8,375}{\sqrt{609,250 - 600,625 \quad 673,250 - 664,225}}$$

$$= \frac{8,375}{\sqrt{86259025}}$$

$$= \frac{8,375}{\sqrt{77840625}}$$

$$= \frac{8,375}{8822.73}$$

$$r = .949$$

- V. **Decision Rule:** If the computed r value is greater than the r tabular value, reject the H_0 .

VI. Conclusion / Implication:

Since the computed value of r which is .949 is greater than the tabular value of .632 at .05 level of significance with 8 degrees of freedom, the null hypothesis is rejected in favor of the research hypothesis. This means that there is a significant relationship between the midterm grades of students and the final examination. It implies that the higher also are the final grades because the value of r is positive. Likewise the lower also are the final grades.

What is the simple linear regression analysis?

The simple linear regression analysis predicts the value of y given the value x.

When do we use the simple linear regression analysis?

The simple linear regression analysis is used when there is a relationship between x independent variable and y dependent variable. This is used in predicting the value of y given the value of x. Like other parametric tests it must also meet some conditions. First, the data should be normally distributed using the level of measurement which is expressed in an interval or ratio data.

The formula for the simple linear regression analysis is,

$$Y=a+bx$$

Why do we use linear regression analysis?

The linear regression analysis is used because we are interested in predicting the value of x, the independent variable. This is used for forecasting and prediction.

How do we use the simple linear regression analysis?

We have to apply the formula

$$Y = a + bx$$

Where:

- y = the dependent variable
- x = the independent variable
- a = the y intercept
- b = the slope of the line

To solve for **b** the formula is

$$b = \frac{\sum xy - \frac{\sum x \sum y}{n}}{\sum x^2 - \frac{(\sum x)^2}{n}}$$

To solve for a the formula is

$$a = \bar{y} - b\bar{x}$$

Use the data in the Pearson Product Moment Coefficient of Correlation r.

Determine the number of observation.

Solve for $\sum xy$, $\sum x$, $\sum y$, $\sum x^2$ apply the formula for b.

To solve for a, find the mean of y, \bar{y} and the mean of x, \bar{x} .

Example. Suppose the midterm report is $x = 88$, what is the value of the final grade?

Solution:

$$\begin{aligned} b &= \frac{\sum xy - \frac{\sum x \sum y}{n}}{\sum x^2 - \frac{(\sum x)^2}{n}} \\ &= \frac{10(64,000) - (775)(815)}{10(60,925) - (775)^2} \\ &= \frac{640,000 - 631,625}{609,250 - 600,625} \\ &= \frac{8375}{8625} \end{aligned}$$

$$b = .971$$

$$a = \hat{y}-b\hat{x}$$

$$\begin{aligned} &= 81.5 - .971 (77.5) \\ &= 81.5 - 75.25 \\ &= 6.25 \end{aligned}$$

The Regression Equation is,

$$\begin{aligned} y &= a + bx \\ y &= 6.25 + .971x \\ y &= 6.25 + .971 (88) \\ &= 6.25 + 85.45 \\ y &= \mathbf{91.70 \text{ or } 92 --- final grade} \end{aligned}$$

Example 2. A study is conducted on the relationship of the number of absences (x) and the grades (y) of 15 students in English. Using r at .05 level of significance and the hypothesis that there is no significant relationship between absences and grades of the students in English, determine the relationship using the following data.

Number of Absences	Grades in English
x	y
1	90
2	85
2	80
3	75
3	80
8	65
6	70
1	95
4	80
5	80
5	75
1	92
2	89
1	80
9	65

Solving by Stepwise Method

I. **Problem:** Is there a significant relationship between the number of absences and the grades of 15 students in an English class?

II. **Hypotheses:**

$$H_0 : \text{ There is no significant relationship between}$$

the number of absences and the grades of 15 students in an English class.

H_1 : There is no significant relationship between the number of absences and the grades of 15 students in an English class.

III. Level of Significance:

$$\begin{aligned} \alpha &= .05 \\ df &= n - 2 \\ &= 15 - 2 \\ &= 13 \\ r_{.05} &= .514 \end{aligned}$$

IV. Statistics:

r Pearson Product Moment Coefficient of Correlation

x	y	x^2	y^2	xy
1	90	1	8100	90
2	85	4	7225	170
2	80	4	6400	160
3	75	9	5625	225
3	80	9	6400	240
8	65	64	4225	520
6	70	36	4900	420
1	95	1	9025	95
4	80	16	6400	320
5	80	25	6400	400
5	75	25	5625	375
1	92	1	8464	92
2	89	4	7921	178
1	80	1	6400	80
9	65	81	4225	585
$\Sigma x = 53$	$\Sigma y = 1201$	$\Sigma x^2 = 281$	$\Sigma y^2 = 97335$	$\Sigma xy = 3950$
n = 15	n = 15			
$\bar{x} = 3.53$	$\bar{y} = 80.07$			

$$r = \frac{\Sigma xy - \frac{\Sigma x \Sigma y}{n}}{\sqrt{(\Sigma x^2 - \frac{(\Sigma x)^2}{n})(\Sigma y^2 - \frac{(\Sigma y)^2}{n})}}$$

$$\begin{aligned}
& \frac{1201 \sum x^2 - (\sum x)^2}{53 \sum x^2 - (\sum x)^2} \\
& = \frac{15(281) - (1201)}{15(3950) - (53)(1201)} \\
& = \frac{59250 - 63653}{\sqrt{4215 - 2809} \sqrt{1460025 - 1442401}} \\
& = \frac{-4403}{\sqrt{140617624}} \\
& = \frac{-4403}{\sqrt{24779344}} \\
& = \frac{-4403}{4977.88}
\end{aligned}$$

r = -.88

V. **Decision Rule:** If the r computed value is greater than or beyond the critical value, reject the H_0 .

VI. **Conclusion:** The computed r value of -.88 is beyond the critical value of -.514 at .05 level of significance with 13 degrees of freedom, so the null hypothesis is rejected. This means that there is a significant relationship between the number of absences and the grades of students in English. Since the value of r is negative, it implies that students who had more absences had lower grades.

Suppose we want to predict the grade (y) of the student who has incurred 7 absences (x). To get the value of y given the value of x, the simple linear regression analysis will be used.

$y = a + bx$ is the regression equation.

Solve for a and b

$$\begin{aligned}
b & = \frac{\sum xy - \frac{\sum x \sum y}{n}}{\sum x^2 - \frac{(\sum x)^2}{n}} \\
& = \frac{15(3950) - (53)(1201)}{53 \sum x^2 - (\sum x)^2} \\
& = \frac{59250 - 63653}{4215 - 2809}
\end{aligned}$$

$$= \frac{-4403}{1406}$$

$$b = -3.13$$

$$a = \bar{y} - b\bar{x}$$

$$= 80.07 - (-3.13)(3.53)$$

$$= 80.07 - (-11.05)$$

$$= 80.70 + 11.05$$

$$a = 91.12$$

$$y = a + bx$$

$$= 91.12 + (-3.13)x$$

$$= 91.12 - 3.13x$$

$$= 91.12 - 3.13(7)$$

$$= 91.12 - 21.91$$

$$y = 69.21 \text{ or } 69 \quad \text{grade}$$

What is the multiple regression analysis?

The multiple regression analysis is used to predict the dependent variables y given the independent variables x_0 .

Aside from making predictions we can also see relationship between the dependent variables and the different independent variables. For instance, we can make better predictions of the performance of newly hired teachers if we consider not only their education but also their years of experience, x_1 personality x_2 , attitude x_3 , and other variables that may influence performance.

When do we use the multiple regression analysis?

we use the multiple regression analysis when predicting y dependent variable with 2 or more independent variables x_8 .

we want to know if there is a relationship that exists between dependent variables and among the independent variables.

Why do we use the multiple regression analysis?

The multiple regression analysis is used because we want to know the extent of influence that the independent variables have on the dependent variables through the $r^2 \times 100\%$ coefficient of determination and to know whether the correlation is positive or negative as indicated in the value of r .

How do we use the multiple regression analysis?

Use the formula and steps in solving for the multiple regression analysis and its example.

Many mathematical formulas can serve to express relationship among more than two variables, but the most commonly used in statistics are linear equations.

$$y = \frac{1 + b_2 x_2 + \dots + b_n x_n}{b_0 + b_1 x_1}$$

where:

- y = the dependent variable to be predicted
- $x_1, x_2 \dots x_n$ = the known independent variables that may influence y
- $b_0, b_1, b_2 \dots b_n$ = numerical constants which must be determined from observed data

For instance, when there are two independent variables $x_1 \wedge x_2$ and we want to fit the equation

$$y = b_0 + b_1 x_1 + b_2 x_2$$

We must solve the three normal equations:

$$\begin{aligned} y &= nb_0 + x_1 b_1 + x_2 b_2 \\ \sum x_1 y &= x_1 b_0 + x_1^2 b_1 + x_1 x_2 b_2 \\ \sum x_1^2 y &= x_2 b_0 + x_1 x_2 b_1 + x_2^2 b_2 \end{aligned}$$

Example 1. The following are data on the ages and incomes of a random sample of 5 executives working for in ABC corporation and their academic achievements while in college.

Income (In thousand pesos) y	Age x ₁	Academic Achievement x ₂
81.7	38	1.50
73.3	30	2.00
89.5	46	1.75
79.0	40	1.75
69.9	32	2.50

- Fit an equation of the form $y = nb_0 + x_1 b_1 + x_2 b_2$ to the equation of the given data.
- Use the equation of the form obtained in (a) to estimate the average income of a 35-year old executive with 1.25 academic achievement.

y	x_1	x_2	x_1^2	x_2^2	$x_1 y$	$x_2 y$	$x_1 x_2$
81.7	38	1.5	1444	2.25	3104.6	122.55	57
73.3	30	2.0	900	4	2199	147.4	60
89.5	46	1.75	2116	3.06	4117	156.62	80.5
79.0	40	1.75	1600	3.06	3160	138.25	70
69.9	32	2.5	1024	6.25	2236.8	174.75	80
$\Sigma y = 393.4$	$\Sigma x_1 = 186$	$\Sigma x_2 = 9.5$	$\Sigma x_1^2 = 7084$	$\Sigma x_2^2 = 18.62$	$\Sigma x_1 y = 14817.4$	$\Sigma x_2 y = 739.58$	$\Sigma x_1 x_2 = 347.5$

Computations:

Solve for b_0 , b_1 , and b_2 using the 3 equations:

1) $y = nb_0 + x_1 b_1 + x_2 b_2$
2) $\Sigma x_1 y = x_1 b_0 + x_1^2 b_1 + x_1 x_2 b_2$
3) $\Sigma x_2 y = x_2 b_0 + x_1 x_2 b_1 + x_2^2 b_2$

1) $393.4 = 5b_0 + 186b_1 + 9.5b_2$
2) $14817.4 = 186b_0 + 7084b_1 + 347.5b_2$
3) $739.58 = 9.5b_0 + 347.5b_1 + 18.62b_2$

Step 1. Eliminate b_0 using equations 1 and 2. Look at their numerical coefficients. Try to divide 186 by 5 and the quotient is 37.2. If you multiply 37.2 by 5 the product is 186. So you can eliminate b_0 by subtraction.

Step 2. Multiply equation 1 by 37.2 making a new equation 4. Subtract equation 2 from equation 4. The result is equation 5.

4) $14634.48 = 186 b_0 + 6919.2 b_1 + 353.4 b_2$
2) $14817.40 = 186 b_0 + 7084.0 b_1 + 347.5 b_2$

5) $-182.92 = 0 - 164.8 b_1 + 5.9 b_2$

Step 3. Eliminate b_0 using equations 1 and 3. Look at their numerical coefficients. Try to divide 9.5 by 5 and the quotient is 1.9. If you multiply 1.9 by 5 the product is 9.5. So you can eliminate b_0 by subtraction.

Step 4. Multiply equation 1 by 1.9 making a new equation 6. Subtract equation 3 from equation 6. The result is equation 7.

$$6) \quad 747.46 = 9.5 \, b_0 + 353.4 \, b_1 + 18.05 \, b_2$$

$$3) \quad 739.58 = 9.5 \, b_0 + 347.5 \, b_1 + 18.62 \, b_2$$

$$5) \quad 7.88 = 0 + 5.9 \, b_1 + .57 \, b_2$$

Step 5. Equations 5 and 7 have b_1 and b_2 to be eliminated. To eliminate b_1 , multiply equation 5 by the numerical coefficient of b_1 of equation 7, making a new equation 8. Likewise multiply equation 7 by the numerical coefficient of b_1 of equation 5, making a new equation 9. Then add.

$$8) \quad -1079.228 = -972.32 \, b_1 + 34.81 \, b_2$$

$$9) \quad +1298.624 = +972.32 \, b_1 + 93.936 \, b_2$$

$$5) \quad 219.396 = 0 - 5.9.126 \, b_2$$

Step 6. Solve for b_1 using either equation 5 or 7. Using equation 5:

$$5) \quad -182.92 = -164.8 \, b_1 + 5.9 \, b_2$$

$$-182.92 = 164.8 \, b_1 + 5.9(-3.71)$$

$$-182.92 = -164.8 \, b_1 - 21.889$$

$$21.889 - 182.92 = -164.8 \, b_1$$

$$-161.031 = -164.8 \, b_1$$

$$.98 = b_1$$

Step 7. Solve for b_0 using either equations 1, 2, or 3. Using equation 1:

$$1) \quad 393.4 = 5 \, b_0 + 186 \, b_1 + 9.5 \, b_2$$

$$393.4 = 5 \, b_0 + 186(.98) + 9.5(-3.71)$$

$$393.4 = 5 \, b_0 + 182.28 - 35.245$$

$$35.245 - 182.28 + 393.4 = 5 \, b_0$$

$$246.365 = 5 \, b_0$$

$$49.27 = b_0$$

a) The regression is

$$y = b_0 + b_1 x_1 + b_2 x_2$$

$$y = 49.27 + .98 x_1 - 3.71 x_2$$

- b) To estimate the average income (y) of a 35-year-old ($x_1 = 35$) executive with 1.25 academic achievements ($x_2 = 1.25$) :