

# TEST OF HYPOTHESIS

## Definition of Hypothesis

A hypothesis is a tentative statement about the relationship between two or more variables. It is a specific, testable prediction about what you expect to happen in a study.

## Diagramming hypotheses:-

Diagramming hypotheses is a useful technique to help clarify your thinking.

Usually, a hypothesis takes the form 'X causes Y' or 'X is related to Y'

$$X \xrightarrow{\quad} Y$$

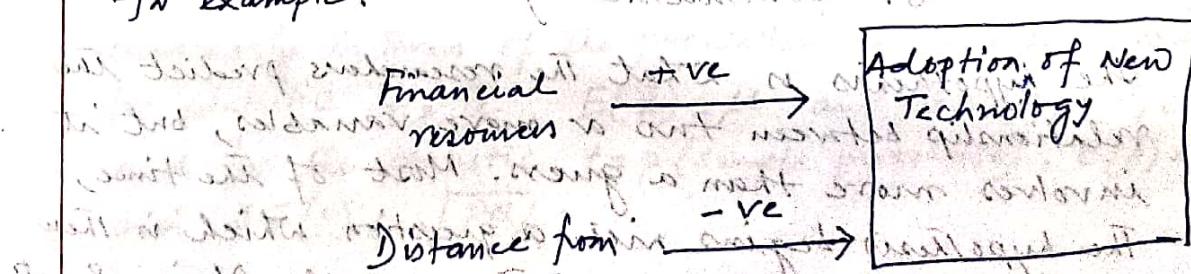
For example, the first hypothesis stated that above could be represented by a diagram as follows-

$$\text{Financial resources} \xrightarrow{+ve} \text{Adoption of new technology}$$

The two variables or concepts are linked by an arrow going from one concept to the other. The arrow indicates that one variable (financial resources) does something to the other variable (adoption of new technology).

The plus sign indicates that the relationship is seen as positive, that is more of the one will lead to more of the other. Not all concepts are have a positive relationship.

Suppose, involving more than two concepts or variables for example:



In this case two concepts, financial resources and distance from market, are related as independent concepts to the dependent concept, adoption of technology.

Here one of the independent concepts is positively related and other negatively related to the dependent concept.

There are endless possibilities. Most research proposals or projects deal only with one small area of the diagram. But it is often useful to make a diagram of more than you plan to study in order to show where your research fits into the larger frame of things and help you to identify factors which may have to taken into account.

### How is a hypothesis used in the scientific method

In the scientific method in Social Science research involves the following steps:-

1. Forming a question

2. Performing background research

3. Creating hypothesis

4. Designing an experiment

5. Collecting data

6. Analyzing the results

7. Drawing conclusions

8. Communicating the results

The hypothesis is what the researchers predict the relationship between two or more variables, but it involves more than a guess. Most of the time, the hypothesis begins with a question which is then explored through background research. It is only at

This point the researcher begin to develop a testable hypothesis

## Types of Statistical Hypotheses:

There are two types of statistical hypotheses:

### 1. Null Hypothesis:

The null hypothesis, denoted by  $H_0$ , is usually the hypothesis that sample observations result purely from chance. The null hypothesis, always states that the treatment has no effect (no change, no difference). According to the null hypothesis, the population mean after treatment is the same as it was before treatment.

### How to set up a Null hypothesis:

The following points may be followed in mind in setting the null hypothesis:-

(i) If we want to test the significance of the difference between a statistic and parameter or between two sample statistics then set up the null hypothesis  $H_0$  that the difference is not significant.

(ii) If we want to test any statement about the population, we set up the null hypothesis that it is true. For example, if the population mean has specified value  $M_0$ , then  $H_0: M = M_0$

(iii) To test the significance of any statistic 't' we compute the standardised variate

$$Z = \frac{t - E(t)}{S.E(t)} \quad [\text{for large Sample test}]$$

For any statistic  $t$ , the values of  $E(t)$  and  $S.E(t)$  are invariably in terms of the population parameters. For example for statistic  $t = \bar{x}$ ,  $E(\bar{x}) = \mu$  and  $S.E(\bar{x}) = \sigma/\sqrt{n}$  and for the statistic  $t = p$  (sample proportion),  $E(p) = P$  and  $S.E(p) = \sqrt{PQ/n}$

## Alternative hypothesis:

For any hypothesis which is complementary to the null hypothesis is called an alternative hypothesis and it is denoted by  $H_1$ . It is very important to explicitly state the alternative hypothesis in respect of any null hypothesis  $H_0$  because the acceptance or rejection of  $H_0$  is meaningful only if it is being tested against a rival hypothesis. For example, if we want to test the null hypothesis that the population has specified mean ( $\mu_0$ ) i.e.

$$H_0: \mu = \mu_0$$

Then the alternative hypothesis could be:

$$(i) H_1: \mu \neq \mu_0 \text{ (i.e. } \mu > \mu_0 \text{ or } \mu < \mu_0\text{)}$$

$$(ii) H_1: \mu > \mu_0$$

$$(iii) H_1: \mu < \mu_0$$

The alternative hypothesis in (i), (ii) or (iii)

(i) is known as a two-tailed alternatives

(ii) is known as right-tailed alternatives

(iii) is known as left-tailed alternatives.

The null hypothesis consists of only a single parameter value and is usually simple while alternative hypothesis is usually composite.

Ans. Here (i) is the random sample obtained from the same population containing the true mean of  $\mu = 50$  and  $H_0: \mu = 50$  is the hypothesis of (H0), (Null hypothesis) i.e. it includes all the values of the population having mean  $\mu = 50$  & (ii) the alternative hypothesis  $H_1: \mu \neq 50$  does not include the true mean of the population.

Ans. The random sample obtained from the same population containing the true mean of  $\mu = 50$  and  $H_0: \mu = 50$  is the hypothesis of (H0), (Null hypothesis) i.e. it includes all the values of the population having mean  $\mu = 50$  & (ii) the alternative hypothesis  $H_1: \mu \neq 50$  does not include the true mean of the population.

## Types of Error in Testing of hypothesis:

Type I error: A type I error occurs when the researcher rejects a null hypothesis when it is true.

The probability of committing a Type I error is called the significance level.

This probability is also called  $\alpha$  (alpha).

Type-II error: A Type-II error occurs when the researcher fails to reject a null hypothesis that is false. The probability of committing a Type-II error is called  $\beta$  (Beta).  
The probability of not committing a Type-II error is called the Power of the test.

In any test procedure, the four possible mutually disjoint and exhaustive decisions are:

i) Reject  $H_0$  when actually it is not true.  
i.e., when  $H_0$  is false.

ii) Accept  $H_0$  when it is true.

iii) Reject  $H_0$  when it is true.

iv) Accept  $H_0$  when it is false.

The decisions in (i) and (ii) are correct decisions while decisions in (iii) and (iv) are wrong decisions. These decisions may be expressed in the following dichotomous table:

Decision from Sample

	Reject $H_0$	Accept $H_0$
$H_0$ (True)	Wrong (Type-I)	Correct
$H_0$ (False) ( $H_1$ True)	Correct	Wrong (Type-II error)

Then we can write,

$P[\text{Reject } H_0 \text{ when it is true}]$

$= P[\text{Type I Error}] = \alpha.$

$P[\text{Accept } H_0 \text{ when it is wrong}]$

$= P[\text{Type-II Error}] = \beta$

Then  $\alpha$  and  $\beta$  are also called the sizes of Type-I and Type-II error respectively.

### Level of Significance :

The maximum size of the type-I error, which we are prepared to risk is known as the level of significance.

$P[\text{Rejecting } H_0 \text{ when } H_0 \text{ is true}] = \alpha.$

Commonly used levels of significance in practice are 5% (0.05) and 1% (0.01). If we adopt 5% level of significance, it implies that 5 samples out of 100 we are likely to reject a correct  $H_0$ .

In other words this implies that we are 95% confident that our decision to reject  $H_0$  correct. Level of significance is always fixed in advance before collecting the sample information.

Thus we can say that  $\alpha$  is the level of significance and  $(1-\alpha)$  is the degree of confidence which is called the confidence coefficient.

signed mark position

Actual $H_0$	Decision	Probability
True ( $H_0$ )	Reject ( $H_0$ )	$(1-\alpha)$
False ( $H_1$ )	Accepted ( $H_0$ )	$\alpha$

### Critical Region :

Suppose we take several samples of the same size from the given population and compute some statistic  $t$  (say,  $\bar{x}$ ,  $p$  etc), for each of these samples. Let  $t_1, t_2, \dots, t_n$  be the values of the statistic for these samples. Each of these values may be used to test some null hypothesis  $H_0$ . The statistics which lead to the rejection of  $H_0$  give us a region called Critical Region ( $C$ ), or Rejection Region ( $R$ ) while those which lead to the acceptance of  $H_0$  gives us a region called Acceptance Region ( $A$ ).

### Critical Values or Significant Value:

The value of test statistic which separates the critical (or rejection) region and the acceptance region is called Critical value or Significant value.

It depends on:

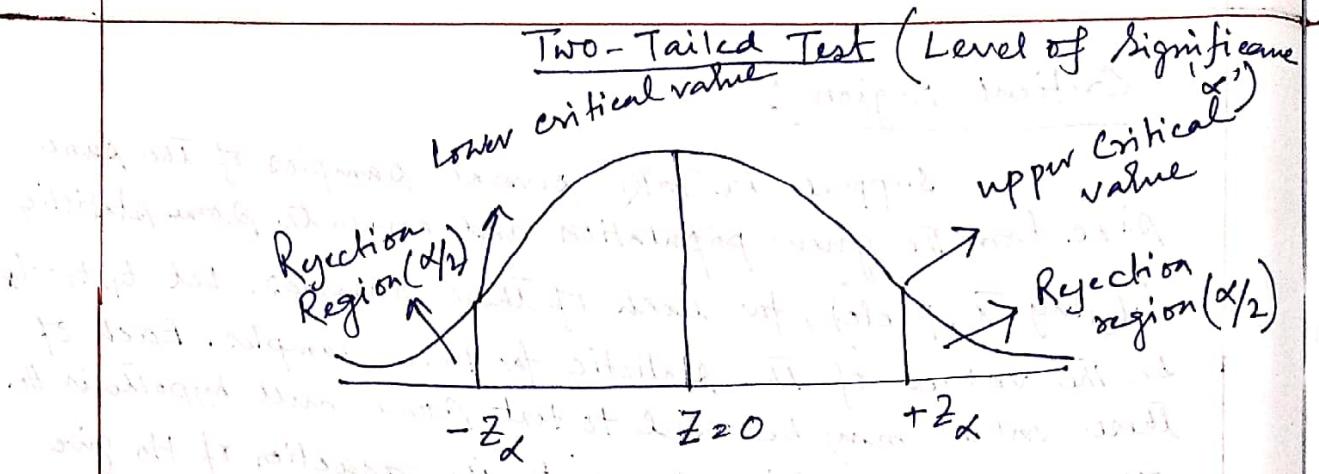
i) The level of significance used

ii) The alternative hypothesis, whether it is two-tailed or single tail test.

The critical value of the test statistic at level of significance ( $\alpha$ ) for two tailed test is given by  $Z_\alpha$  where  $Z_\alpha$  is determined by the equation

$$P(|Z| > Z_\alpha) = \alpha$$

i.e.  $Z_\alpha$  is the value of so that total area of the critical region on both tails is  $\alpha$ . Since normal probability curve is a symmetrical curve —



from the normal distribution —

$$\begin{aligned} P(Z > z_\alpha) + P(Z < -z_\alpha) &= \alpha \\ \Rightarrow P(Z > z_\alpha) + P(Z > z_\alpha) &= \alpha. \end{aligned}$$

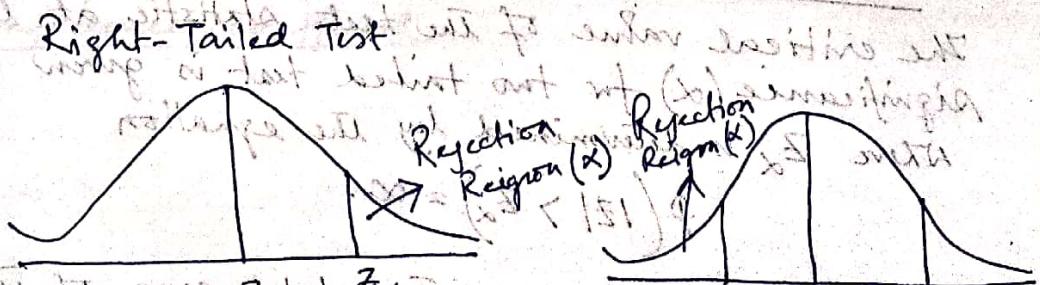
(By symmetry for the normal distribution)

In example  $\Rightarrow 2P(Z > z_\alpha) = \alpha$

~~for one-tailed test, the rejection region (with lower tail) is~~  
~~understanding,  $P(Z > z_\alpha) = \alpha/2$  because it~~

i.e. the area of each tail is  $\alpha/2$ . Thus  $z_\alpha$  is the value such that area to the right of  $z_\alpha$  and to the left of  $-z_\alpha$  is  $\alpha/2$ . (From the above figure).

In case of single-tail alternative, the critical value  $z_\alpha$  is determined so that the total area to the right of it (for right-tailed test) is  $\alpha$  and for left-tailed test the total area to the left of  $-z_\alpha$  is  $\alpha$ .



similarly for left-tailed test the area to the left of -z\_alpha is  $\alpha$

Thus we can say that, The significant or critical value of  $Z$  for a single tail-tailed test (left or right) at the level of significance ' $\alpha$ ' is same as the critical value of  $Z$  for a two-tailed test at the level of significance ' $2\alpha$ '.

The following critical values of  $Z$  at commonly used levels of significance for both two-tailed and single-tailed tests.

Critical value of ( $Z_\alpha$ ) of $Z$		
Critical values ( $Z_\alpha$ )	Level of Significance ( $\alpha$ )	
	1%	5%
Two-tailed Test	$ Z_\alpha  = 2.58$	$ Z_\alpha  = 1.96$
Right-tailed Test	$Z_\alpha = 2.33$	$Z_\alpha = 1.645$
Left-tailed Test	$Z_\alpha = -2.33$	$Z_\alpha = -1.645$

If  $n$  is small, then the sampling distribution of the test statistics  $Z$  will not be normal and in that case we can't use the above significant values, which have been obtained from the normal probability curves. In this case viz  $n$  small (usually less than 30), we use the significant values based on the exact sampling distribution of  $Z$ , which turns out to be Student's  $t$ -distribution.

Student's  $t$ -distribution is a bell-shaped curve centered around zero with some interesting features.

The shape of 'S' for large  $n$  is very similar to that of the standard normal distribution.

As  $n$  decreases, the distribution becomes more and more skewed to the left.

As  $n$  decreases, the distribution becomes more and more skewed to the right.

As  $n$  decreases, the distribution becomes more and more skewed to the left.

As  $n$  decreases, the distribution becomes more and more skewed to the right.

As  $n$  decreases, the distribution becomes more and more skewed to the left.

As  $n$  decreases, the distribution becomes more and more skewed to the right.

As  $n$  decreases, the distribution becomes more and more skewed to the left.

## Procedure for Testing of Hypothesis

We now summarise below the various steps in testing of a statistical hypothesis in a systematic manner:-

1. Null hypothesis: Set up the Null hypothesis ( $H_0$ )

2. Alternative hypothesis: Set up the alternative hypothesis ( $H_1$ ).

This enables us to decide whether we have to use single-tailed or two-tailed test.

3. Level of significance:

choose the appropriate level of significance ( $\alpha$ ). This should be decided before the sample is drawn.

4. Test statistic or Test Criterion:

compute the test statistic

$Z = \frac{t - E(t)}{S.E.(t)}$  under Null hypothesis.

5. Conclusion:

After calculating the calculated value of 'Z' in step(4) it is compared with the tabulated value of Z at the given level of significance ( $\alpha$ ).

If the calculated value of 'Z' (in absolute value) is less than the tabulated value, then we say that it is not significant and the null hypothesis is accepted. If the calculated value of 'Z' is more than the tabulated value, then  $H_0$  is rejected.

## Test for Single Proportion

The sampling distribution follows a binomial probability distribution. Hence its standard error is given by the formula  $\sqrt{NPQ}$  under null hypothesis.

Where  $n$  = sample size

$p$  = probability of success for each trial

$q = 1-p$  = Probability of failure for each trial

for large  $n$ , binomial distribution tends to normal distribution. Hence for large  $n$ ,

$$Z_1 = \frac{X - E(X)}{\sqrt{NPQ}}$$

$$\text{approx} \approx 0.0002 \text{ standard deviation}$$

Example 1: A die is thrown 324 times. It showed odd points 181 times. Would you say that die is fair?

In 324 throws of a six-faced die, odd points appeared 181 times. Would you say that die is fair?

$$\text{Sol: } X = 181, N = 324, P = \frac{1}{2}$$

Set, the hypothesis be that the die is unbiased.

The probability of getting odd points =  $\frac{1}{2}$

Expected probability number of odd points

$$\text{in 324 throws} = 324 \times \frac{1}{2} = 162$$

$$\text{Actual no. is } X - E(X) = 181 - 162 = 19$$

$$\text{Standard deviation } SE = \sqrt{NPQ} = \sqrt{324 \times \frac{1}{2} \times \frac{1}{2}} = 9$$

$$\therefore Z_1 = \frac{X - E(X)}{\sqrt{NPQ}} = \frac{19}{9} = 2.11 \text{ standard deviation}$$

Calculated value of  $Z_1$  at 5% level of significance for two tail test is  $\pm 1.96$ . Since calculated value is more than tabulated value, the hypothesis is rejected at 5% level of significance. Hence the die is biased.

### Example-2:

A die is thrown 9000 times and a throw of 3 or 4 is observed 3240 times. Show that the die cannot be regarded as an unbiased one.

Soln:

Hence, the die is unbiased.

$P$  = probability of Success

= probability of getting 3 or 4

$$= \frac{1}{6} + \frac{1}{6} = \frac{1}{3}$$

$$\therefore Q = 1 - P = 1 - \frac{1}{3} = \frac{2}{3}$$

Sample Size ( $n$ ) = 9000

$$\therefore \text{Expected number} = 9000 \times \frac{1}{3} = 3000$$

$$\text{Observed value } X - E(X) = 3240 - 3000 = 240$$

$$SE = \sqrt{NPQ} = \sqrt{9000 \times \frac{1}{3} \times \frac{2}{3}} = \sqrt{2000} = 44.73$$

$$Z = \frac{X - E(X)}{\sqrt{NPQ}} = \frac{240}{44.73} = 5.36$$

Since  $Z_{\text{calculated}} > Z_{\text{tabulated}}$  at 1%, 5%, and 10%.

level of Significance.

Hence The hypothesis is rejected at all levels

and therefore the die is almost certainly biased.

### Example-3:

In a sample of 500 from a town, 280 are tea drinkers and the rest are coffee drinkers. Can we assume that coffee and tea are equally popular in the town at 1% level of significance?

Sol<sup>n</sup>:

Let us take the hypothesis that tea and coffee are equally popular.

The expected frequency for tea and coffee drinkers

would be  $\frac{500}{2} = 250$ .

$$SE = \sqrt{NPQ}$$

$$= \sqrt{500 \times \frac{1}{2} \times \frac{1}{2}} = 11.2$$

$$\therefore Z = \frac{X - E(X)}{\sqrt{NPQ}} = \frac{280 - 250}{11.2} = 2.68$$

Tabulated value of  $Z$  at 1% level of significance for two-tailed test is 2.58. Since calculated value is more than the tabulated value, the hypothesis is rejected at 1% level of significance. Hence tea and coffee are not equally popular.

In a large sample test, we can determine whether the sample would have come from a population having a specified proportion  $p$ .

This can be done as follows:

(i) Null hypothesis

It is assumed that the given sample would have come from a population having a specified proportion  $p$ .

## LARGE SAMPLE TESTS

### Large Sample Theory:

The sample size ( $n$ ) is greater than 30 ( $n > 30$ ) it is known as a large sample. For a large sample the sampling distributions of statistic are Normal (Z test).

### Small Sample theory:

If the sample size ( $n$ ) is less than 30 ( $n < 30$ ). It is known as small sample. For small sample the sampling distributions are t, F and  $\chi^2$  distribution.

### Test of significance:

\* Test of significance for large sample.

\* Large sample test or ~~Asymptotic test~~ Asymptotic test or Z test ( $n > 30$ )

\* Test of significance for small samples ( $n < 30$ )

### Sampling from attributes:

There are two types of test for attributes -

a) Test for single proportion

b) Test for equality of two proportions

### Test for single proportion:

In a large sample size ( $n$ ), we may examine whether the sample would have come from a population having a specified proportion  $P = P_0$ .

We may proceed as follows:-

i) Null hypothesis ( $H_0$ ):

$H_0$ : The given sample would have come from a population with specified proportion  $P = P_0$ .

ii) Alternative Hypothesis ( $H_1$ ):

$$H_1: P \neq P_0 \quad (\text{Two Sided})$$

$$P > P_0 \quad (\text{one Sided - right tailed})$$

$$P < P_0 \quad (\text{one Sided - Left tailed}).$$

iii) Test statistic:

Follows the Z statistic with  
mean  $\mu = 0$  and  $\sigma^2 = 1$ .

iv) Level of Significance:

The level of significance may be fixed at either 5% or 1%.

v) Expected value or critical value:

In case of test statistic,  $Z_e$ , the expected value is

$$Z_e = 1.96 \text{ at } 5\% \text{ level of test}$$

$$= 2.58 \text{ at } 1\% \text{ level two tailed test}$$

$$Z_e = 1.65 \text{ at } 5\% \text{ level}$$

$$= 2.33 \text{ at } 1\% \text{ level of one tailed test.}$$

vi) Inference:

If the observed value of the test statistic ( $Z_0$ ) exceeds the table of ( $Z_e$ ), we reject the  $H_0$  hypothesis.

Example: If the observed value of the test statistic is 2.15, which is not exceeded by the critical value of 1.96 at 5% significance level, then we accept  $H_0$ .

Ques: A sample of size 30 has a mean of 10.5 and standard deviation of 2.5. Is there sufficient evidence to suggest that the true mean is greater than 10?

## Large Sample Tests for a Population Mean

In This section we describe and demonstrate the procedure for conducting a test of hypothesis about the mean of a population in the case sample size ( $n$ ) is at least 20. The central limit theorem states that  $\bar{X}$  is approximately normally distributed, then has Mean  $\mu_{\bar{X}} = \mu$  and standard deviation  $\sigma_{\bar{X}} = \sigma/\sqrt{n}$  where  $\mu$  and  $\sigma$  are the mean and the standard deviation of the population. This implies that the statistic

$$\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \text{ has the standard normal distribution.}$$

Two situations may arise:-

- i) If we know  $\sigma$  then the statistic is our test statistic.
- ii) If we do not know  $\sigma$  then we replace it by the sample standard deviation ( $s$ ).

Since the sample is large the resulting test statistic still has a distribution that is approximately standard normal.

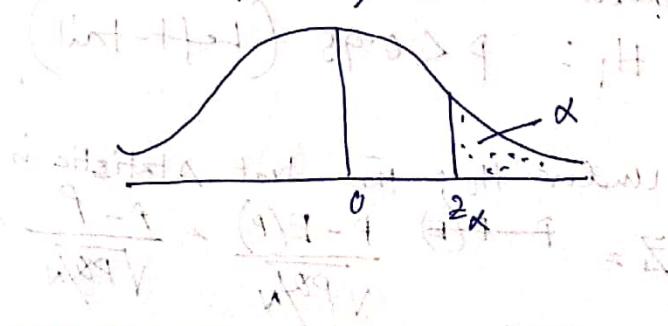
For a single population mean -  $\mu = \bar{X}$

\* If  $\sigma$  is known:  $Z = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}}$

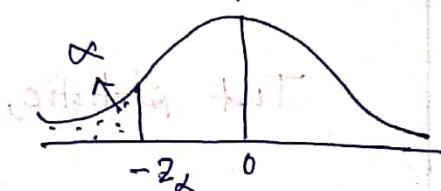
\* If  $\sigma$  is unknown:  $t = \frac{\bar{X} - \mu_0}{s/\sqrt{n}}$

The test statistic has the standard normal distribution. The distribution of the standardized test statistic and the corresponding rejection region for each form of the alternative hypothesis (left-tailed, right-tailed or two-sided tailed). This is shown in the following diagram.

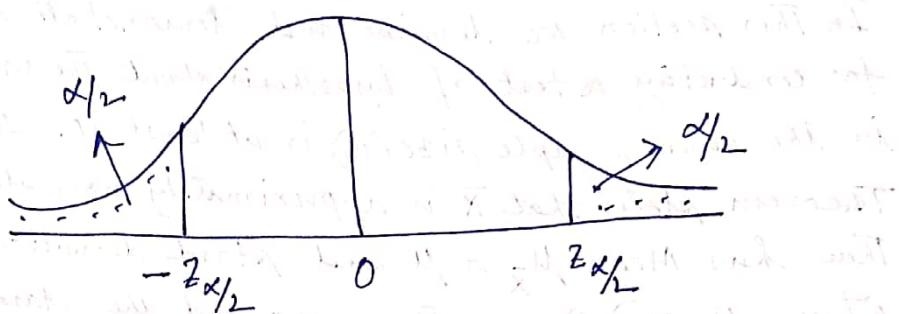
$$H_1: \mu > \mu_0$$



$$H_1: \mu \neq \mu_0$$



$$H_1: \mu \neq \mu_0 \text{ (two-tailed test)}$$



Important formulae and results

- ① A manufacturer claimed that at least 95% of the equipment which he supplied to a factory conformed to significant specifications. An examination of a sample of 200 pieces of equipment revealed that 18 were faulty. Test his claim at a significant level of (i) 0.05; (ii) 0.01.

Sol: We are given that,  $n = 200$

$$n = 200$$

$X$  = No. of pieces conforming to specifications  
in the sample =  $200 - 18 = 182$

Let  $P$  denote the probability of pieces conforming to specifications in the sample =  $\frac{182}{200}$

Null Hypothesis:  $H_0$ :  $P = 0.95$   
i.e., the proportion of pieces conforming to specifications in the lot is 95%.

Alternative Hypothesis

$$H_1: P < 0.95 \text{ (Left-tail)}$$

Test statistic,

Under  $H_0$ , the test statistic is

$$Z = \frac{P - E(P)}{\sqrt{P(1-P)/N}} = \frac{P - P}{\sqrt{P(1-P)/N}} \sim N(0)$$

$$\text{Now, } Z = \frac{0.91 - 0.95}{\sqrt{0.95 \times 0.05 / 200}} = -2.6$$

**Conclusion:**

At 5% level of significance, since the alternative hypothesis is one-tailed (left), we shall apply left-tailed test for testing significance of  $Z$ . The significant value of  $Z$  at 5% level of significance for left-tail test is  $-1.645$ . Since the computed value is less than the tabulated value, then we can say that,  $Z$  is significant and we reject the null hypothesis at 5% level of significance. Hence the manufacturer's claim is refuted or rejected at 5% level of significance.

i) Significance level at 1%:

The critical value of  $Z$  at 1% level of significance for single tailed (left tailed) test is  $-2.33$ . Since the computed value  $Z = -2.6$  is less than  $-2.33$  (i.e.  $1.21 > 2.33$ ),  $H_0$  is rejected at 1% level of significance also.

A statement like : off market estimate test

$$(1.21) = \frac{-1.9}{\sqrt{0.95 \times 0.05 / 200}}$$

cannot be aligned with reality

$$H_{0.01} = \frac{6.0 \text{ future}}{200 \times 0.05} = 6$$

As the test statistic is greater than the critical value calculated at 1% level of significance for left-tailed test, hence it is rejected at 1% level of significance. Hence the manufacturer's claim is refuted or rejected at 1% level of significance.

- Q) In a big city 325 men out of 600 men were found to be smokers. Does this information support the conclusion that the majority of men in this city are smokers? [state the hypothesis clearly]

Ans: We are given that,  $n = 600$

$$\text{No of smokers} = 325$$

Let  $p = \text{sample proportion of smokers}$

$$(\text{Sample}) = \frac{325}{600} = 0.5417$$

Null Hypothesis ( $H_0$ ):

$H_0$ : The number of smokers and non-smokers are equal in the city

So that,  $p = \text{Population proportion of smokers}$

(Population) in the city

$$= \frac{1}{2} = 0.5$$

$$\therefore Q = 1 - p = 1 - 0.5 = 0.5$$

Alternative Hypothesis ( $H_1$ ):  $P > 0.5$  (Right Tailed)

Test Statistic Under  $H_0$ ; the statistic is

$$Z = \frac{P - E(P)}{S.E.(P)} = \frac{P - P}{\sqrt{\frac{PQ}{n}}} \sim N(0, 1)$$

Since the sample is large.

$$\text{Now } Z = \frac{0.5417 - 0.5}{\sqrt{\frac{0.5 \times 0.5}{600}}} = 2.04$$

The significant value of  $Z$  for a right-tail test at 5% level of significance is 1.645. Since the calculated value of  $Z$  is greater than 1.645. It is significant and null hypothesis is rejected at 5% level of significance. Hence we conclude the majority of men in the city are smokers.

The significant value of  $Z$  for a right-right-tail test is 2.33. Since the calculated value of  $Z$  is less than 2.33, it is not significant at 1% level of significance and the null hypothesis may be accepted at 1% level of significance.

- (\*) In a sample of 400 parts manufactured by a factory, the number of defective parts was found to be 30. The company, however, claimed that only 5% of their product is defective. Is the claim tenable?

Soln. We are given that,  $n = 400$

$X = \text{No. of defectives in the sample} = 30$

Let,  $P = \text{Proportion of defectives in the sample.}$

$$\therefore P = \frac{X}{n} = \frac{30}{400} = 0.075$$

Null hypothesis ( $H_0$ ):

$$H_0: P = 0.075 \Rightarrow Q = 0.95$$

Alternative Hypothesis ( $H_1$ ):

$$H_1: P > 0.05 \text{ (Right-tailed alternative)}$$

Hence the test is one-tailed (right-tailed) test.

Test Statistics,

$\frac{\hat{P} - P}{\sqrt{\frac{PQ}{n}}}$  Under  $H_0$  for large samples,

$$\text{Standard deviation of } \hat{P} = \sqrt{\frac{P(1-P)}{n}} \approx \sqrt{\frac{PQ}{n}}$$

$$\therefore \frac{0.075 - 0.05}{\sqrt{0.05 \times 0.95 / 400}} = 2.27.$$

For right-tailed test, The significant value of  $Z$  at 5% level of significance is 1.645. Since the computed value of  $Z = 2.27$  is greater than 1.645, it is significant at 5% level of significance and hence null hypothesis is rejected. However we conclude that the company's claim of  $P$  is not tenable.

## Test of Significance for Difference of proportions

Let us compare two large populations, i.e A and B with respect to the prevalence of a certain attribute among their members.

Let us suppose we take two independent large sample size  $n_1$  and  $n_2$  from the populations A and B respectively and let  $x_1$  and  $x_2$  be the observed number of successes in these samples respectively.

Then,

$P_1 = \text{Observed proportion of Success in the sample from population A}$

$$A = \frac{x_1}{n_1}$$

$P_2 = \text{Observed proportion of Success in the sample from population B}$

$$B = \frac{x_2}{n_2}$$

If the population proportions are  $P_1$  and  $P_2$  then the corresponding sample proportions provide unbiased estimates for them,

$$E(P_1) = P_1 \text{ and } E(P_2) = P_2$$

$$\text{and } \text{Var}(P_1) = \frac{P_1(1-P_1)}{n_1} \text{ and } \text{Var}(P_2) = \frac{P_2(1-P_2)}{n_2}$$

Since for large samples,  $P_1$  and  $P_2$  are normally distributed, their difference  $P_1 - P_2$  is also normally distributed,

Then

$$E(t) = E(P_1 - P_2)$$

$$\text{and } t = P_1 - P_2$$

$$\text{var}(t) = \text{Var}(P_1 - P_2) = \text{Var}(P_1) + \text{Var}(P_2)$$

because samples are independent,  
 $\text{Var}(t) = \frac{P_1 Q_1}{n_1} + \frac{P_2 Q_2}{n_2}$

$$\text{S.E.}(t) = \sqrt{\frac{P_1 Q_1}{n_1} + \frac{P_2 Q_2}{n_2}}$$

Hence, the standardised variable corresponding to  
 $t = P_1 - P_2$  is given by —

$$Z = \frac{(P_1 - P_2) - E(P_1 - P_2)}{\text{S.E.}(P_1 - P_2)}$$

$$Z = \frac{(P_1 - P_2) - (P_1 - P_2)}{\sqrt{\frac{P_1 Q_1}{n_1} + \frac{P_2 Q_2}{n_2}}} \sim N(0, 1).$$

Student's t-test

Two sample t-test

t-statistic test

$$(1-\alpha) \text{ a.s. } \frac{s - \bar{s}}{\sqrt{\frac{(s-\bar{s})^2}{n-2}}}$$

$$\left(\frac{1}{n_1} + \frac{1}{n_2}\right)^{-1} \sqrt{s - \bar{s}} \sim \text{t-distribution}$$

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{s_{11}}{n_1} + \frac{s_{22}}{n_2}}} \sim t(n_1 + n_2 - 2)$$

$$8.9 - 8 = 0.9 \sim t(18)$$

- \* A machine puts out 16 imperfect articles in a sample of 500. After machine is overhauled it puts out 3 imperfect articles in a batch of 100. Has the machine improved.

We are given that,  $n_1 = 500$  and  $n_2 = 100$

$P_1$  = proportion of defective in the 1st sample

$$= \frac{16}{500} = 0.032$$

$P_2$  = proportion of defective in the 2nd sample

$$= \frac{3}{100} = 0.03$$

# Null hypothesis:

$$H_0: P_1 = P_2$$

i.e. There is no significant difference in the machine before overhauling and after overhauling.

Alternative hypothesis:

$$H_1: P_2 < P_1 \text{ or } P_1 > P_2$$

Test statistic:

Under  $H_0$ , since the sample are large, the test statistic is

$$Z_1 = \frac{P_1 - P_2}{S.E.(P_1 - P_2)} \sim N(0, 1)$$

$$\text{Now } S.E.(P_1 - P_2) = \sqrt{P\bar{Q}\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}$$

$$\text{Where } \hat{P} = \frac{n_1 P_1 + n_2 P_2}{n_1 + n_2} = \frac{16 + 3}{500 + 100} = \frac{19}{600} = 0.032$$

$$\hat{Q} = 1 - \hat{P} = 0.968$$

$$S.E. (P_1 - P_2) = \sqrt{0.032 \times 0.968 \left( \frac{1}{500} + \frac{1}{100} \right)}$$

$$= 0.0193$$

$$\therefore Z_1 = \frac{0.032 - 0.030}{0.0193}$$

$$= 1.04$$

Since  $Z < 1.645$  (Right tailed test), it is not significant at 5% level of significance. Hence we may accept the null hypothesis and conclude that the machine has not improved after overhauling.

\* Before an increase in excise duty on tea, 400 people out of a sample of 500 persons were found to be tea drinkers. After an increase in duty, 400 people were found tea drinkers in a sample of 600 people. Using standard error of proportion, state whether there is a significant decrease in consumption of tea.

Sol: :- We are given that,  $n_1 = 500$  and  $n_2 = 600$ .

$P_1$  = Sample proportion of tea drinkers before increase in excise duty  $= \frac{400}{500} = 0.80$

$P_2$  = Sample proportion of tea drinkers after increase in excise duty  $= \frac{400}{600} = 0.67$ .

Null Hypothesis :

$H_0 : P_1 = P_2$  i.e. There is no significant difference in the consumption of tea before and after the increase in excise duty.

Alternative Hypothesis:  $H_1 : P_1 > P_2 \Rightarrow (P_2 < P_1)$  i.e Right -tailed test.

Test Statistic:

Under the null hypothesis the test statistic is

$$Z = \frac{P_1 - P_2}{S.E.(P_1 - P_2)} = \frac{P_1 - P_2}{\sqrt{P \sigma \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim N(0,1)$$

Where

$$\hat{P} = \frac{n_1 P_1 + n_2 P_2}{n_1 + n_2} = \frac{400 + 400}{500 + 600} = \frac{8}{11}$$

$$\therefore \hat{Q} = 1 - \hat{P} = 1 - \frac{8}{11} = \frac{3}{11}$$

$$\therefore Z = \frac{0.80 - 0.67}{\sqrt{\frac{8}{11} \times \frac{3}{11} \left( \frac{1}{500} + \frac{1}{600} \right)}} = 34.81$$

Since  $Z > 1.645$ , as well as  $Z > 2.33$  (since test is one-tailed), it is highly significant at both 5% and 1% level of significance.

Hence we reject the null hypothesis,  $H_0$  and conclude that there is a significant decrease in the consumption of tea after increase in the excise duty.

$$0.8 - \frac{0.67}{500} = \text{True value is}$$

$$0.793 = \frac{0.67}{500} = \text{True value is}$$

which is significantly less than 0.80.

It is to be noted that the null hypothesis is rejected at the 1% level of significance.

and at  $(\alpha = 0.05) \Rightarrow 0.8 < 0.80$  : H<sub>0</sub> is rejected with 95%

confidence level.

It is to be noted that the null hypothesis is rejected at the 5% level of significance.

It is to be noted that the null hypothesis is rejected at the 1% level of significance.