

sw
7/4/2021

Testing:

UNIT → IV

- (1) Large Sample Test
 - Means
 - Proportions
- (2) Small Sample test
 - Means
 - Variance
- (3) Chi-square Test

Procedure for Testing Hypothesis:-

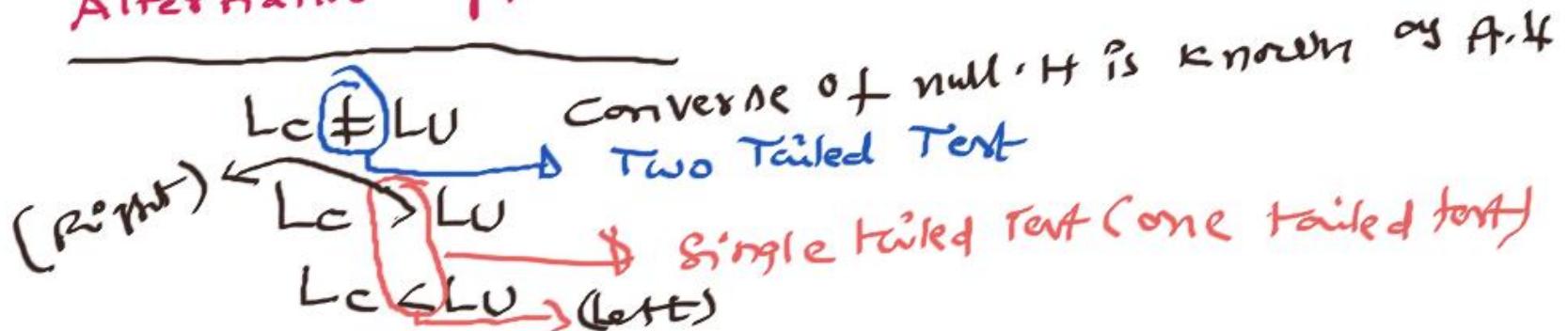
Step-i :- Hypothesis [Assumptions]

Null Hypothesis (H_0) or H_N :

$H_L = H_U$ [Region of rejection is known as N.F.]

Alternative Hypothesis (H_1 or H_A):

$A \neq B$
 $A < B$
 $A > B$



II: degrees of freedom (d.f.):

"The no of independent observations are actively participating of Testing Hypothesis".

e.g., - if size is $n \Rightarrow d.f$ is " $n-1$ "

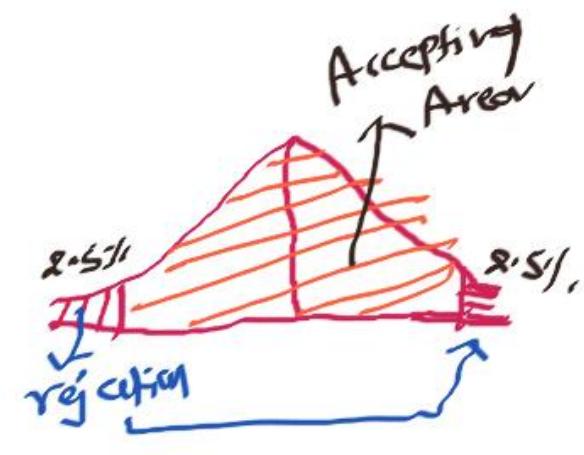
III:- Levels of Significance (L.O.S)

90%, 75%, 98%, and 99%.

↳ Accepting

10%, 5%, 2%, & 1%.

↳ rejection



IV : Test statistics

$$(n) < 30$$

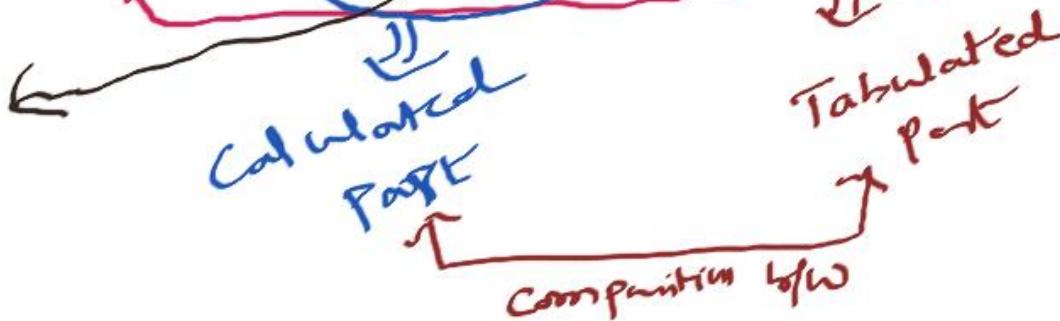
\Rightarrow Small Sample Test

If the number of observations (n) less than 30 then it is known as small sample test.

Defined as

$$t = \frac{\bar{x} - E(x)}{S \cdot E(x)} \quad df \sim T_{n-1}$$

Standard Error



$$SE = \sqrt{\text{Sample Variance}}$$

Large sample test:-

If the no. of observations $(n) \geq 30$, then the test is known as Large Sample Test

$$Z = \frac{x - E(x)}{S.d(x)} \sim N(0,1)$$

calculated part

Tabulated part

V Calculations / procedure

VI Inference / conclusion

If $\text{Calculated value} < \text{Table Value}$
at different d.f in different L.O.S our null hypothesis
(H₀) will accepted otherwise
 $\text{Calculated value} > \text{Table Value at different}$
d.f in different L.O.S our null hypothesis will be rejected

Types of Errors:-

Type-I Error:- [Producer's Risk]

Def:- Rejecting H_0 When H_0 is "true" is known as Type-I error \Rightarrow which is denoted by " α ".

Type-II error:- [Consumer's Risk]

Accepting H_0 When H_0 is "false". is known as

Type-II error

Large sample test

Table values
$$z = \frac{x - E(x)}{S.d(x)}$$

Two tailed Table value

Loss:	10%	1.645
95%	5%	1.96
98%	2%	2.33
99%	1%	2.58

Notations

	Sample	Population
Size	n	N
Mean	\bar{x}	μ
Variance	s^2	σ^2
Proportion	p	$\cdot p$

Sample:-

A part of the population is known as sample.

Estimation:-

Studying samples and drawing the conclusions about the population is known as estimation.

Types of Estimation:-

(1) Point Estimation

(2) Interval Estimation

→ By studying single observation and drawing its conclusion about populations is known as "Point Estimation"

⇒ By studying the set of observations and drawing its conclusion about a population is known as "Interval Estimation".

N

27/4/21
2.20 to 4.30 pm

Testing the significance difference in single mean

sample
single mean

Large sample
 σ -known

$$H_0: \bar{x} = \mu \quad \left[\text{i.e.: There is no significant difference b/w sample mean \& population} \right]$$

$$H_1: \bar{x} \neq \mu \quad (\text{Two tailed test})$$

LOS: at different levels.

Test Statistics:-

$$Z = \frac{\bar{x} - E(x)}{S.d} \sim N(0,1)$$

calculated

Table value

Procedure:- Let x_1, x_2, \dots, x_n are n -random samples drawn from the same normal population Mean ' M ' and variance ' σ^2/n '. $\left[\because \bar{x} \sim N(M, \sigma^2/n) \right]$

Here we are testing our N.H using Large Sample test.
But \bar{x} is single sample mean (\bar{x})

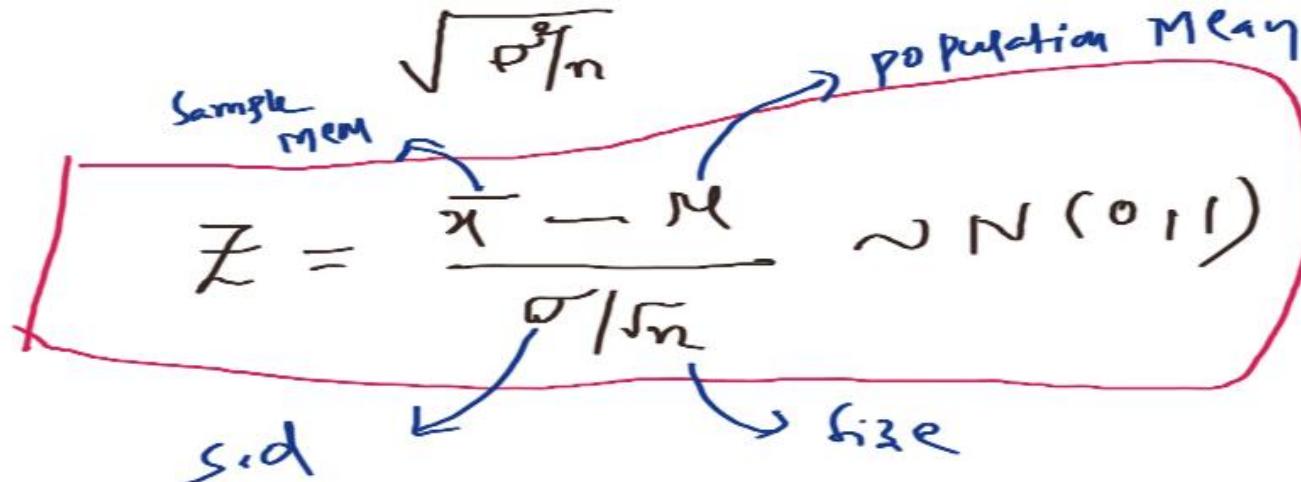
$$Z \leq \frac{\bar{x} - E(\bar{x})}{? \text{ s.d. } (\bar{x})} \sim N(0, 1)$$

Now $E(\bar{x}) = M$ $\left(\because \bar{x} \sim N(M, \sigma^2/n) \right)$

$$\text{Var}(\bar{x}) = \sigma^2/n \quad (\text{ij})$$

Finally

$$Z = \frac{\bar{x} - \mu}{\sqrt{\sigma^2/n}} \sim N(0, 1)$$



Inference:- Since $Z_{cal} < Z_{tab}$ our H_0 will accepted

otherwise rejected.

Testing i.e. significance differences b/w two sample mean
(Large sample test)

$$H_0: \mu_1 = \mu_2$$

$$H_1: \mu_1 \neq \mu_2$$

Loss:- at different levels

Test statistics:-

$$Z = \frac{x - E(x)}{S.d(x)} \sim N(0, 1)$$

Procedure :-

Let x_1, x_2 are random samples with size n_1, n_2 drawn from the two different normal population with means μ_1, μ_2 and variances $\frac{\sigma_1^2}{n_1}, \frac{\sigma_2^2}{n_2}$ respectively.

Here, testing N.H. w.r.t Large sample test

Now, q.v is difference b/w two sample means $(\bar{x}_1 - \bar{x}_2)$

Test statistics $Z = \frac{(\bar{x}_1 - \bar{x}_2) - (E(\bar{x}_1 - \bar{x}_2))}{\sqrt{\text{Var}(\bar{x}_1 - \bar{x}_2)}}$ ~ $N(0, 1)$

Let us determine $E(\bar{x}_1 - \bar{x}_2)$ & $V(\bar{x}_1 - \bar{x}_2)$

$$\begin{aligned} \text{Now } E(\bar{x}_1 - \bar{x}_2) &= E(\bar{x}_1) - E(\bar{x}_2) \\ &= \mu_1 - \mu_2 \\ &= 0 \end{aligned}$$

$$\begin{aligned} \bar{x}_1 &\sim N(\mu_1, \sigma_1^2/n_1) \\ \bar{x}_2 &\sim N(\mu_2, \sigma_2^2/n_2) \\ H_0: \mu_1 &= \mu_2 \end{aligned}$$

$$V(\bar{x}_1 - \bar{x}_2) = V(\bar{x}_1) + V(\bar{x}_2)$$

$$= \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}$$

$$\sqrt{V(\bar{x}_1 - \bar{x}_2)} = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}.$$

finally, test statistics rewritten as

$$Z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0, 1)$$

case:- $\sigma_1^2 = \sigma_2^2$

$$Z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\sigma^2 \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \sim N(0, 1)$$

Inference:- since $Z_{\text{cal}} < Z_{\text{tabulated}}$, null hypothesis
will be accepted.
otherwise rejected.



Testing the significance difference in single sample

Proportion (Large sample test)

$$H_0: p = P$$

$$H_1: p \neq P$$

(ratio)

Loss: at different levels

Test statistics:-

$$Z = \frac{x - E(x)}{S.d(x)} \sim N(0,1)$$

Procedure:- Let x_i is a random sample size in following
 Binomial dist with parameter (n, p)
 Here, we are N.H using large sample test and also
 $\eta \cdot V$ is single sample proportion $(\hat{p} = \frac{x}{n})$ $\begin{cases} E(\eta) = n \cdot p \\ V(\eta) = npq \end{cases}$

$$Z = \frac{\hat{p} - E(\hat{p})}{\sqrt{V(\hat{p})}} \sim N(0, 1)$$

Let us compute the values $E(\hat{p})$ & $V(\hat{p})$

$$\text{Now } E(\hat{p}) = E(\bar{Y}_n) = \frac{1}{n} E(x) = \frac{1}{n} (np) = P$$

$$\begin{aligned} V(\hat{P}) &= V(\bar{x}/n) = \frac{1}{n^2} V(x) \\ &= \frac{1}{n^2} (\cancel{n} PQ) = \frac{PQ}{n} \end{aligned}$$

Therefore, test statistics is

$$\boxed{\hat{Z} = \frac{\hat{P} - P}{\sqrt{\frac{PQ}{n}}} \sim N(0, 1)} \quad Q = 1 - P \quad \left(\hat{P} = \frac{x}{n} \right)$$

if 'P' is not given in the data, we should consider
 $"P = 0.5"$

Inference:- Since $Z_{\text{cal}} < Z_{\text{tab}}$, Null will be accepted. otherwise rejected.

Testing for significance differences b/w two sample proportions

(Large sample test):

$$H_0: P_1 = P_2 = P$$

$$H_1: P_1 \neq P_2$$

Loss: at different levels.

Test statistic: $Z = \frac{x - E(x)}{\text{s.d}(x)} \sim N(0, 1).$

Procedure: Let x_1, x_2 are two random samples with sizes n_1, n_2 drawn from the same Binomial population

The population parameter P, Q :

Here, we are testing Null using Large Sample test.
But our σ -v is difference b/w two sample proportion
denoted by $(P_1 - P_2)$,

Now the test statistics for Large Sample defined
as

$$Z = \frac{(\hat{p}_1 - \hat{p}_2) - E(\hat{p}_1 - \hat{p}_2)}{\sqrt{\text{Var}(\hat{p}_1 - \hat{p}_2)}} \sim N(0, 1)$$

let us find the values of $E(\hat{p}_1 - \hat{p}_2)$ and $\text{Var}(\hat{p}_1 - \hat{p}_2)$

$$E(\hat{p}_1 - \hat{p}_2) = E(\hat{p}_1) - E(\hat{p}_2)$$

$$\begin{cases} \hat{p}_1 = \frac{x_1}{n_1} \\ \hat{p}_2 = \frac{x_2}{n_2} \end{cases}$$

$$= E\left(\frac{x_1}{n_1}\right) - E\left(\frac{x_2}{n_2}\right)$$

$$= \frac{1}{n_1} E(x_1) - \frac{1}{n_2} E(x_2)$$

$$= \frac{1}{n_1} \cancel{\mu_1 p} - \frac{1}{n_2} \cancel{\mu_2 p} = 0_{II} .$$

$$\begin{aligned}
 \text{Var}(P_1 - P_2) &= \text{Var}(P_1) + \text{Var}(P_2) \\
 &= \text{Var}\left(\frac{x_1}{n_1}\right) + \text{Var}\left(\frac{x_2}{n_2}\right) \\
 &= \frac{1}{n_1^2} \text{Var}(x_1) + \frac{1}{n_2^2} \text{Var}(x_2) \\
 &= \frac{1}{n_1^2} (\cancel{n_1} PQ) + \frac{1}{n_2^2} (\cancel{n_2} PQ) \\
 &= \frac{PQ}{n_1} + \frac{PQ}{n_2} \\
 &\Rightarrow PQ \left(\frac{1}{n_1} + \frac{1}{n_2} \right)
 \end{aligned}$$

$$\text{Var}(P_1 - P_2) = PQ \left(\frac{1}{n_1} + \frac{1}{n_2} \right)$$

Therefore, the test statistics of difference b/w two proportions is

$$Z = \frac{P_1 - P_2}{\sqrt{PQ \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim N(0, 1)$$

Hence $p_1 = \frac{x_1}{n_1}$, $p_2 = \frac{x_2}{n_2}$

$$\hat{P} = \frac{x_1 + x_2}{n_1 + n_2} \text{ or } \frac{n_1 p_1 + n_2 p_2}{n_1 + n_2}$$

$$Q = 1 - P$$

Inference: - Since $Z_{\text{cal}} < Z_{\text{tab}}$, N.H will be accepted, otherwise refuted.

Problems:-

① The average life of its computer is 1150 hours and S.D is 42 hours. A random sample of 60 computer were tested whose average life is 1155 hours. Test whether overall average life is significant or not at 5% level.

Sol:- $H_0: \mu = 1150$ ^{Pop Mean}

$H_1: \mu \neq 1150$

$\text{LOS: } 0.05 \checkmark$

Test statistics:-

$$Z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \sim N(0,1)$$

Data:-

$$\bar{x} = 1155$$

$$\mu = 1150$$

$$\sigma = 42$$

$$n = 800$$

Calculation

$$Z_{\text{cal}} = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} = \frac{1155 - 1150}{42/\sqrt{800}} = 3.36$$

$$Z_{cal} = 3.36$$

$$Z_{tab} = 1.96$$

Inference:- Since $Z_{cal} > Z_{tab}$ at 5% LOS
our Null will be rejected.

i.e average life of computer is not equal to
1/50 hours ?

Prob:- Average Indian life expectancy is 72 years according ^{Medical} Statist. A random sample 50 people observed whose average life was 71.9 years and s.d is 11.5 years. Test whether average life of Indians is significant or not at 5% LOS?

$$\text{Sol:- } H_0: \mu = 72$$

$$H_1: \mu \neq 72$$

$$\text{LOS: } 5\% \text{ LOS} = 0.05$$

Test statistics:

$$Z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}} \sim N(0, 1)$$

Data :-

$$\bar{x} = 71.9 \quad \mu = 72 \quad \sigma = 11.5 \quad n = 500$$

Calculation:-

$$Z_{\text{cal}} = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}} = \frac{71.9 - 72}{11.5 / \sqrt{500}} = -0.194$$

$$|Z_{\text{cal}}| = 0.194, \dots$$

$$Z_{\text{tab}} \text{ at } 5\% \text{ LOS} = 1.96$$

$$|Z_{\text{cal}}| = 0.194$$

Inference:- Since $Z_{\text{cal}} < Z_{\text{tab}}$, our N.H will
be accepted.

i.e The average Indian life is 72 years.

③ The average length of bolt is (3.6) inch and s.d is (0.35) in, another sample average length is (3.72) inch and s.d (1.1) inch where sample size is (55) . Test whether bolt samples are drawn from same population or not?

$$\text{Sol:-- } H_0: \mu_1 = \mu_2$$

$$H_1: \mu_1 \neq \mu_2$$

LOS, 5% LOS or 0.05

Test statistics:-

$$Z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0, 1)$$

Data:-

$$\bar{x}_1 = 3.6 \quad \sigma_1 = 0.35$$

$$\bar{x}_2 = 3.72 \quad \sigma_2 = 1.1$$

$$n_1 = n_2 = 55$$

Calculation

$$Z_{\text{cal}} = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2 + \sigma_2^2}{n}}}$$

$$Z_{\text{cal}} = \frac{3.6 - 3.72}{\sqrt{\frac{(0.35)^2 + (1.1)^2}{55}}} \\ = -3$$

$$|Z_{\text{cal}}| = 3$$

$$\left. \begin{array}{l} Z_{\text{tab}} \\ \text{at } 5\% \\ \text{LOS} \end{array} \right\} = 1.96$$

Inference: Since $Z_{\text{cal}} > Z_{\text{tab}}$ at 5% LOS,
our H_0 will be rejected.

i.e both samples are not belongs same population

\bar{x} -

Single Sample Proportion:

$$Z = \frac{\hat{p} - p}{\sqrt{pq/n}} \sim N(0,1)$$

Here $p = x/n$

Two Sample Proportion:

$$Z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{pq(n_1 + n_2)}} \sim N(0,1)$$

$$\hat{p}_1 = \frac{x_1}{n_1}, \quad \hat{p}_2 = \frac{x_2}{n_2}$$

$$P = \frac{x_1 + x_2}{n_1 + n_2} = \frac{n_1 \hat{p}_1 + n_2 \hat{p}_2}{n_1 + n_2}$$

$$Q = 1 - P$$

— x —

SL
4/5/21
2:20 PM
430PM

Problems on proportion test:

A random sample of 400 items, out of 39 are defective, company claiming that 6.5% are defective - check whether company claim can be accepted or not at 5%?

- Sol
- $H_0: P = 0.065$
 - $H_1: P \neq 0.065$
 - LOS: 5% or 0.05
 - Test statistics:

$$Z = \frac{\hat{P} - P}{\sqrt{PQ/n}} \sim N(0,1)$$

Data:- $x = 39 : n = 400$

$$\begin{aligned} P &= 0.065 & Q &= 1 - 0.065 \\ &&&= 0.935 \end{aligned}$$

Cal :-

$$Z_{\text{cal}} = \frac{\hat{p} - p}{\sqrt{pq/n}} \sim N(0, 1)$$

$$\hat{p} = \frac{x}{n} = \frac{39}{400} = 0.0975$$

$$Z_{\text{cal}} = \frac{0.0975 - 0.0625}{\sqrt{\frac{0.0625 \times 0.9375}{400}}} = 3.27$$

$$\left. \begin{array}{l} Z_{\text{tab}} \\ \text{at 5% LOS} \end{array} \right\} = 1.96$$

Inference :-

Since $Z_{\text{cal}} > Z_{\text{tab}}$
at 5% LOS our
 H_0 will be rejected.
 $\therefore P \neq 0.0625$.

In a T/F a student answers ⁷⁰~~out~~ correct out of 130.
 Whether True | False question are uniformly distributed or
 not?

$$\text{Sol:- } H_0: P = 0.5$$

$$H_1: P \neq 0.5$$

Loss: 5% or 0.05

Test statistics:

$$Z = \frac{\hat{P} - P}{\sqrt{PQ/n}} \sim N(0,1)$$

data:

$$n = 130: x = 70,$$

$$\hat{P} = \frac{x}{n} = \frac{70}{130} = 0.538$$

$$P = 0.5, Q = 0.5$$

$$Z_{\text{cal}} = \frac{\hat{P} - P}{\sqrt{PQ/n}} = \frac{0.538 - 0.5}{\sqrt{\frac{0.5 \times 0.5}{130}}} \\ Z_{\text{cal}} = 0.883$$

$$Z_{\text{cal}} = 0.883$$

$$Z_{\text{tab}} \left. \begin{array}{l} \\ \text{at 5% LOS} \end{array} \right\} = 1.96$$

Inference :- Since $Z_{\text{cal}} < Z_{\text{tab}}$ at 5% LOS, our H_0 will be accepted.

i.e. The true / False question equal.

A dice rolled 350 times, out of 210 times face
3 or 4 are occurred. Check whether die is unbiased
 or not?

$$\text{Sol:- } H_0: P = \left[\frac{1}{6} + \frac{1}{6} \right] = \frac{2}{6} =$$

$$H_1: P \neq \frac{2}{6}$$

LOS: S.F. LOS

Test statistics:

$$Z = \frac{p - P}{\sqrt{\frac{pq}{n}}} \sim N(0,1)$$

data:

$$x = 210 : n = 350$$

$$p = \frac{x}{n} = \frac{210}{350} = 0.6$$

$$P = \frac{2}{6} = 0.33$$

$$Q = 0.67$$

$$Z_{\text{cal}} = \frac{p - P}{\sqrt{\frac{pq}{n}}} = \frac{0.6 - 0.33}{\sqrt{\frac{0.33 \times 0.67}{350}}} = 10.75$$

$$Z_{tab} = 1.96$$

5% LOS

Inference: $Z_{cal} > Z_{tab}$ at 5% LOS our H_0 will be rejected.

i.e die is not unbiased

— X —

A Survey is conducted regarding online examination system.
320 boys are favour out of 350 and 290 girls are favour out
of 350. Check whether opinion is uniform or not?

Sol:- $H_0: P_1 = P_2$ $H_1: P_1 \neq P_2$

Loss: 5% or 0.05^-

Test statistics:

$$Z = \frac{p_1 - p_2}{\sqrt{pq(\frac{1}{n_1} + \frac{1}{n_2})}} \sim N(0,1)$$

Data:-

$$n_1 = 320 : x_1 = 300$$

$$n_2 = 350 : x_2 = 290$$

$$p_1 = \frac{x_1}{n_1} = \frac{300}{320} = 0.937$$

$$p_2 = \frac{x_2}{n_2} = \frac{290}{350} = 0.829$$

$$\begin{aligned} P &= \frac{x_1 + x_2}{n_1 + n_2} = \frac{300 + 290}{320 + 350} \\ &= 0.880 \end{aligned}$$

$$Q = 1 - 0.880 = 0.12$$

$$\begin{aligned} Z_{cal} &= \frac{p_1 - p_2}{\sqrt{pq(\frac{1}{n_1} + \frac{1}{n_2})}} \\ &= \frac{0.937 - 0.829}{\sqrt{0.88 \times 0.12 \left(\frac{1}{320} + \frac{1}{350} \right)}} \end{aligned}$$

$$Z_{cal} = 4.32$$

$$Z_{tab} = 1.96$$

Inference: Since $Z_{\text{cal}} > Z_{\text{tab}}$ at 5% LOS: our H_0 will be rejected.

	Pro	Mutual	-x-
S ₁	36%	480	
S ₂	45%	520	

Test at 5% LOS?

Sol: $p_1 = 0.36$: $p_2 = 0.45$

$$P = \frac{n_1 p_1 + n_2 p_2}{n_1 + n_2} = \frac{480 \times 0.36 + 520 \times 0.45}{480 + 520}$$

$$= 0.4068$$

$$\theta = 1 - 0.4068 = 0.5932$$

$$Z_{\text{cal}} = \frac{P_1 - P_2}{\sqrt{PQ(P_{n_1} + P_{n_2})}} = \frac{0.36 - 0.45}{\sqrt{0.4068 \times 0.5932 \left(\frac{1}{480} + \frac{1}{520}\right)}} \\ = -2.90$$

$$|Z| = 2.90$$

$$Z_{\text{tab}} = 1.96$$

at 5%.

Inference:- Since $Z_{\text{cal}} > Z_{\text{tab}}$ at 5% LOS
our H_0 will be rejected.

Test for Significance difference in Single Sample Mean (σ unknown)

$H_0: \bar{x} = \mu$ (no significance difference b/w S.M and P.M) (Small test)
($n < 30$)

$H_1: \bar{x} \neq \mu$

df :- $n - 1$

LOS: at different levels

Test statistics:

$$T = \frac{\bar{x} - E(x)}{S \cdot E(x)} \stackrel{df}{\sim} T_{n-1}.$$

Procedure:-

Let x_1, x_2, \dots, x_n are the random samples drawn from the same normal population with mean μ and variance σ^2/n with same size n .
Here, we are testing H_0 vs H_1 using small sample test.
But \bar{x} is single sample Mean (\bar{x})

Now the test statistics is

$$T = \frac{\bar{x} - E(\bar{x})}{S \cdot E(\bar{x})} \sim t_{n-1}.$$

Let us find $E(\bar{x})$; $V(\bar{x})$

Now $E(\bar{x}) = \mu$ $(\because \bar{x} \sim N(\mu, \frac{\sigma^2}{n}))$

$V(\bar{x}) = \frac{\sigma^2}{n}$ (,,)

$$\begin{aligned} S \cdot E(\bar{x}) &= \sqrt{\sigma^2/n} \\ &= \sqrt{s^2/n} \quad (\sigma \text{ unknown}) \end{aligned}$$

$$S \cdot E(\bar{x}) = s/\sqrt{n}$$

∴ Test Statistics

$$T = \frac{\bar{x} - \mu}{s/\sqrt{n}} \stackrel{df}{\sim} t_{n-1}$$

$$\begin{cases} \bar{x} = \frac{\sum x_i}{n} \\ s^2 = \frac{\sum (x_i - \bar{x})^2}{n-1} \end{cases}$$

Inference:

Since t_{cal} value is $< t_{tab}$ at different d.f
in different LOS our H_0 will be accepted otherwise

Rejected -

Testing the significance

differences

bw

two sample Mean

(small n₁, n₂)

σ^2 Unknown

$$H_0: \mu_1 = \mu_2 : \sigma_1^2 = \sigma_2^2$$

$$H_1: \mu_1 \neq \mu_2 : \sigma_1^2 \neq \sigma_2^2$$

df :- $n_1 - 1 + n_2 - 1 = n_1 + n_2 - 2$

Los : at different levels.

Test statistics:

$$t = \frac{x - E(x)}{S \cdot E(x)} \stackrel{df}{\sim} t_{n-1}$$

Procedure:- Let x_1, x_2 are two random samples with sizes n_1, n_2 drawn from two different normal populations with means μ_1, μ_2 and variances $\frac{\sigma_1^2}{n_1}, \frac{\sigma_2^2}{n_2}$ respectively.

Here we are testing H_0 using small sample test.

But our g.v is $(\bar{x}_1 - \bar{x}_2)$

Now the test statistics is written as

$$t = \frac{(\bar{x}_1 - \bar{x}_2) - E(\bar{x}_1 - \bar{x}_2)}{S \cdot E(\bar{x}_1 - \bar{x}_2)}$$

? df $\sim t_{n_1+n_2-2}$

Let us find $E(\bar{x}_1 - \bar{x}_2)$ and $V(\bar{x}_1 - \bar{x}_2)$

$$\begin{aligned}E(\bar{x}_1 - \bar{x}_2) &= E(\bar{x}_1) - E(\bar{x}_2) \\&= \mu_1 - \mu_2 \\&= 0\end{aligned}$$

$$V(\bar{x}_1 - \bar{x}_2) = V(\bar{x}_1) + V(\bar{x}_2)$$

$$= \frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2} \quad (H_0: \sigma_1^2 = \sigma_2^2)$$

$$= \sigma^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)$$

$$= s^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right) \quad (\sigma \text{ unknown})$$

$$\bar{x}_1 \sim N(\mu_1, \frac{\sigma_1^2}{n_1})$$

$$\bar{x}_2 \sim N(\mu_2, \frac{\sigma_2^2}{n_2})$$

$$(H_0: \mu_1 = \mu_2)$$

Finally the test statistic

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s^2(\frac{1}{n_1} + \frac{1}{n_2})}} \stackrel{df}{\sim} t_{n_1+n_2-2}$$

Here $\bar{x}_1 = \frac{\sum x_{1i}}{n_1}$ $\bar{x}_2 = \frac{\sum x_{2j}}{n_2}$

s^2 (pooled sample variance) = $\frac{\sum (x_{1i} - \bar{x}_1)^2 + \sum (x_{2j} - \bar{x}_2)^2}{n_1 + n_2 - 2} = \frac{(n_1-1)s_1^2 + (n_2-1)s_2^2}{n_1+n_2-2}$

Inference:

Since $t_{cal} < t_{tab}$ at different levels in different d.f's our H_0 will be accepted otherwise rejected.

Equality of Variance (F-test)

$$H_0: \sigma_1^2 = \sigma_2^2$$

$$H_1: \sigma_1^2 \neq \sigma_2^2$$

$$\text{d.f.: } (n_1 - 1, n_2 - 1)$$

always
 $F_{\text{cal}} > 1$

Loss: at different level.

Test Statistics:

$$F = \frac{s_1^2}{s_2^2} \sim F(n_1 - 1, n_2 - 1) \quad (\text{or}) \quad F = \frac{s_2^2}{s_1^2}$$

f.t. s.t.
Nr > Dr

$$F = \frac{s_2^2}{s_1^2} \sim F(n_2 - 1, n_1 - 1)$$

Inference: Since $F_{cal} < F_{tab}$ at different d.f in different levels
our H_0 will be accepted otherwise rejected.

— * —
A company claims life of its ^{average life} computer is 1150 hours.

A random sample of 6 computer were observed whose
average lives are 1160, 1175, 1140, 1110, 1200, 1185. Test at
2.5% LOS whether company average life is significant
or not?

Sol:- $H_0: \mu = 1150$ | d.f: $n=6$
 $H_1: \mu \neq 1150$ | $n-1=5$
LOS: 0.02

Test Statistics:

$$t = \frac{\bar{x} - M}{S/\sqrt{n}} \sim t_{n-1}$$

$$\bar{x} = \frac{\sum x_i}{n}$$

$$S^2 = \frac{\sum (x_i - \bar{x})^2}{n-1}$$

Calculations:-

x_i	$(x_i - \bar{x})^2$
1160	
1175	
1140	
1110	
1200	
1185	
	$\sum (x_i - \bar{x})^2 = 5333.28$

$$\bar{x} = 1161.67$$

$$S^2 = \frac{\sum (x_i - \bar{x})^2}{n-1} = \frac{5333.28}{5} = 1066.65$$

$$S = \sqrt{1066.65} = 32.659$$

$$t_{\text{cal}} = \frac{\bar{x} - M}{S/\sqrt{n}} = \frac{1161.67 - 1150}{32.659/\sqrt{6}}$$

$$\approx 0.8752$$

$$t_{\text{tab}} \left. \begin{array}{l} \\ \end{array} \right\} = 3.365$$

$\left. \begin{array}{l} 5df \\ 0.02 \end{array} \right\}$

Inference: Since $t_{cal} < t_{tab}$ at 5% LOS in 5df our
 H_0 will be accepted. i.e. the average life computer
is 1150 hours.

A random sample of 11 students I.Q are 110, 109, 111, 108, 111, 110, 112, 113, 110, 110, 109. The average I.Q level is 111.5, Test whether it is significant or not at 2.5% LOS?

Sol:-

$$H_0: \mu = 111.5$$

$$H_1: \mu \neq 111.5$$

$$\text{d.f: } n = 11 \\ n-1 = 10$$

LOS: 0.02 or 2%.

Test Statistics

$$t = \frac{\bar{x} - \mu}{s/\sqrt{n}} \quad \text{d.f} \sim t_{n-1}$$

Cals:

$$n = 11 : \quad \bar{x} = 111.5$$

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05/5/21

JT3-

-1PM to
3.20PM

x_i	$\frac{(x_i - \bar{x})^2}{(x_i - 110.27)^2}$
110	$(110 - 110.27)^2 =$
109	
111	
108	
111	
110	
112	
113	
110	
110	$(109 - 110.27)^2 =$
109	
$\sum x_i = 1213$	
$\sum (x_i - \bar{x})^2 = 20.1819$	

$$\bar{x} = \frac{1213}{11} = 110.27$$

$$s^2 = \frac{\sum (x_i - \bar{x})^2}{n-1}$$

$$= \frac{20.1819}{10}$$

$$s^2 = 2.01819$$

$$s = \sqrt{2.01819}$$

$$= 1.4206$$

$$t_{cal} = \frac{\bar{x} - \mu}{s/\sqrt{n}}$$

$$t_{cal} = \frac{110.27 - 115.5}{1.4206 / \sqrt{11}}$$

$$= -2.8719$$

$$|t_{cal}| = 2.8719$$

$$t_{tab} \left. \begin{array}{l} t_{tab} \\ 10 d.f \\ 0.02 \\ \text{two-tailed} \end{array} \right\} = 2.764$$

Inf:- Since $t_{cal} > t_{tab}$ at 10 d.f in 0.02 L.S, H_0 will be rejected. i.e $\mu \neq 110$.

Test the equality of Means as well as Variance for following data set:

Sales	8	11	15	4	7	6	3	9
Sales after Adv	11	10	15	6	10	8	10	14

Test whether Sales are Uniform or not at 5% LOS?

$$H_0: \mu_1 = \mu_2: \sigma_1^2 = \sigma_2^2$$

$$H_1: \mu_1 \neq \mu_2: \sigma_1^2 \neq \sigma_2^2$$

$$\text{d.f.} = n_1 + n_2 - 2 = 8 + 8 - 2 = 14$$

LOS: at 5% or 0.05

Test statistics

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

Cal :-

x_1	x_2	$(x_1 - \bar{x}_1)^2$	$(x_2 - \bar{x}_2)^2$
8	11	$(x_1 - 7.875)^2$	$(x_2 - 10.5)^2$
11	10		
15	15		
4	6		
7	10		
6	8		
3	10		
9	14		
<u>63</u>	<u>84</u>	$\sum (x_1 - \bar{x}_1)^2 =$	$\sum (x_2 - \bar{x}_2)^2 =$
		$= 105.0154$	$= 60$

$$\bar{x}_1 = \frac{63}{8} = 7.875$$

$$\bar{x}_2 = \frac{84}{8} = 10.5$$

$\checkmark s_1^2 = \frac{\sum (x_1 - \bar{x}_1)^2}{n-1}$

$$= \frac{105.0154}{7} =$$

$s_1^2 = 15.022 \text{ //}$

$\checkmark s_2^2 = \frac{\sum (x_2 - \bar{x}_2)^2}{n-1}$

$s_2^2 = \frac{60}{7} = 8.5714 \text{ //}$

$$s^2 = \frac{\sum (x_1 - \bar{x}_1)^2 + \sum (x_2 - \bar{x}_2)^2}{n_1 + n_2 - 2}$$

$$= \frac{105.0154 + 60}{14} = 11.786$$

$$t_{cal} = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s^2(\frac{1}{n_1} + \frac{1}{n_2})}} = \frac{7.875 - 10.5}{\sqrt{11.786 (\frac{1}{8} + \frac{1}{8})}} = -1.529$$

$$|t_{cal}| = 1.529$$

$$\left. \begin{array}{l} t_{tab} \text{ at } 14 \text{ df} \\ 0.05 \text{ LOS} \end{array} \right\} = 2.145$$

at 2 tailed

Inf:- since $t_{cal} < t_{tab}$ at
14 df in 0.05 LOS, H_0

will be accepted.

i.e. there no significance
difference b/w the population
means (Sales - Uniform)

F-test:

$$F = \frac{s_1^2}{s_2^2} \text{ at } F(n_1-1, n_2-1) \\ \text{Nr} > \text{Dr}$$

$$F_{cal} = \frac{s_1^2}{s_2^2} = \frac{15.022}{8.5714} = 1.7525$$

$$\left. \begin{array}{l} F_{tab}(7,7) \\ \text{at 5% LOS} \end{array} \right\} = 3.79$$

Inf:- since $F_{cal} < F_{tab}$ at (7,7) df
in 0.05 LOS, H_0 will be accepted.

Test the equality of Means and equality Variances for the following data at 5% LOS:

x_1	7	14	3	8	2	6	5	1	4
x_2	8	3	6	2	1	7	9		

$$\text{Sol:- } H_0: \mu_1 = \mu_2: \sigma_1^2 = \sigma_2^2$$

$$H_1: \mu_1 \neq \mu_2: \sigma_1^2 \neq \sigma_2^2$$

$$\text{d.f: } n_1 + n_2 - 2 = 9 + 7 - 2 \\ = 14$$

LOS: 0.05

Te

Test Statistics:-

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s^2(\frac{1}{n_1} + \frac{1}{n_2})}} \text{ df } \sim t_{n_1+n_2-2}$$

x_1	x_2	$(x_1 - \bar{x}_1)^2$	$(x_2 - \bar{x}_2)^2$
7	8	$(x_1 - 5.55)^2$	$(x_2 - 5.14)^2$
14	3		
3	6		
8	2		
2	1		
6	7		
5	9		
1			
4			
$\sum (x_1 - \bar{x}_1)^2 =$		$\sum (x_2 - \bar{x}_2)^2 =$	
$= 122.22$		$= (58.85)$	

$$\bar{x}_1 = \frac{50}{9} = 5.55$$

$$\bar{x}_2 = \frac{36}{7} = 5.14$$

$$s_1^2 = \frac{\sum (x_1 - \bar{x}_1)^2}{n_1 - 1} = \frac{122.22}{8} = 15.277$$

$$s_2^2 = \frac{\sum (x_2 - \bar{x}_2)^2}{n_2 - 1} = \frac{58.85}{6} = 9.8095$$

$$s^2 = \frac{\sum (x_1 - \bar{x}_1)^2 + \sum (x_2 - \bar{x}_2)^2}{n_1 + n_2 - 2} = \frac{122.22 + 58.85}{14} = 12.93$$

$$\therefore t_{\text{cal}} = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s^2(\frac{1}{n_1} + \frac{1}{n_2})}} = \frac{5.55 - 5.14}{\sqrt{12.933 \left(\frac{1}{9} + \frac{1}{7} \right)}} = 0.226$$

$$t_{\text{cal}} = 0.226$$

$$\left. \begin{array}{l} t_{\text{tab}} \\ \text{at } 14 \text{ df} \\ \text{in } 0.05 \end{array} \right\} = 2.145$$

Inf:- Since $t_{\text{cal}} < t_{\text{tab}}$
 at 14 d.f in 0.05 LOS
 our H_0 will be accepted
 So, no significance difference
 b/w population Means

F-test:

$$F = \frac{s_1^2}{s_2^2} \sim F(n_1-1, n_2-1) \quad (n_r > D_r)$$

$$F_{\text{cal}} = \frac{15.277}{9.8095} = 1.557$$

$$\left. \begin{array}{l} F_{\text{tab}} \\ (9-1, 7-1) \\ \text{at } 5\% \end{array} \right\} = 4.15$$

Inf:- Since $F_{\text{cal}} < F_{\text{tab}}$ at
 (8, 6) d.f in 0.05 LOS, H_0 will be
 accepted.

Test the equality of Means and Equality of Variance
following results at 5% LOS:

	Size	Sum of the observation	Sum of Squares of observation from Mean
S_1	9	450	1785
S_2	11	510	2145

Sol:-

$$H_0: \mu_1 = \mu_2 : \sigma_1^2 = \sigma_2^2$$

$$H_1: \mu_1 \neq \mu_2 : \sigma_1^2 \neq \sigma_2^2$$

$$df: n_1 + n_2 - 2 = 9 + 11 - 2 = 18$$

LOS: 5% LOS

Test statistics:

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s^2 \left(\frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim t_{n_1 + n_2 - 2}$$

data:

$$\left[\begin{array}{lll} n_1 = 9 & \sum x_{1i} = 450 & \sum (x_{1i} - \bar{x}_1)^2 = 1785 \\ n_2 = 11 & \sum x_{2j} = 510 & \sum (x_{2j} - \bar{x}_2)^2 = 2145 \end{array} \right]$$

$$\bar{x}_1 = \frac{\sum x_{1i}}{n_1} = \frac{450}{9} = 50$$

$$\bar{x}_2 = \frac{\sum x_{2j}}{n_2} = \frac{510}{11} = 46.36$$

$$s_1^2 = \frac{\sum (x_{1i} - \bar{x}_1)^2}{n_1 - 1} = \frac{1785}{8} = 223.125$$

$$s_2^2 = \frac{\sum (x_{2j} - \bar{x}_2)^2}{n_2 - 1} = \frac{2145}{10} = 214.5$$

$$s^2 = \frac{\sum (x_{1i} - \bar{x}_1)^2 + \sum (x_{2j} - \bar{x}_2)^2}{n_1 + n_2 - 2}$$

$$= \frac{1785 + 2145}{18}$$

$$= 218.333$$

.

$$t_{cd} = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s^2(\frac{1}{n_1} + \frac{1}{n_2})}} = \frac{50 - 46.36}{\sqrt{218.33 \left(\frac{1}{9} + \frac{1}{11} \right)}} = 0.548$$

$$\left. \begin{array}{l} t_{tab} \\ \text{at } 18 \text{ df} \\ 0.05 \end{array} \right\} = 2.101$$

Inf:- Since $t_{cd} < t_{tab}$ at
18 df in 0.05 LOS, H_0 will
be accepted.

F-test:-

$$F = \frac{s_1^2}{s_2^2} \sim F(n_1-1, n_2-1)$$

$n_1 > n_2$

$$F_{cal} = \frac{s_1^2}{s_2^2} = \frac{223.125}{214.5} = 1.0402$$

$$\left. \begin{array}{l} F_{tab}(9-1, 11-1) \\ \text{at } 0.05 \end{array} \right\} = 3.07$$

Inf:- Since $F_{cal} < F_{tab}$ at
(8, 10) df in 0.05 LOS H_0 will
be accepted.

Chi-square Test (χ^2 -test) (Goodness of fit)

$$H_0: P_1 = P_2 = P_3 = \dots = P_n$$

$$H_1: P_1 \neq P_2 \neq P_3 \neq \dots \neq P_n$$

d.f.: - $n-1$

$$(r-1 \times c-1)$$

↓ ↓
 no of rows no of columns

Los: - at different levels

Test statistics:-

$$\chi^2 = \sum_{i=1}^n \frac{(O_i - E_i)^2}{E_i}$$

$\sim \chi^2(n-1)$

O - observed freq's
(given data)

E: Expected freq
(estimated freq)

χ^2 -test table:

O_i	E_i	$(O_i - E_i)^2$	$\frac{(O_i - E_i)^2}{E_i}$
			$\sum \frac{(O_i - E_i)^2}{E_i} =$

Total of the last column
itself calculated value

We should
learn how
find expected

Inference :- Since $\chi^2_{cal} < \chi^2_{tab}$ at different df in different levels our H_0 will be accepted otherwise rejected.

def → Chi-square:-

Sum of the Squares of Standard normal η_i 's
is known as Chi-square χ^2 -Variable.

Methods:-

1) χ^2 -test for independent attributes

General
Method

fitting of the
disn

(2) χ^2 -test for

$R \times C$ tables

They have different procedures for
Computing expected values.

Method - I :-

χ^2 - test for independence of attributes:

$$\text{Test statistics } \} = \sum \frac{(O_i - E_i)^2}{E_i} \underset{\text{df}}{\sim} \underline{\chi^2_{m-1}}$$

$$\text{Here } E_i = \text{Expected value} = \frac{\sum O_i}{n} = \overline{O}$$

i.e "Every observed value has same expected value"

- x -

A die rolled, the following table shows different face value and freq, Test whether die is unbiased or not at 5% LOS

$\sum f_i = 750$
 $T_{T_3} = 1115 \neq 21$
 $2.20 \text{ to } 4.30$

face value	1	2	3	4	5	6
freq	110	115	140	165	95	125

Sol:-

$$\sum f_i = 750$$

$$\bar{x} = \frac{\sum f_i}{6} = 125$$

$$\bar{o} = 125$$

$$\text{Expected freq (e}_i^2) = 125$$

o	E	$(o - E)^2$	$\frac{(o - E)^2}{E_i}$
110	125	225	1.8
115	125	100	0.8
140	125	225	1.8
165	125	1600	12.8
95	125	900	7.2
125	125	0	0
<u>24.4</u>			

$$\chi^2_{\text{cal}} = 24.4$$

$$\left. \begin{array}{l} \chi^2_{\text{tab 5 d.f}} \\ 0.05 \text{ LOS} \end{array} \right\} = 11.070$$

Inference:- Since $\chi^2_{\text{cal}} > \chi^2_{\text{tab}}$ at 5 d.f in 0.05 LOS
our H_0 will be rejected.
i.e Given die is not unbiased

-x-

The aircraft accidents during the weekdays are given below. Test whether accidents are uniform or not at 5% LOS $\frac{1}{2}$

day	M	T	W	Th	F	Sat	Sun
f_i	3	5	2	1	4	4	5

Sol:-

$$\sum f_i = \sum o_i = 24$$

$$\bar{o} = \frac{24}{7} = 3.42$$

o_i	ε	$(o - \varepsilon)^2$	$\frac{(o - \varepsilon)^2}{\varepsilon_i}$
3	3.42	0.1764	0.0515
5	3.42	2.4964	0.7299
2	3.42	2.016	0.5894
1	3.42	5.8564	1.7123
4	3.42	0.1764	0.0515
4	3.42	0.1764	0.0515
5	3.42	2.4964	0.7299

$$\chi^2_{\text{cal}} = 3.916$$

$$\left. \begin{array}{l} \chi^2_{\text{tab}} \\ 6 \text{ d.f} \\ 0.05 \end{array} \right\} = 12.592$$

Inference:- since $\chi^2_{\text{cal}} < \chi^2_{\text{tab}}$ at 6 d.f in 0.05
our H_0 will accepted. i.e., aircraft accidents are
uniform.

Method : 2 Test the Chi-square Goodness of fit by fitting
data with Poisson Dist:

x	0	1	2	3	4	5	6	
f	1150	900	749	550	410	295	78	$N = 4132$

$$P(x) = \frac{e^{-\lambda} \lambda^x}{x!} \quad \lambda > 0$$

Sol:- Table - I (Mean Table)

x	f	fx	$\sum f_x = 7631$	$\bar{x} = \frac{\sum f_x}{\sum f} = \frac{7631}{4132}$
.	.	.	$\sum f = 4132$	$= 1.846$

Table - 2:- Fitting of the poisson distⁿ

x	f	$P(x) = \frac{e^{-\lambda} \cdot \lambda^x}{x!}$	$(\lambda = 1.641)$	[To find expected values]
0	1150	$P(0) = \frac{e^{-1.641} (1.641)^0}{0!} = 0.1578$		$E(0) = 4132 \times 0.1578 = 652.02$
1	900	$P(1) = \frac{P(0) \cdot \lambda}{1!} = 0.2914$		$E(1) = 1244.06$
2	749	$P(2) = \frac{P(1) \cdot \lambda}{2!} = 0.2689$		$E(2) = 1111.09$
3	550	$P(3) = \frac{P(2) \cdot \lambda}{3!} = 0.1655$		$E(3) = 683.846$
4	410	$P(4) = \frac{P(3) \cdot \lambda}{4!} = 0.0763$		$E(4) = 314.03$
5	295	$P(5) = \frac{P(4) \cdot \lambda}{5!} = 0.0281$		$E(5) = 116.109$
6	178	$P(6) = \frac{P(5) \cdot \lambda}{6!} = 0.0086$		$E(6) = 35.532$

Table 3:- Chi-square goodness of fit

O	E	$(O - E)^2$	$\frac{(O - E)^2}{E_i}$	
1150	652.02	247984.08	380.332	
900	1204.06	92452.4836	76.783	
749	1111.09		118.0004	
550	683.846		26.197	
410	314.03		29.329	
295	116.109		275.620	
78	33.5352		58.956	
$\sum \frac{(O - E)^2}{E_i} = 965.2174$				

1

$$\therefore \chi^2_{\text{cal}} = 965.2174$$

$$\left. \begin{array}{l} \chi^2_{\text{tab}} \\ 6 \text{ d.f} \\ 0.05 \end{array} \right\} = 12.592$$

Inference:- Since $\chi^2_{\text{cal}} > \chi^2_{\text{tab}}$ at 6 d.f in
0.05 LOS our H_0 will be rejected.

-x-

Fit the poisson dist and test the chisquare goodness of fit.

x	0	1	2	3	4	5	6	.
f	850	710	640	520	445	310	98	N = 3573

Sol:- Mean Table

x	f	fx

$$\sum fx = 8318$$

$$\sum f = 3573$$

$$\left| \begin{array}{l} \bar{x} = \frac{\sum fx}{N} = \frac{8318}{3573} \\ \quad \quad \quad = 2.380 \\ \bar{x} = \lambda = 2.380 \end{array} \right.$$

Table (2) :- fitting of the Poisson Disⁿ

x	f	$P(x) = \frac{e^{-\lambda} \cdot \lambda^x}{x!}$	$\lambda = 2.380$	$E(x) = N \cdot P(x)$
0		$P(0) = \frac{e^{-2.380} (2.380)^0}{0!} = 0.0925$		$E(0) = 333.683$
1		$P(1) = \frac{P(0) \cdot \lambda}{1} = 0.2206$		$E(1) = 786.595$
2		$P(2) = \frac{P(1) \cdot \lambda}{2} = 0.2619$		$E(2) = 936.049$
3		$P(3) = \frac{P(2) \cdot \lambda}{3} = 0.20784$		$E(3) = 742.376$
4		$P(4) = \frac{P(3) \cdot \lambda}{4} = 0.1236$		$E(4) = 441.556$
5		$P(5) = \frac{P(4) \cdot \lambda}{5} = 0.0588$		$E(5) = 210.212$
6		$P(6) = \frac{P(5) \cdot \lambda}{6} = 0.0233$		$E(6) = 83.336$

Table -3:- Chi-square Table

O_i	E_i	$(O_i - E_i)^2$	$\frac{(O_i - E_i)^2}{E_i}$
850	333.683		7.98.911
710	786.595		7.458
640	936.049		93.632
520	142.376		66.611
445	441.556		0.0268
310	210.212		0.0564
98	83.336		9.580
		$\sum \frac{(O_i - E_i)^2}{E_i} =$	969.975

$$\therefore \chi^2_{\text{cal}} = 969.275$$

$$\left. \begin{array}{l} \chi^2_{\text{tab}} \\ 6 \text{ d.f} \end{array} \right\}_{0.05} = 12.592$$

Inf:- Since $\chi^2_{\text{cal}} > \chi^2_{\text{tab}}$ at 6 d.f in
0.05 LOS our H_0 will be rejected.

METHOD-II:- χ^2 method ($r \times c$ tables)

Test statistics:-

$$\chi^2 = \sum \frac{(O_i - E_i)^2}{E_i} \sim \chi^2_{(r-1) \times (c-1)}$$

↓
no of rows ↓
 no of columns

2x2 Contingency table:-

		S-II		Row Total	Grand Total =
		A	B		
S-I	A	a	b	a+b	= Sum Row totals =
	B	c	d	c+d	= Sum Column totals =
Column Total		a+c	b+d	$a+b+c+d=N$	

How find the expected freal (rxc method)

$$E(a) = \frac{R \cdot T \times C \cdot T}{G \cdot T} = \frac{(a+b) \times (a+c)}{a+b+c+d} = \frac{(a+b) \times (a+c)}{N}$$

$$E(b) = \frac{R \cdot T \times C \cdot T}{G \cdot T} = \frac{(a+b) \times (b+d)}{N}$$

$$E(c) = \frac{R \cdot T \times C \cdot T}{G \cdot T} = \frac{(C+q) \times (a+c)}{N}$$

$$E(d) = \frac{R \cdot T \times C \cdot T}{G \cdot T} = \frac{(C+d) \times (b+d)}{N}$$

.

Test the Chi-square Goodness of fit at 5% LOS for the following cells?

	α	β
I	5	15
II	10	9
III	4	6

Sol:-

	α	β	RT
I	5	15	20
II	10	9	19
III	4	6	10
	19	30	$49 = N$

Let us find expected freais

$$E(5) = \frac{20 \times 19}{49} = 7.7551$$

$$E(15) = \frac{20 \times 30}{49} = 12.244$$

$$E(10) = \frac{19 \times 19}{49} = 7.36$$

$$E(9) = \frac{19 \times 30}{49} = 11.6326$$

$$E(4) = \frac{10 \times 19}{49} = 3.8775$$

$$E(6) = \frac{10 \times 30}{49} = 6.122$$

Chi Square Table:

O	E	$(O - E)^2$	$\frac{(O - E)^2}{E}$
5	7.9551		0.9787
15	12.244		0.5191
10	7.36		0.9465
9	11.6326		0.595
4	3.8775		0.0038
6	6.122		0.0024
			3.1455

$$\therefore \chi^2_{\text{cal}} = 3.1445$$

$$\begin{aligned} \chi^2_{\text{tab}} \\ (3-1) \times (2-1) = 2 \text{ df} \end{aligned} \left. \right\} = 5.991$$

at 0.05 LOS

Infer: since $\chi^2_{\text{cal}} < \chi^2_{\text{tab}}$ at 2 d.f in
0.05 LOS our H_0 will be accepted.

test the Chi-square Goodness of fit for the following data at 5% LOS?

	α	β	γ
A	18	21	19
B	15	17	10
C	11	8	10

Sol:-

	α	β	γ	R.T
A	19.78	20.68	17.53	58
B	14.32	14.97	12.69	42
C	19.89	10.34	8.76	29
C.T	44	46	39	129 = N

50

$$\left. \begin{array}{l} T_{13} \\ 12 | 5 | 21 \\ 1.20 | 50 \\ 3.308^m \end{array} \right\}$$

$$E(18) = \frac{58 \times 44}{129} = 19.78$$

$$E(21) = \frac{58 \times 46}{129} = 20.68$$

$$E(19) = \frac{58 \times 39}{129} = 17.53$$

$$E(15) = \frac{42 \times 44}{129} = 14.32$$

$$E(17) = \frac{42 \times 46}{129} = 14.97$$

$$E(10) = \frac{42 \times 39}{129} = 12.69$$

$$E(11) = \frac{29 \times 44}{129} = 9.89$$

$$E(18) = \frac{29 \times 46}{129} = 10.34$$

$$E(10) = \frac{29 \times 39}{129} = 8.76$$

Chi-square Table:-

O_i	E_i	$(O - E)^2$	$\frac{(O - E)^2}{E}$
18	19.78		0.1610
21	20.68		0.0049
19	17.53		0.1232
15	14.32		0.0322
17	14.97		0.2752
10	12.69		0.5702
11	9.89		0.1245
8	10.34		0.5295
10	8.76		0.1755

$$\chi^2_{\text{cal}} = \sum \frac{(O_i - E_i)^2}{E_i} = 1.9962$$

$$\left. \begin{array}{l} \chi^2_{\text{tab}} \\ (3-1) \times (3-1) = 4 \\ \text{at } 0.05 \end{array} \right\} = 9.488$$

Inference:- Since $\chi^2_{\text{cal}} < \chi^2_{\text{tab}}$ at 4 d.f in 0.05 LOS
our H₀ will be accepted.
(i.e) there is no significance difference within the cells.

Key points:-

1. Test statistics for single mean (σ known) $\Rightarrow Z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \sim N(0,1)$
2. Test statistics for difference of two mean (σ known):

$$Z = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0,1)$$

$$\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

3. Test Matrices for single sample proportion

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

4. Test statistics difference b/w two sample proportions.

$$Z = \frac{p_1 - p_2}{\sqrt{pq(\frac{1}{n_1} + \frac{1}{n_2})}} \sim N(0,1) \quad \left| \begin{array}{l} p_1 = \frac{x_1}{n_1}, \quad p_2 = \frac{x_2}{n_2} \\ p = \frac{x_1 + x_2}{n_1 + n_2} \sim \frac{n_1 p_1 + n_2 p_2}{n_1 + n_2} \\ (q = 1 - p) \end{array} \right.$$

5. Test statistics for single sample mean (σ unknown)

$$t = \frac{\bar{x} - \mu}{s/\sqrt{n}} \stackrel{df}{\sim} t_{n-1} \quad s = \sqrt{\frac{\sum(x_i - \bar{x})^2}{n-1}}$$

6. Test statistics for difference b/w two sample Mean (σ unknown)

$$t = \frac{\bar{x}_1 - \bar{x}_2}{\sqrt{s^2(\frac{1}{n_1} + \frac{1}{n_2})}} \stackrel{df}{\sim} t_{n_1 + n_2 - 2} \quad \left| \begin{array}{l} s^2 = \frac{\sum(x_{1,i} - \bar{x}_1)^2 + \sum(x_{2,i} - \bar{x}_2)^2}{n_1 + n_2 - 2} \end{array} \right.$$

$$S^2 = \frac{(n_1 - 1) s_1^2 + (n_2 - 1) s_2^2}{n_1 + n_2 - 2}$$

7. F-test :-

$$F = \frac{s_1^2}{s_2^2} \sim F(n_1 - 1, \nu_1, \nu_2) \quad (N_r > D_r)$$

$$= \frac{s_2^2}{s_1^2} \sim F(n_2 - 1, \nu_1, \nu_2)$$

8. Chi-square test (Goodness fit)

(a) $\chi^2 = \sum \frac{(O - E)^2}{E_i} \sim \chi^2_{n-1}$ [Independent attributes]

(b) $\chi^2 = \sum \frac{(O - E)^2}{E} \sim \chi^2_{(r-1)(c-1)}$ [r x c]