

Unit-5 (Lec-2)

(Maths-4)

Topic: Student t-test of Significance (Part-1)

Test of Significance ✓

→ t-test ✓

→ F-Test ✓

→ Chi-square Test (χ^2 -Test) ✓

→ One way ANOVA ✓

Test of Significance of Small samples: →

When sample size less than 30 ✓ ⇒ small sample.

t-distribution is used when

* Sample size ≤ 30 ✓

* Population S.D. is unknown

$$\rightarrow \boxed{t = \frac{\bar{x} - \mu}{s/\sqrt{n}}}, S = \sqrt{\frac{\sum (x - \bar{x})^2}{n-1}}$$

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

$\bar{x}$ → sample mean ✓

μ → Population mean ✓

n → sample size ✓

S → S.D. of Population ✓

If S.D. of sample is given 's' then

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

✓ s and S relation is

$$(n-1) \underline{S^2} = \underline{s^2 n}$$

✓ t-test for single sample Mean ✓

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

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

$$\checkmark \quad \underline{v} = \underline{d.f.} = \underline{\checkmark n - 1}$$

↳ degree of freedom

d.f. = $\underline{v} = \checkmark n - k$, k is no. of independent constraints

t-test for two sample means

$$\checkmark \quad t_{cal} = \frac{(\checkmark \bar{x} - \checkmark \bar{y})}{S \sqrt{\frac{1}{\checkmark n_1} + \frac{1}{\checkmark n_2}}}$$

$\checkmark \bar{x}$ = mean of sample 1

$\checkmark \bar{y}$ = " " " " 2

$\checkmark n_1, \checkmark n_2$ size of sample 1 and sample 2

$$S^2 = \frac{n_1 \checkmark s_1^2 + n_2 \checkmark s_2^2}{n_1 + n_2 - 2} = \frac{\sum (\checkmark x_i - \checkmark \bar{x})^2 + \sum (\checkmark y_i - \checkmark \bar{y})^2}{n_1 + n_2 - 2}$$

$$\checkmark \quad v_1 = n_1 - 1, \quad v_2 = n_2 - 1$$

$$\underline{v = n_1 + n_2 - 2}$$

Note: - $\checkmark |t_{cal}| < t_{tab} \Rightarrow$ Null hypothesis H_0 accepted

$\checkmark |t_{cal}| > t_{tab} \Rightarrow$ Null hypothesis H_0 rejected

$\left. \begin{array}{l} t \\ v = \\ 5\% \rightarrow .05 \end{array} \right\} t_{tab} (.05)$ level of sig. $\rightarrow (5\%, 1\%)$
 $\uparrow$
 (By default)

STATISTICAL TABLES

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TABLE A.2

t Distribution: Critical Values of t

Degrees of freedom	Two-tailed test: One-tailed test:	Significance level					
		10% 5%	5% 2.5%	2% 1%	1% 0.5%	0.2% 0.1%	0.1% 0.05%
1		6.314	12.706	31.821	63.657	318.309	636.619
2		2.920	4.303	6.965	9.925	22.327	31.599
3		2.353	3.182	4.541	5.841	10.215	12.924
4		2.132	2.776	3.747	4.604	7.173	8.610
5		2.015	2.571	3.365	4.032	5.893	6.869
6		1.943	2.447	3.143	3.707	5.208	5.959
7		1.894	2.365	2.998	3.499	4.785	5.408
8		1.860	2.306	2.896	3.355	4.501	5.041
9		1.833	2.262	2.821	3.250	4.297	4.781
10		1.812	2.228	2.764	3.169	4.144	4.587
11		1.796	2.201	2.718	3.106	4.025	4.437
12		1.782	2.179	2.681	3.055	3.930	4.318
13		1.771	2.160	2.650	3.012	3.852	4.221
14		1.761	2.145	2.624	2.977	3.787	4.140
15		1.753	2.131	2.602	2.947	3.733	4.073
16		1.746	2.120	2.583	2.921	3.686	4.015
17		1.740	2.110	2.567	2.898	3.646	3.965

18	1.734	2.101	2.552	2.878	3.610	3.922
19	1.729	2.093	2.539	2.861	3.579	3.883
20	1.725	2.086	2.528	2.845	3.552	3.850
21	1.721	2.080	2.518	2.831	3.527	3.819
22	1.717	2.074	2.508	2.819	3.505	3.792
23	1.714	2.069	2.500	2.807	3.485	3.768
24	1.711	2.064	2.492	2.797	3.467	3.745
25	1.708	2.060	2.485	2.787	3.450	3.725
26	1.706	2.056	2.479	2.779	3.435	3.707
27	1.703	2.052	2.473	2.771	3.421	3.690
28	1.701	2.048	2.467	2.763	3.408	3.674
29	1.699	2.045	2.462	2.756	3.396	3.659
30	1.697	2.042	2.457	2.750	3.385	3.646
32	1.694	2.037	2.449	2.738	3.365	3.622
34	1.691	2.032	2.441	2.728	3.348	3.601
36	1.688	2.028	2.434	2.719	3.333	3.582
38	1.686	2.024	2.429	2.712	3.319	3.566
40	1.684	2.021	2.423	2.704	3.307	3.551
42	1.682	2.018	2.418	2.698	3.296	3.538
44	1.680	2.015	2.414	2.692	3.286	3.526
46	1.679	2.013	2.410	2.687	3.277	3.515
48	1.677	2.011	2.407	2.682	3.269	3.505
50	1.676	2.009	2.403	2.678	3.261	3.496
60	1.671	2.000	2.390	2.660	3.232	3.460
70	1.667	1.994	2.381	2.648	3.211	3.435
80	1.664	1.990	2.374	2.639	3.195	3.416
90	1.662	1.987	2.368	2.632	3.183	3.402
100	1.660	1.984	2.364	2.626	3.174	3.390
120	1.658	1.980	2.358	2.617	3.160	3.373

Ex. The lifetime of electric bulbs for a random sample of 10 from a large consignment gave the following data : ✓ n=10

Item	1	2	3	4	5	6	7	8	9	10
Life in '000 hrs	4.2	4.6	3.9	4.1	5.2	3.8	3.9	4.3	4.4	5.6

Can we accept the hypothesis that the average lifetime of bulb is 4000 hrs ?

Sol ✓ $H_0: \mu = 4000 \text{ hrs}$ There is no significant diff b/w sample mean and Population Mean n=10
 $H_1: \mu \neq 4000 \text{ hrs}$ (Two tailed Test)

x	$x - \bar{x} = 2.44$	$(x - 4.4)^2$
4.2	-.2	.04
4.6	.2	.04
3.9	-.5	.25
4.1	-.3	.09
5.2	.8	.64
3.8	-.6	.36
3.9	-.5	.25

$$S = \sqrt{\frac{3.12}{9}}$$

$$S = .589$$

$$t = \frac{\bar{x} - \mu}{S/\sqrt{n}}$$

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

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

$$\bar{x} = \frac{44}{10} = 4.4$$

4.3	- .1	.01
4.4	+ 0	0
5.6	1.2	1.44
$\Sigma x = 44$		$\Sigma (x_i - \bar{x})^2 = 3.12$

$$t = \frac{4.4 - 4}{.589 / \sqrt{10}}$$

$$= \frac{.4 \times \sqrt{10}}{.589}$$

$$v = n - 1 = 10 - 1 = 9$$

$$v = 9$$

$$t_{cal} = 2.213$$

level of significance = 5%

$$v = 9, t_{0.05} = 2.262$$

$$|t_{cal}| < t_{tab} \Rightarrow H_0 \text{ accepted}$$

i.e. Avg lifetime could be 4000 hrs.

Ex. A sample of 18 items has mean 24 units and S.D. 3 units. Test the hypothesis that it is random sample from Normal population with mean 27 units.

Sol
 $H_0: \mu = 27$, There is no significant difference b/w population mean and sample mean

$$\left\{ \begin{array}{l} n = 18 \checkmark \\ \bar{x} = 24 \\ s = 3 \end{array} \right.$$

$H_1: \mu \neq 27$ (Two tailed Test)

$$t = \frac{\bar{x} - \mu}{s / \sqrt{n-1}} = \frac{24 - 27}{3 / \sqrt{17}} = \frac{-3 \sqrt{17}}{3}$$

$$t_{cal} = -\sqrt{17} = -4.123$$

$$\checkmark |t_{cal}| = 4.123$$

level of sig $\rightarrow 5\%$ 0.05

$$v = d.f = n-1 = 18-1 = 17$$

$$d.f = 17, \quad t_{tab} = t_{0.05} = 2.110 \quad \checkmark$$

Conclusion $\underline{|t_{cal}| > t_{tab}} \Rightarrow \underline{H_0 \text{ rejected}}$

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Unit-5 (Lec-3)

(Maths-4)

Topic: Student t-test of Significance (Part-2)

t-test for significance between two sample Mean

Size

Sample Mean, Pop. Mean, Pop Var.

✓ Sample (1) ✓ n_1 , $x_1, x_2, x_3 \dots x_{n_1}$, $\bar{x}$, μ_1 , σ ✓ Sample (2) ✓ n_2 , $y_1, y_2, y_3 \dots y_{n_2}$, $\bar{y}$, μ_2 , σ

$$t = \frac{\bar{x} - \bar{y}}{S \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

If s_1, s_2 S.D. are given for samples then

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

OR

$$S^2 = \frac{\sum (x_i - \bar{x})^2 + \sum (y_i - \bar{y})^2}{n_1 + n_2 - 2}$$

✓ Degree of freedom

$$\rightarrow v_1 = n_1 - 1$$

 $\Rightarrow$ for sample (1)

$$\rightarrow v_2 = n_2 - 1$$

 $\Rightarrow$ for " (2)

$$d.f. = v = n_1 + n_2 - 2$$

level of significance = (5%, 1%)

Note:-

$$|t_{cal}| < t_{tab} \Rightarrow H_0 \text{ Accepted}$$

$$|t_{cal}| > t_{tab} \Rightarrow H_0 \text{ rejected}$$

$$\left. \begin{array}{l} H_0 \\ H_1 \end{array} \right\}$$

$$|t_{cal}| > t_{tab}$$

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90		1.662	1.987	2.368	2.632	3.183	3.402
100		1.660	1.984	2.364	2.626	3.174	3.390
120		1.658	1.980	2.358	2.617	3.160	3.373

Ex. Samples of sizes 10 and 14 were taken from two Normal populations with S.D 3.5 and 5.2. The sample means were found to be 20.3 and 18.6. Test whether the means of two populations are same at 5% level.

	Size	S.D.	Sample Mean
Sample (1)	10	3.5 = s_1	20.3 = $\bar{x}$
	11		
	14		
			18.6 = $\bar{y}$

Sample (2) | $\begin{matrix} 14 \\ n_2 \end{matrix}$ $S_2 = 5.2$ 10.6 σ

H_0 : There is no significant difference, $\mu_1 = \mu_2$

H_1 : $\mu_1 \neq \mu_2$ (Two tailed)

$$t = \frac{\bar{x} - \bar{y}}{S \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

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

$$= \frac{10(3.5)^2 + 14(5.2)^2}{10 + 14 - 2}$$

$$S^2 = \frac{501.06}{22} = 22.775$$

$$S = 4.772$$

$$t = \frac{20.3 - 18.6}{4.772 \sqrt{\frac{1}{10} + \frac{1}{14}}}$$

$$= 0.8604$$

$$t_{cal} = 0.8604$$

$$d.f = n_1 + n_2 - 2 = 10 + 14 - 2 = 22$$

t at 5% level of significance for 22 d.f

$$t_{tab} = t_{0.05} = 2.074$$

Conclusion:

$$|t_{cal}| = 0.8604 < t_{0.05}$$

$\Rightarrow$ Null hypothesis H_0 is accepted

$\Rightarrow$ No significant difference b/w their means

Ex. ✓ The height of 6 randomly chosen sailors in inches are 63, 65, 68, 69, 71 and 72. Those of 9

randomly chosen soldiers are 61, 62, 65, 66, 69, 70, 71, 72 and 73. Test whether the sailors are on average taller than soldiers.

$X_1 \rightarrow$ Sailors $n_1 = 6$

$X_2 \rightarrow$ Soldiers $n_2 = 9$

H_0 : There is no significant difference i.e. $\mu_1 = \mu_2$

H_1 : $\mu_1 > \mu_2$ (One tail test)

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

$$t = \frac{\bar{x}_1 - \bar{x}_2}{S \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

Sample 1 Sailors

$\checkmark X_1$:	<u>63</u>	65	68	69	71	72	$\Sigma x_1 = 408$
$x_1 - \bar{x}_1$:	-5	-3	0	1	3	4	
$(x_1 - \bar{x}_1)^2$:	25	9	0	1	9	16	$\rightarrow \Sigma (x_1 - \bar{x}_1)^2 = 60$

mean $\bar{x}_1 = \frac{\Sigma x_1}{n_1} = \frac{408}{6} = 68$, $\Sigma (x_1 - \bar{x}_1)^2 = 60$ ✓✓

Sample 2 Soldiers

$\checkmark X_2$:	<u>61</u>	62	65	66	69	70	71	72	73	$\Sigma x_2 = 609$
$x_2 - \bar{x}_2$:	-6.66	-5.66	-2.66	-1.66	1.34	2.34	3.34	4.34	5.34	
$(x_2 - \bar{x}_2)^2$:	44.36	32.035	7.0756	2.7556	1.7956	5.4756	11.1556	18.8356	28.5156	

$$\bar{X}_2 = \frac{\sum X_2}{n_2} = \frac{609}{9} = \underline{\underline{67.66}} \quad \sum (X_2 - \bar{X}_2)^2 = 152.0002$$

$$S^2 = \frac{\sum (X_1 - \bar{X}_1)^2 + \sum (X_2 - \bar{X}_2)^2}{n_1 + n_2 - 2} = \frac{60 + 152.0002}{6 + 9 - 2} = 16.3077$$

$$S = 4.038$$

$$t = \frac{\bar{X}_1 - \bar{X}_2}{S \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} = \frac{68 - 67.66}{4.038 \sqrt{\frac{1}{6} + \frac{1}{9}}} = 0.1569$$

$$t_{cal} = 0.1569$$

$$d.f. = n_1 + n_2 - 2 = 6 + 9 - 2 = \underline{\underline{13}} \leftarrow$$

$$\text{level of signif} = 5\% \leftarrow$$

$$t_{0.05} = \underline{\underline{1.771}}$$

$$\text{Conclusion : } |t_{cal}| = 0.1569 < t_{0.05}$$



$\Rightarrow H_0$ hypothesis accepted.

$\Rightarrow$ No significant difference b/w their average

Sailors are not average taller than
Soldiers

Practice Question

Q → The mean life of 10 electric motors was found

to be 1450 hrs with S.D of 423 hrs. A second sample of 17 motors chosen from different batch showed mean life 1280 hrs with S.D of 398 hrs. Is there a significant difference b/w means of two samples.

==

Topic: Chi-Square Test (χ^2 -Test) [Most Imp. Topic]

χ^2 -test

$$\chi^2 = \sum \left[\frac{(O_i - E_i)^2}{E_i} \right]$$

O_i = Observed frequency $\checkmark \leftarrow (O_i - E_i)^2 / E_i$

E_i = Expected frequency $\leftarrow$

$\checkmark$ for simple data

$$E_i = \frac{\sum O_i}{n}$$

$n \rightarrow$ No of Observed

Degree of freedom

$$d.f = \nu = n - 1$$

n = No of observations.

$$\Rightarrow \nu = \frac{(p-1)(q-1)}{(R-1)(C-1)}$$

for tabular data

$\checkmark P \rightarrow$ No of Rows —

$q \rightarrow$ No. of columns

Level of Significance (α) $\rightarrow$ 5% or 1%

Note:

$$\checkmark \chi^2_{cal} < \chi^2_{tab}$$

$\Rightarrow$ Null hypothesis H_0 accepted

$$\chi^2_{cal} > \chi^2_{tab}$$

$\Rightarrow$ Null hypothesis H_0 Rejected

$$\checkmark d.f = \nu = n - k$$

k = No. of independent Constraints.

Ex. :- The following table gives the number of accidents that took place in an industry during various days of week. Test if accidents are uniformly

distributed over the week.

Day:-	MON	TUE	WED	THUR	FRI	SAT	
No of Accidents:-	14	18	12	11	15	14	$\leftarrow O_i$

Sol:- H_0 : Accidents are Uniformly distributed

$$n = 6$$

$$v = n - 1 = 6 - 1 = 5$$

level of significance $(\alpha) = 5\%$

Day	Observed (O_i)	Expected (E_i) = $\sum O_i / n$	$(O_i - E_i)^2 / E_i$
MON	14 ✓	$84/6 = 14$ ✓	0
TUE	18 ✓	$84/6 = 14$ ✓	$16/14 = 1.1428$
WED	12	14	$4/14 = .2857$
THUR	11	14	$9/14 = .6428$
FRI	15	14	$1/14 = .0714$
SAT	14	14	0
$\sum O_i = 84$ ✓			$\sum \left[\frac{(O_i - E_i)^2}{E_i} \right] = 2.1427$ ✓

$$\chi^2_{\text{cal}} = 2.1427$$

$$E_i = \frac{\sum O_i}{n}, \quad \chi^2 = \sum \left[\frac{(O_i - E_i)^2}{E_i} \right]$$

level of sig $(\alpha) = 5\%$, $v = 5$

$$\chi^2_{(5\%, 5)} = 11.07 \quad (\text{Table})$$

Conclusion

$$\chi^2_{\text{cal}} = 2.1427 < \chi^2_{\text{tab}}$$

H_0 (Null hypothesis Accepted) $\Rightarrow$ Accidents are

Uniformly distributed.

Ex. A die is thrown 276 times and results of these throws are given below

No. appeared on die	1	2	3	4	5	6
freq	40	32	29	59	57	59

$59 \leftarrow O_i$

Test whether the die is biased OR Not

Aktu-2019

Solⁿ H_0 : Die is Unbiased.

$$n = 6$$

$$d.f = v = n - 1 = 6 - 1 = 5$$

$$\text{level of sig } (\alpha) = 5\%$$

No appeared	Observed O_i	Expected (E_i) $= \sum O_i / n$	$(O_i - E_i)^2 / E_i$
1	40	276/6 = 46	36/46
2	32	46	$(14)^2 / 46 = 196/46$
3	29	46	$(17)^2 / 46 = 289/46$
4	59	46	$(13)^2 / 46 = 169/46$
5	57	46	$(11)^2 / 46 = 121/46$
6	59	46	$(13)^2 / 46 = 169/46$

$$\sum O_i = 276$$

$$\sum \left[\frac{(O_i - E_i)^2}{E_i} \right] = \frac{980}{46}$$

$$\chi^2 = \sum \left[\frac{(O_i - E_i)^2}{E_i} \right] = \frac{980}{46} = 21.30$$

$$\chi^2$$

$$\chi^2_{\text{cal}} = 21.30$$

$$v = n - 1 = 5$$

Level of sig (α) = 5%

$$\chi^2_{(5\%, 5)} = 11.07 = \chi^2_{\text{tab}}$$

Conclusion

$$\chi^2_{\text{cal}} = 21.30 > \chi^2_{\text{tab}}$$

$\Rightarrow$ Null hypo. H_0 rejected

$\Rightarrow$ Die is biased

TABLE A.4

χ^2 (Chi-Squared) Distribution: Critical Values of χ^2

ν	Degrees of freedom	Significance level		
		5%	1%	0.1%
1		3.841	6.635	10.828
2		5.991	9.210	13.816
3		7.815	11.345	16.266
4		9.488	13.277	18.467
5		11.070	15.086	20.515
6		12.592	16.812	22.458
7		14.067	18.475	24.322
8		15.507	20.090	26.124
9		16.919	21.666	27.877
10		18.307	23.209	29.588

Practice Question

Q → Demand for a particular part in a factory was found to vary from day to day
In a sample study,

Day	MON	TUE	WED	THUR	FRI	SAT
No. of Parts demand	1124	1125	1110	1120	1126	1115

Test the hypothesis that the No. of parts demanded does not depend on the day of week.

(Given $\chi^2_{(5\%, 5)} = 11.07$, $\chi^2_{(5\%, 6)} = 12.59$)



$$\chi^2_{(5\%, \underset{\uparrow}{7})} = \underline{14.07} \quad)$$

Unit-5 (Lec-5) ✓
(Maths-4)

Topic: Chi-Square Test (Part-2) Most Imp Topic

$$\chi^2 = \sum \left[\frac{(O_i - E_i)^2}{E_i} \right]$$

$$E_i = \frac{\sum O_i}{n}$$

O_i = Observed frequency ✓ → (Given)

E_i = Expected frequency →

✓ $v = d.f = n - 1$ simple data

$n \rightarrow$ No of observation

✓ In case of Binomial distribution $d.f = n - 1$ ←

" " Poisson " " = $n - 2$ ←

" " Normal " " = $n - 3$ ←

level of significance (α) — (5% OR 1%) ✓

If ✓ $(\chi^2_{cal}) > \chi^2_{tab}$ ⇒ H_0 Rejected

χ^2_{cal} < χ^2_{tab} ⇒ H_0 accepted

TABLE A.4

χ^2 (Chi-Squared) Distribution: Critical Values of χ^2

Degrees of freedom	Significance level		
	5%	1%	0.1%
1	3.841	6.635	10.828
2	5.991	9.210	13.816
3	7.815	11.345	16.266

4	9.488	13.277	18.467
→ 5	<u>11.070</u>	15.086	20.515
6	12.592	16.812	22.458
7	14.067	18.475	24.322
8	15.507	20.090	26.124
9	16.919	21.666	27.877
10	18.307	23.209	29.588

Ex. : → A survey of 320 families with 5 children shows the following distribution :

No of children	5B 0G	4B 1G	3B 2G	2B 3G	1B 4G	0B 5G	Total
No of families	<u>18</u> 0 ₁	<u>56</u> 0 ₂	<u>110</u> 0 ₃	<u>88</u> 0 ₄	<u>40</u> 0 ₅	<u>8</u> 0 ₆	<u>320</u>

Test hypothesis that data are Binomially distributed and Male and female birth have equal probability. (Given $\chi^2_{(5\%, 5)} = \underline{11.07}$)

Sol $p = \text{prob for B.}$ $q = \text{" " G.}$ $p = q = \frac{1}{2}$ $P(r) = {}^n C_r p^r q^{n-r}$
 $r = 0, 1, 2, 3, 4, 5$ $N = 320$
 $p \rightarrow \text{prob of success}$
 $q \rightarrow \text{" " failure}$

$$E_6 \quad P(r=0) = P(\underline{0B, 5G}) = N \cdot {}^n C_r p^r q^{n-r} = 320 \cdot {}^5 C_0 \left(\frac{1}{2}\right)^0 \left(\frac{1}{2}\right)^5 \\ = 320 \cdot 1 \cdot \frac{1}{2^5} = 10 \text{ family}$$

$$E_5 \quad P(r=1) = P(\underline{1B, 4G}) = 320 \cdot {}^5 C_1 \left(\frac{1}{2}\right)^1 \left(\frac{1}{2}\right)^4 = 320 \cdot 5 \cdot \frac{1}{2^5} \\ = 50 \text{ families}$$

$$E_4 \quad P(r=2) = P(\underline{2B, 3G}) = 320 \cdot {}^5 C_2 \left(\frac{1}{2}\right)^2 \left(\frac{1}{2}\right)^3 = 320 \cdot \frac{5 \times 4}{2 \times 1} \cdot \frac{1}{2^5}$$

$$E_3 \quad P(r=3) = P(\underline{3B, 2G}) = 320 \cdot {}^5C_3 \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^2 = 320 \cdot \frac{5 \times 4 \times 3}{2 \times 1} \left(\frac{1}{2}\right)^5 = 100 \text{ families} \checkmark$$

$$E_2 \quad P(r=4) = P(\underline{4B, 1G}) = 320 \cdot {}^5C_4 \left(\frac{1}{2}\right)^4 \left(\frac{1}{2}\right)^1 = 320 \cdot 5 \cdot \frac{1}{2^5} = 50 \checkmark$$

$$E_1 \quad P(r=5) = P(\underline{5B, 0G}) = 320 \cdot {}^5C_5 \left(\frac{1}{2}\right)^5 \left(\frac{1}{2}\right)^0 = 320 \cdot 1 \cdot \frac{1}{2^5} = 10 \checkmark$$

H_0 : Data are Binomially distributed

No of children	Observed freq O_i	Expected freq E_i	$(O_i - E_i)^2 / E_i$
5B, 0G	18 ✓	10 ✓	64/10 = 6.4
4B, 1G ✓	56 ✓	50 ✓	36/50 = .72
3B, 2G ✓	110 ✓	100 ✓	100/100 = 1
2B, 3G ✓	88 ✓	100 ✓	144/100 = 1.44
1B, 4G	40 ✓	50 ✓	100/50 = 2
0B, 5G	8 ✓	10 ✓	4/10 = .4
	$\Sigma O_i = 320$	$\Sigma E_i = 320$	11.96

$$\chi^2_{\text{cal}} = 11.96$$

$$v = 6 - 1 = 5$$

At 5% level of sig, $\chi^2_{\text{tab}} = 11.07$

$$\chi^2_{\text{cal}} = 11.96 > \chi^2_{\text{tab}}$$

$$\chi^2_{cal} = 10.10 > \chi^2_{table}$$

$\Rightarrow H_0 : \underline{\underline{\text{Rejected}}}$

Topic: Chi-Square Test (Part - 3) ✓

χ^2 -Test As a test of Independence ✓

- * Sample data is given in tabular form, called Contingency table
- * By Using χ^2 -test, we can find whether or not two attributes are related. ✓✓

$$* \text{ d.f} = v = (p-1)(q-1) \quad \text{OR} \quad (R-1)(C-1)$$

p = No of Rows ✓

q = No of col. ✓

$$\chi^2 = \sum \left[\frac{(O_i - E_i)^2}{E_i} \right]$$

Contingency Table

a	b
c	d

Observed frequency

→	$a = O_1$	$b = O_2$	$a+b$	← Sum
→	$c = O_3$	$d = O_4$	$c+d$	←
→	$a+c$	$b+d$	$N = a+b+c+d$	← Total
	↑ Sum			

Expected frequencies

$E_1 = \frac{(a+b)(a+c)}{a+b+c+d}$	$E_2 = \frac{(a+b)(b+d)}{a+b+c+d}$
$E_3 = \frac{(c+d)(a+c)}{a+b+c+d}$	$E_4 = \frac{(c+d)(b+d)}{a+b+c+d}$

$$O_1 \rightarrow E_1$$

$$O_2 \rightarrow E_2$$

$$O_3 \rightarrow E_3 \quad O_4 \rightarrow E_4$$

↑ Contingency table

TABLE A.4

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9	16.919	21.666	27.877
10	18.307	23.209	29.588

Ex. From the following table regarding the color of eyes of father and son, test if the colour of son's eye is associated with that of the father

[2020-21]

Eye colour of father	Eye Colour of SON	
	Light	Not Light
	Light	Not Light
Light	471	51
Not Light	140	230

Sol

H_0 : The colour of son's eye is not associated with father's eye colour i.e. they are independent.

	SON		
	Light	Not Light	
Father	O_1 471	O_2 51	522
	O_3 148	O_4 230	378
	619	281	900



(2x2)

Expected frequencies

Eye color of SON Eye color of FATHER	Light	Not Light	Total
Light	$E_1 = \frac{522 \times 619}{900}$ $= 359.02$	$E_2 = \frac{522 \times 281}{900}$ $= 162.98$	$\Rightarrow 522$
Not Light	$E_3 = \frac{378 \times 619}{900}$ $= 259.98$	$E_4 = \frac{378 \times 281}{900}$ $= 118.02$	$\Rightarrow 378$
	619	281	900

Calculation of χ^2

O_i	471 ✓	51	148	230
E_i	<u>359.02 ✓</u>	162.98	259.98	118.02
$\frac{(O_i - E_i)^2}{E_i}$	$\frac{(471 - 359.02)^2}{359.02}$ $= 34.927$	76.939	48.232	106.249

$\Sigma = 266.347$

$$\chi^2_{cal} = \boxed{266.347}$$

$$v = (R-1)(C-1) = (2-1)(2-1) = 1$$

$$v = 1$$

level of sig (α) = 5%

$$\chi^2_{(.05, 1)} = 3.841$$

$$\chi^2_{cal} = 266.347 > \chi^2_{tab}$$

$\Rightarrow H_0$ Rejected

$\Rightarrow$ Color of son's eye associated with father's eye colour.

Ex. gdp

To test effectiveness of inoculation against cholera the following table was obtained

	Attacked ✓✓	Not Attacked ✓✓	Total
Inoculated ✓	30 ✓	160 ✓	190 ←
Not Inoculated ✓	140 ✓	460 ✓	600 ←
	170	620	790

Use χ^2 -test to defend or refute the statement
that the inoculation prevents attack from cholera.
[2009, 2018]

Sol

H_0 : The inoculation does not prevent attack.

Under H_0

observed 2x2

	O_1	O_2	Total
O_1	30	160	190 ✓
O_3	140	460	600 ✓

Expected Frequencies

$E_1 = \frac{190 \times 170}{790}$ $= 40.886$	$E_2 = \frac{190 \times 620}{790}$ $= 149.11$
$E_3 = \frac{600 \times 170}{790}$	$E_4 = \frac{600 \times 620}{790}$

170	460	600	790 = 129.11	790 = 470.89
<u>170</u>	<u>620</u>	N = <u>790</u>		

Calculation of χ^2 ✓

O_i	O_1	O_2	O_3	O_4	
	30 ✓	160	140	460	
E_i	40.886 ✓	149.11	129.11	470.89	
$\frac{(O_i - E_i)^2}{E_i}$	$\frac{(30 - 40.886)^2}{40.886}$ = 2.898	.795	.918	.252	Σ = 4.863

$$\chi^2_{cal} = 4.863 \quad \checkmark$$

$$v = (R-1)(C-1) = (2-1)(2-1) = 1$$

$$\alpha = 5\%$$

$$\chi^2_{(5\%, 1)} = 3.841$$

$$\chi^2_{cal} > \chi^2_{tab}$$

$\Rightarrow H_0$ Rejected

$\Rightarrow$ we defend that inoculation prevents attack.

Topic: F-Test for significance of variance

Let
 Sample (1) $x_1, x_2, x_3 \dots x_{n_1}$
 Sample (2) $y_1, y_2, y_3 \dots y_{n_2}$ } be two random samples
 drawn from sample
 population or population with equal variance.

F-test uses variance ratio to test the
 significance of difference between variances of
 two samples

$$F = \frac{S_1^2}{S_2^2} \quad (S_1 > S_2)$$

	Size	Variance	
Sample (1)	n_1	S_1^2	$S_1 = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n_1 - 1}}$
Sample (2)	n_2	S_2^2	$S_2 = \sqrt{\frac{\sum (y_i - \bar{y})^2}{n_2 - 1}}$

$$\bar{x} = \sum x_i / n_1$$

$$\bar{y} = \sum y_i / n_2$$

Degree of freedom d.f.

(n_1)

for sample 1

(n_2)

" sample 2

$$\begin{aligned} \checkmark \quad v_1 &= n_1 - 1 \leftarrow \\ v_2 &= n_2 - 1 \leftarrow \end{aligned}$$

level of significance $\rightarrow$ (5% or 1%)

If $F_{cal} < F_{tab} \Rightarrow$ Null hypothesis is Accepted

$F_{cal} > F_{tab} \Rightarrow$ Null hypothesis is Rejected

F-Table

The F-distribution for various degrees of freedom v_1 and v_2 is given in the following table :

Table 12.3 : Values of F (F-distribution) for 5% and 1% level, where v_1 is the number of degree of freedom for greater estimate of variance and v_2 for the smaller.

$v_2 \backslash v_1$	1	2	3	4	5	6	8	12	24	∞
1	161.4	199.5	215.7	224.6	230.2	234.0	238.9	243.9	249.0	254.0
2	18.51	19.00	19.16	19.25	19.30	19.32	19.37	19.41	19.45	19.50
3	10.13	9.55	9.28	9.12	9.01	8.94	8.84	8.74	8.64	8.53
4	7.71	6.94	6.59	6.39	6.26	6.16	6.04	5.91	5.77	5.63
5	6.61	5.79	5.41	5.19	5.05	4.95	4.82	4.68	4.53	4.36
6	5.99	5.14	4.76	4.53	4.39	4.28	4.15	4.00	3.84	3.67
7	5.59	4.74	4.35	4.12	3.97	3.87	3.73	3.57	3.41	3.23
8	5.32	4.46	4.07	3.84	3.69	3.58	3.44	3.28	3.12	2.93
9	5.12	4.26	3.86	3.63	3.48	3.37	3.23	3.07	2.90	2.71
10	4.96	4.10	3.71	3.48	3.33	3.22	3.07	2.91	2.74	2.54
12	4.75	3.88	3.49	3.26	3.11	3.00	2.85	2.69	2.50	2.30
14	4.60	3.74	3.34	3.11	2.96	2.85	2.70	2.53	2.35	2.13
16	4.49	3.63	3.24	3.01	2.85	2.74	2.59	2.42	2.24	2.01
18	4.41	3.55	3.16	2.93	2.77	2.66	2.51	2.34	2.15	1.92
20	4.35	3.49	3.10	2.87	2.71	2.60	2.45	2.28	2.08	1.84

Ex: $\rightarrow$ Two Independent sample of sizes 7 and 6 had the following values :

✓ Sample A 28 30 32 33 31 29 34 $\leftarrow$
 ✓ Sample B 29 30 30 24 27 28 $\leftarrow$

Examine whether the samples have been drawn from Normal population having the same variance

Sol $H_0 : \sigma_1^2 = \sigma_2^2$, samples have been drawn from Normal population with same variance.

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

$$S_1^2 = \frac{\sum (x_i - \bar{x})^2}{n_1 - 1}$$

$$S_2^2 = \frac{\sum (x_i - \bar{x}_2)^2}{n_2 - 1}$$

$$F = \frac{S_1^2}{S_2^2} \quad (S_1 > S_2)$$

Sample A			Sample B		
X_1	$X_1 - \bar{X}_1$ $X_1 - 31$	$(X_1 - \bar{X})^2$	X_2	$(X_2 - \bar{X}_2)$ $X_2 - 28$	$(X_2 - \bar{X}_2)^2$
28	-3	9	↓ 29	1	1
30	-1	1	30	2	4
32	1	1	30	2	4
33	2	4	24	-4	16
31	0	0	27	-1	1
29	-2	4	28	0	0
34	3	9			
		<u>28</u>			<u>26</u>
<u>217</u>			168		

$$\bar{X}_1 = \frac{\sum X_1}{n_1} = \frac{217}{7} = 31$$

$$\bar{X}_2 = \frac{\sum X_2}{n_2} = \frac{168}{6} = 28$$

$$S_1^2 = \frac{\sum (X_1 - \bar{X}_1)^2}{n_1 - 1}$$

$$= \frac{28}{6} = 4.666$$

$$S_2^2 = \frac{\sum (X_2 - \bar{X}_2)^2}{n_2 - 1}$$

$$= \frac{26}{5} = 5.2$$

$$F = \frac{S_2^2}{S_1^2} = \frac{5.2}{4.666}$$

$$f_{cal} = 1.1158$$

$$v_2 = n_1 - 1 = 7 - 1 = 6$$

$$v_1 = n_2 - 1 = 6 - 1 = 5$$

$$F_{tab} = 4.39$$

level of sig = 5%

$$F_{cal} = 1.1158 < F_{tab} \Rightarrow H_0 \text{ accepted}$$

there is no significant diff b/w variances
=

Ex. Two Random samples drawn from 2 Normal Populations are as follows

A	17	27	18	25	27	29	13	17
B	16	16	20	27	26	25	21	

Test whether the samples are drawn from same Normal Population.

[Hint $\rightarrow$ To test samples are drawn from same population we test

(i) equality of Means by t-test

(ii) equality of Population Variance by F-test

First F-Test

H_0 : There is no significant difference b/w

Population variance =

$$\text{i.e. } \sigma_1^2 = \sigma_2^2$$

$$S_1^2 = \frac{\sum (x - \bar{x})^2}{n_1 - 1}$$

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

$$F = \frac{\sqrt{S_1^2}}{\sqrt{S_2^2}} = \underline{1.7319}$$

($S_1 > S_2$)

$$\bar{X} = 21.625, S_1^2 = 36.267$$

$$\bar{Y} = 21.57, S_2^2 = 20.94$$

Ans:- H_0 accepted

T-test $\checkmark H_0: \mu_1 = \mu_2$ Population means are equal
 $H_1: \mu_1 \neq \mu_2$

$$t = \frac{\bar{x} - \bar{y}}{S \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

$$S^2 = \frac{\sum (x - \bar{x})^2 + \sum (y - \bar{y})^2}{n_1 + n_2 - 2}$$

$$S^2 = 29.196 \leftarrow$$

$$S = 5.403$$

$$t = 0.2449$$

$$d.f = n_1 + n_2 - 2$$

$$t_{tab}(5\%, d.f) = \underline{\hspace{2cm}}$$

Ans:- H_0 Accepted

Final Ans: $\rightarrow$ Two samples have been drawn from same Normal Population.

Topic : ANOVA (Analysis of Variance)

ANOVA is a statistical tool that can be used for comparison among more than two groups. It is used for testing the hypothesis of equality of more than two normal population means.

'Analysis of Variance' is carried on the basis of Ratio between the variances.

The Variance Ratio is obtained by dividing the variance between samples by variance within samples.

$$F = \frac{\text{Variance between the sample}}{\text{Variance within the samples}}$$

This ratio forms F-statistic

Remember that in F-statistic, variance b/w the samples greater than variance within samples

Techniques of ANOVA

* One way ANOVA → used to Analysis of effects of one independent factor on dependent variable.

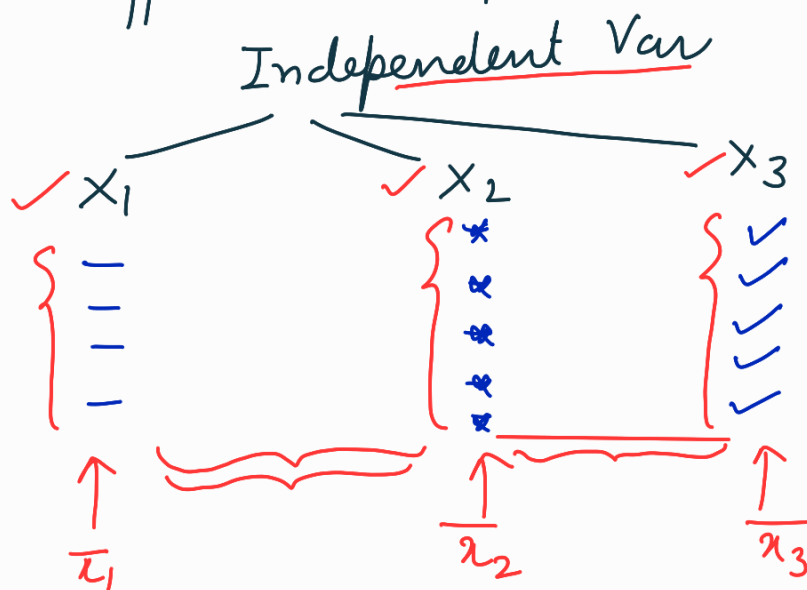
* Two way ANOVA → used to Analysis of effects of two independent factors on dependent variable

Assumptions: →

- * Each sample is randomly drawn. ✓
- * Populations, from which the samples are drawn, are Normally distributed. ✓
- * Each sample is independent of other sample. ✓
- * Each of population has same variances and identical means.

ONE WAY ANOVA One independent factor

effects different samples



Total Variation

Variance b/w samples Variance within samples
(error variance)

✓ Null hypothesis $H_0: \bar{\mu}_1 = \bar{\mu}_2 = \bar{\mu}_3$, i.e. means of populations from which samples are drawn are equal

Alternate hypothesis $\Rightarrow H_1: \underline{\text{at least one mean is different}}$

✓ Short cut Method :-

$\checkmark X_1$ $\checkmark X_1^2$	$\checkmark X_2$ X_2^2	$\checkmark X_3$ X_3^2
$n_1=4$ $\left\{ \begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array} \right.$	$n_2=4$ $\left\{ \begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array} \right.$	$n_3=4$ $\left\{ \begin{array}{c} \vdots \\ \vdots \\ \vdots \end{array} \right.$
ΣX_1 ΣX_1^2	ΣX_2 ΣX_2^2	ΣX_3 ΣX_3^2

* Grand Total G.T. = $\Sigma X_1 + \Sigma X_2 + \Sigma X_3$ ←

* Correction factor C.F. = $\frac{(G.T.)^2}{n}$ $n \rightarrow$ total No. of items.

* Sum of squares between samples (SSC) ←

$$\frac{(\Sigma X_1)^2}{n_1} + \frac{(\Sigma X_2)^2}{n_2} + \frac{(\Sigma X_3)^2}{n_3} - C.F.$$

d.f = $c-1$ d.f = $3-1=2$

* Total Sum of squares (SST)

$$\Sigma X_1^2 + \Sigma X_2^2 + \Sigma X_3^2 - C.F.$$

d.f = $n-1$ =

* Sum of squares within the samples (SSE)^{error}

$$\Rightarrow SST = SSC + SSE$$

$$SSE = SST - SSC$$

d.f = $n-c$

* Mean Sum of squares between samples (Variance)

$$MSC = \frac{SSC}{d.f} = \frac{SSC}{c-1}$$

* Mean Sum of squares within samples (Variance)

$$MSE = \frac{SSE}{n-c} = \frac{SSE}{d.f}$$

* F-Statistic $F = \frac{MSC}{MSE}$ $F_{cal} \cdot d.f. F_{(n,d)}$

* level of significance $(\alpha) = 5\%$

* If $\underline{F_{cal}} < \underline{F_{tab}}$ at 5% level of significance $\Rightarrow$ H_0 Accepted ✓

$F_{cal} > F_{tab} \Rightarrow H_0$ Rejected

✓✓ ANOVA TABLE

Source of Variation	Sum of Squares	Degree of freedom	Mean sum of squares	F
✓ Between Samples	<u>SSC</u> ✓	<u>C-1</u> ✓	<u>MSC</u> = $SSC / (C-1)$	<u>F</u> = $\frac{MSC}{MSE}$ <u>MSC > MSE</u>
✓ Within Samples	<u>SSE</u> ✓	<u>n-C</u> ✓	<u>MSE</u> = $SSE / (n-C)$	
Total	<u>SST</u> ✓	<u>n-1</u> ✓		

$$F = \frac{\text{Variance b/w samples}}{\text{Variance within samples}} = \frac{MSC}{MSE}$$

$F(n, D)$
 ↙ ↘
 D.f. of Nr d.f. of Dr

Ex. It is desired to compare three hospitals with regards to number of deaths per month. A sample of death records were selected from the records of each hospital and the number of deaths was as given below. From these data, suggest a difference in the no. of deaths per months among 3 hospitals

✓ A	B	C
3	6	7
4	3	3
3	3	4
5	4	6
0	4	5

✓ [2020-21]
AKTU

Given
 ✓ $F_{2,12} = 3.89$
 at 5%

Sol. Null hypothesis H_0 : There is no significant difference

Null hypothesis: No. of deaths among three hospitals ✓
 H_1 : There is significant difference in No. of deaths ✓

✓ A	✓ B	✓ C
x_i	x_2	x_3
3	6	7
4	3	3
3	3	4
5	4	6
0	4	5
$\Sigma x_1 = 15$	$\Sigma x_2 = 20$	$\Sigma x_3 = 25$
$\Sigma x_1^2 = 59$	$\Sigma x_2^2 = 86$	$\Sigma x_3^2 = 135$

$$G.T = \Sigma x_1 + \Sigma x_2 + \Sigma x_3 = 15 + 20 + 25 = 60 \quad \checkmark$$

C.F. = $\frac{(G.T)^2}{n} = \frac{(60)^2}{15} = \frac{3600}{15} = 240$

correction factor

Sum of squares b/w samples (SSC) ✓

$$\Rightarrow \frac{(\Sigma x_1)^2}{n_1} + \frac{(\Sigma x_2)^2}{n_2} + \frac{(\Sigma x_3)^2}{n_3} - C.F.$$

$$= \frac{(15)^2}{5} + \frac{(20)^2}{5} + \frac{(25)^2}{5} - 240 = 10 = SSC$$

Total sum of squares (SST) ✓

$$= \Sigma x_1^2 + \Sigma x_2^2 + \Sigma x_3^2 - C.F.$$

$$= 59 + 86 + 135 - 240 = 40 = SST$$

ANOVA Table



↗ Variance

Source of Variation	Sum of Squares	Degree of freedom	Mean sum of squares	F
✓ Between Samples	$SSC = \underline{10}$	$C-1 = 3-1 = \underline{2}$	$MSC = SSC / (C-1) = 10/2 = \underline{5}$	$F = \frac{MSC}{MSE} = 5/2.5 = \underline{2}$
✓ Within Samples	$SSE = \underline{30}$	$n-C = 15-3 = \underline{12}$	$MSE = SSE / (n-C) = 30/12 = \underline{2.5}$	
Total	$SST = 40$	$n-1 = 15-1 = \underline{14}$		

$$F_{cal} = 2$$

$$F_{(2,12) cal} = 2$$

$$F_{(2,12) tab} = 3.89$$

$F_{cal} < F_{tab}$ Accepted H_0

There is no significant diff.

Ex. The following figures relate to the production in kg. of three varieties I, II, III of wheat sown in 12 plots.

✓ Variety I	14	16	18		
✓ Variety II	14	13	15	22	
✓ Variety III	18	16	19	19	20

Is there any significant difference in production of three Varieties ? Given $F_{2,9} = 4.26$ at 5% level significance.

Difference in population means

Sol H_0 : No significant difference
 $\mu_1 = \mu_2 = \mu_3$

H_1 : There is significant difference ✓

Use Code data by subtracting 12 from each figures

X_1 ^I $\rightarrow X_1^2$	X_2 ^{II} $\rightarrow X_2^2$	X_3 ^{III} $\rightarrow X_3^2$
$\rightarrow \checkmark 2$ 4	$\checkmark 2$ 4	6 36
$\rightarrow \checkmark 4$ 16	$\checkmark 1$ 1	4 16
$\rightarrow \checkmark 6$ 36	$\checkmark 3$ 9	7 49
	$\checkmark 10$ 100	7 49
		8 64
$\Sigma X_1 = 12$	$\Sigma X_2 = 16$	$\Sigma X_3 = 32$
$\Sigma X_1^2 = 56$	$\Sigma X_2^2 = 114$	$\Sigma X_3^2 = 214$

$$G.T = \Sigma X_1 + \Sigma X_2 + \Sigma X_3 = 12 + 16 + 32 = 60 \checkmark$$

$$\checkmark C.F. = \frac{(G.T)^2}{n} = \frac{(60)^2}{12} = 300$$

Sum of squares between Samples (SSC)

$$= \frac{(\Sigma X_1)^2}{n_1} + \frac{(\Sigma X_2)^2}{n_2} + \frac{(\Sigma X_3)^2}{n_3} - C.F.$$

$$= \frac{(12)^2}{3} + \frac{(16)^2}{4} + \frac{(32)^2}{5} - 300 = 16.8 = SSC$$

Total sum of squares (SST)

$$= \Sigma X_1^2 + \Sigma X_2^2 + \Sigma X_3^2 - C.F.$$

$$= 56 + 114 + 214 - 300$$

$$SST = 84 \leftarrow$$

Sum of squares within Samples

$$SSE = SST - SSC$$

$$= 84 - 16.8 = 67.2$$

$$SSE = 67.2$$

↓ ANOVA Table

→ Variance

Source of Variation	Sum of Squares	Degree of freedom	Mean Sum of Squares	F
✓ Between Samples	$SSC = 16.8$	$C-1 = 2$	$MSC = 16.8/2 = 8.4$ ✓	$F = \frac{MSC}{MSE} = 8.4/7.46 = 1.125$
✓ Within Samples	$SSE = 84 - 16.8 = 67.2$	$n-C = 9$	$MSE = 67.2/9 = 7.46$ ✓	
Total	$SST = 84$	$n-1 = 11$		

$$F_{cal}(2,9) = 1.125$$

$$F_{tab}(2,9) = 4.86$$

$$F_{cal} < F_{tab} \Rightarrow H_0 \text{ Accept}$$

==

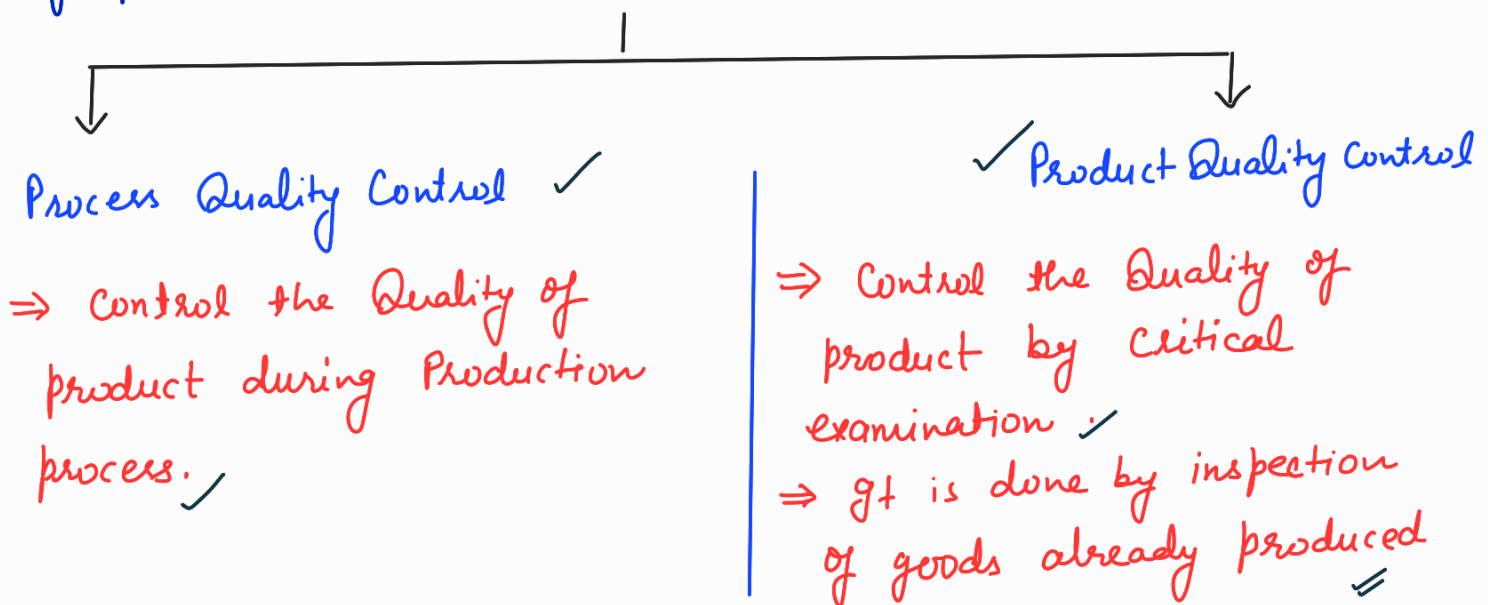
Topic: Statistical Quality Control (SQC) ✓

Statistical Quality Control is statistical method to ensure the quality of a product such as

- * standards of quality acceptable to customers and economical ✓
- * to locate ✓ and identify ✓ the process of faults to control defectives. ✓
- * to take corrective measures to maintain quality of products. ✓

Techniques of SQC

There are two broad ways of controlling the quality of product

**Control Charts**

The most common working statistical tools in Quality Control are Control charts and it is based upon the fact that variability does exist in all repetitive processes.

This variability is due to chance cause as well as assignable cause. A control chart eliminates

error entirely due to assignable causes.

In SQC, Control charts is used to study and control the repetitive process ✓

Control limits are defined within which variations are acceptable. ✓

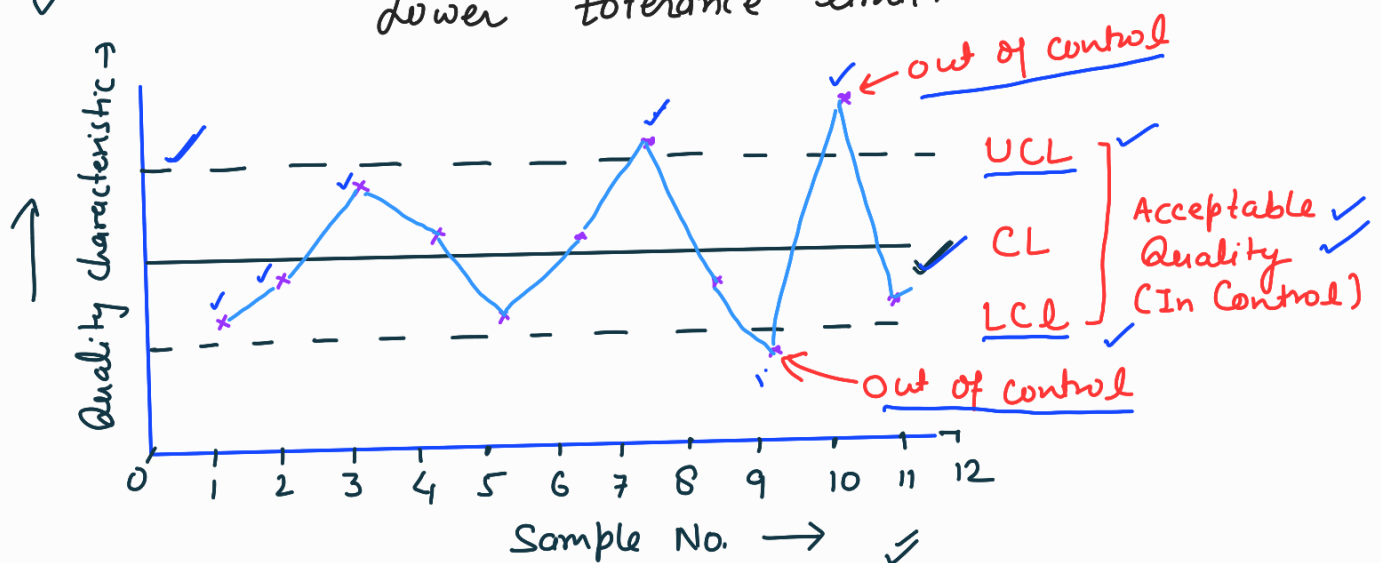
Any variations beyond these limits, can be expected to be caused by assignable cause which need to be controlled and investigated. ✓

A Control chart consists of 3 Horizontal lines ✓

* CL (control line) (solid line) represents desired control level of Process. ✓

* UCL (upper control limit) (Dotted line) represents upper tolerance limit. ✓

* LCL (lower control limit) (Dotted line) represents lower tolerance limit. ✓



Types of Control Charts

* Control charts for Variables → are useful to measure quality characteristics and control fully automatic process. like diameter of hole, thickness of pipe, length of bolt etc. ✓

They are used for measurable quality characteristics

→ Control chart for Sample Mean ($\bar{X}$ -chart) ✓

⇒ Control chart for Sample Range (R-chart) ✓

* ✓ Control charts for Attributes → Inspection by Attributes,
It is used to maintain and achieve an acceptable quality level. It is classified as "Good" or "Bad", "Ok" or "Not Ok", "Rejected" or "Accepted". ✓

⇒ Control charts for fraction defective (p-chart) ✓

⇒ Control chart for number of defective (np-chart) ✓

⇒ Control chart for number of defects (C-chart) ✓

Unit-5 (Lec-10)

(Maths-4)

Topic: **Statistical Quality Control (SQC)**

Control charts for Variables

A random sample of size n is drawn during manufacturing process

Sample mean $\bar{x} = \frac{x_1 + x_2 + x_3 \dots + x_n}{n}$

Sample Range $R = x_{\max} - x_{\min}$

Let k consecutive samples are selected

x_1	x_2	...	x_k
↓	↓		↓
$\bar{x}_1$	$\bar{x}_2$		$\bar{x}_k$
R_1	R_2		R_k

$$\bar{\bar{x}} = \frac{\check{x}_1 + \check{x}_2 + \check{x}_3 + \dots + \check{x}_k}{\check{k}} = \frac{\sum \bar{x}}{\text{No. of Samples}}$$

$$\bar{R} = \frac{R_1 + R_2 + R_3 + \dots + R_k}{k}$$

For X-chart (Mean chart)

$$\check{CL} = \begin{cases} \check{\bar{x}}, & \text{when tolerance limits are not given} \\ \check{\mu}, & \text{when tolerance limits are given} \end{cases}$$

$$\check{\mu} = \frac{1}{2} [LCL + UCL]$$

lower Control limit

Upper Control limit

Standard Error of Mean

$$\check{\sigma}_x = \frac{\bar{\sigma}}{\sqrt{n}}$$

The limits are
$$\left. \begin{aligned} UCL &= \bar{\bar{x}} + 3\sigma_x \\ LCL &= \bar{\bar{x}} - 3\sigma_x \end{aligned} \right\} \begin{array}{l} \text{Control limits are} \\ UCL = \bar{\bar{x}} + A_2 \bar{R} \\ LCL = \bar{\bar{x}} - A_2 \bar{R} \end{array}$$

A_2 are given in table ←

A_2 depends on sample size

For R-chart

$$CL = \bar{R}$$

$$LCL = \bar{R} D_3$$

$$UCL = \bar{R} D_4$$

D_3, D_4 can be found by table or given in Question. It is depend on sample

Size.

Process Capability is given by $\sigma = \frac{\bar{R}}{d_2}$, $\sigma = \text{S.D.}$

d_2 can be obtained by table and depend on sample size.

Ex: The following are mean length and ranges of lengths of a finished product from 10 samples each of size 5. The specification limits for length are $200 \pm 5 \text{ cm}$. Construct $\bar{X}$ and R-chart and examine whether the process is under control and state your recommendation

✓ Sample No	1	2	3	4	5	6	7	8	9	10
Mean ($\bar{X}$)	<u>201</u>	<u>198</u>	202	200	203	204	199	<u>196</u>	199	<u>201</u>
Range (R)	5	0	7	3	3	7	2	8	5	6

Assume for $n=5$ $A_2 = 0.58$, $D_4 = 2.11$, $D_3 = 0$

Solⁿ Control limits for $\bar{X}$ chart $\rightarrow$

Given limits 200 ± 5 (specification limits / Tolerance limits)

$$\bar{R} = \frac{R_1 + R_2 + R_3 + \dots + R_{10}}{10} = \frac{\sum R}{10}$$

$$= \frac{5 + 0 + 7 + 3 + 3 + 7 + 2 + 8 + 5 + 6}{10} = \frac{46}{10} = \boxed{4.6 = \bar{R}}$$

$$\boxed{CL = \mu = 200}$$

$$\begin{array}{|l} \text{Given} \\ A_2 = .58 \end{array}$$

$$UCL_x = \bar{\bar{x}} + A_2 \bar{R} = \underline{\mu} + A_2 \bar{R} = 200 + (.58 \times 4.6) = 202.668$$

$$LCL_x = \bar{\bar{x}} - A_2 \bar{R} = \underline{\mu} - A_2 \bar{R} = 200 - (.58 \times 4.6) = 197.332$$

$$D_3 = 0$$

$$D_4 = 2.11$$

$$n = 5$$

Control limits for R chart

$$\text{Control limit } CL = \bar{R} = 4.6$$

$$UCL_R = \bar{R} D_4 = 4.6 \times 2.11 = 9.706$$

$$LCL_R = \bar{R} D_3 = 4.6 \times 0 = 0$$

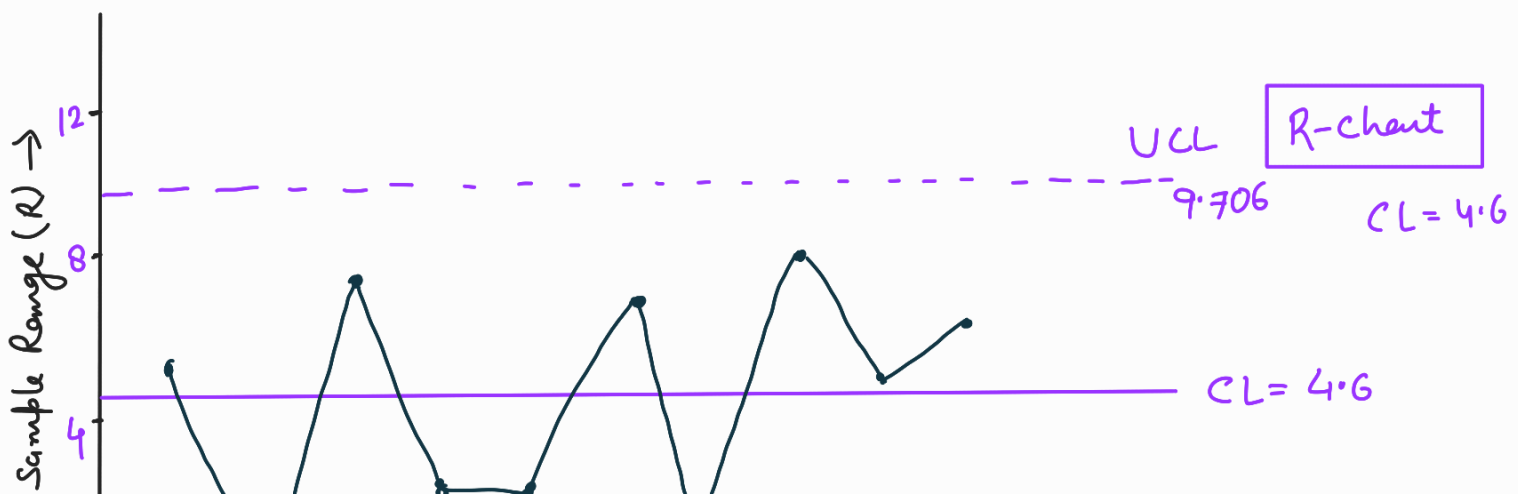
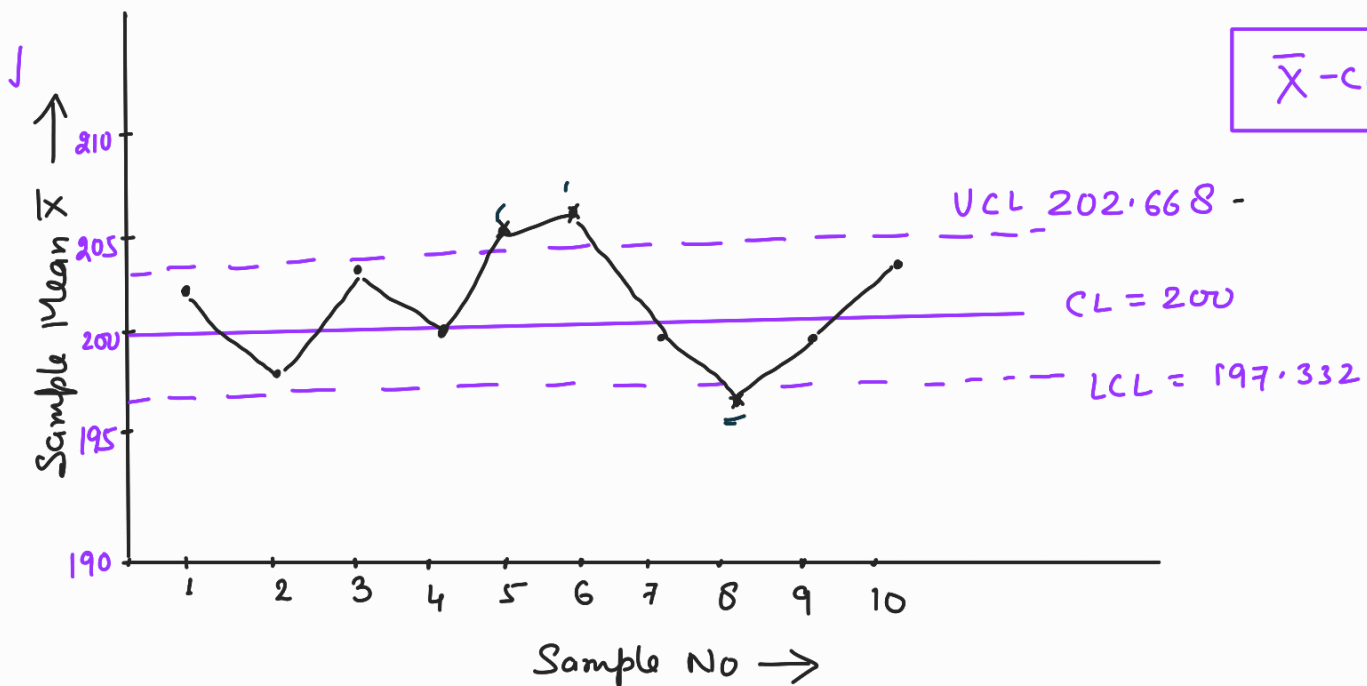
Given

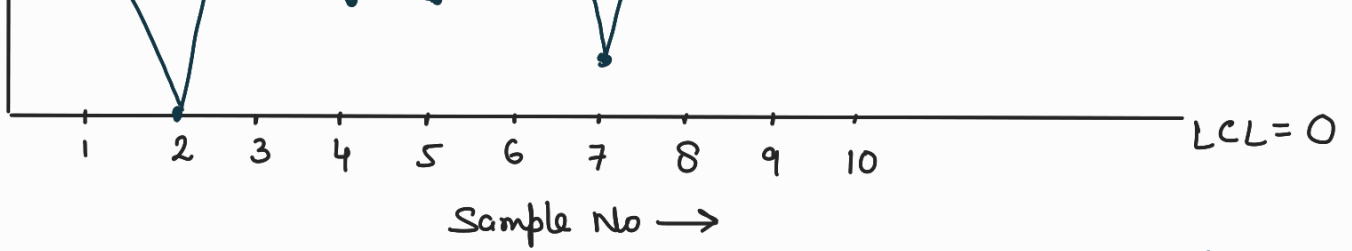
$$A_2 = .58$$

$$D_3 = 0$$

$$D_4 = 2.11$$

$$n = 5$$





Process Variability is under control. But $\bar{X}$ -chart, is not in statistical control. process should be halted.

Ex. A company manufactures screws to a nominal diameter $.500 \pm .030$ cm. Five samples were taken randomly from the manufactured lots and 3 measurements were taken on each sample at different lengths. Following are readings

Sample No ↓	Measurement per sample x (in cm)		
	1	2	3
- 1	.488	.489	.505
- 2	.494	.495	.499
- 3	.498	.515	.487
- 4	.492	.509	.514
- 5	.490	.508	.499

Calculate Control limits for $\bar{X}$, R charts. Draw $\bar{X}$, R Charts and examine whether the process is in Statistical Control? $[A_2 = 1.02, D_4 = \underline{2.57}, D_3 = 0 \text{ for } n = 3]$

<u>Sol</u>	x_1	x_2	x_3	x_4	x_5
Sample	1	2	3	4	5
	.488	.494 ✓	.498	.492	.490
	.489	.495	.515	.509	.508
	.505	.499	.487	.514	.499
Mean =	$\frac{1.482}{3}$	$\frac{1.489}{3}$	$\frac{1.5}{3}$	$\frac{1.515}{3}$	
	= .494	= .496	= .500	= .505	= .499
Range =	.505 - .488	.499 - .494	.515 - .487	.514 - .492	.508 - .490
	= .017	= .005	= .028	= .022	= .018

Range = Max - min

$$\bar{\bar{x}} = \frac{\bar{x}_1 + \bar{x}_2 + \bar{x}_3 + \bar{x}_4 + \bar{x}_5}{5} = \frac{.494 + .496 + .5 + .505 + .499}{5}$$

$$\boxed{\bar{\bar{x}} = .4988}$$

$$\bar{R} = \frac{R_1 + R_2 + R_3 + R_4 + R_5}{5} = \frac{.017 + .005 + .028 + .022 + .018}{5}$$

$$= \frac{.090}{5} = .018$$

$$\boxed{\bar{R} = .018}$$

Trial Control limits for $\bar{x}$ -chart

$$CL = \bar{\bar{x}} = .4988$$

$$UCL = \bar{\bar{x}} + A_2 \bar{R} = .4988 + (1.02) \times .018 = .5172$$

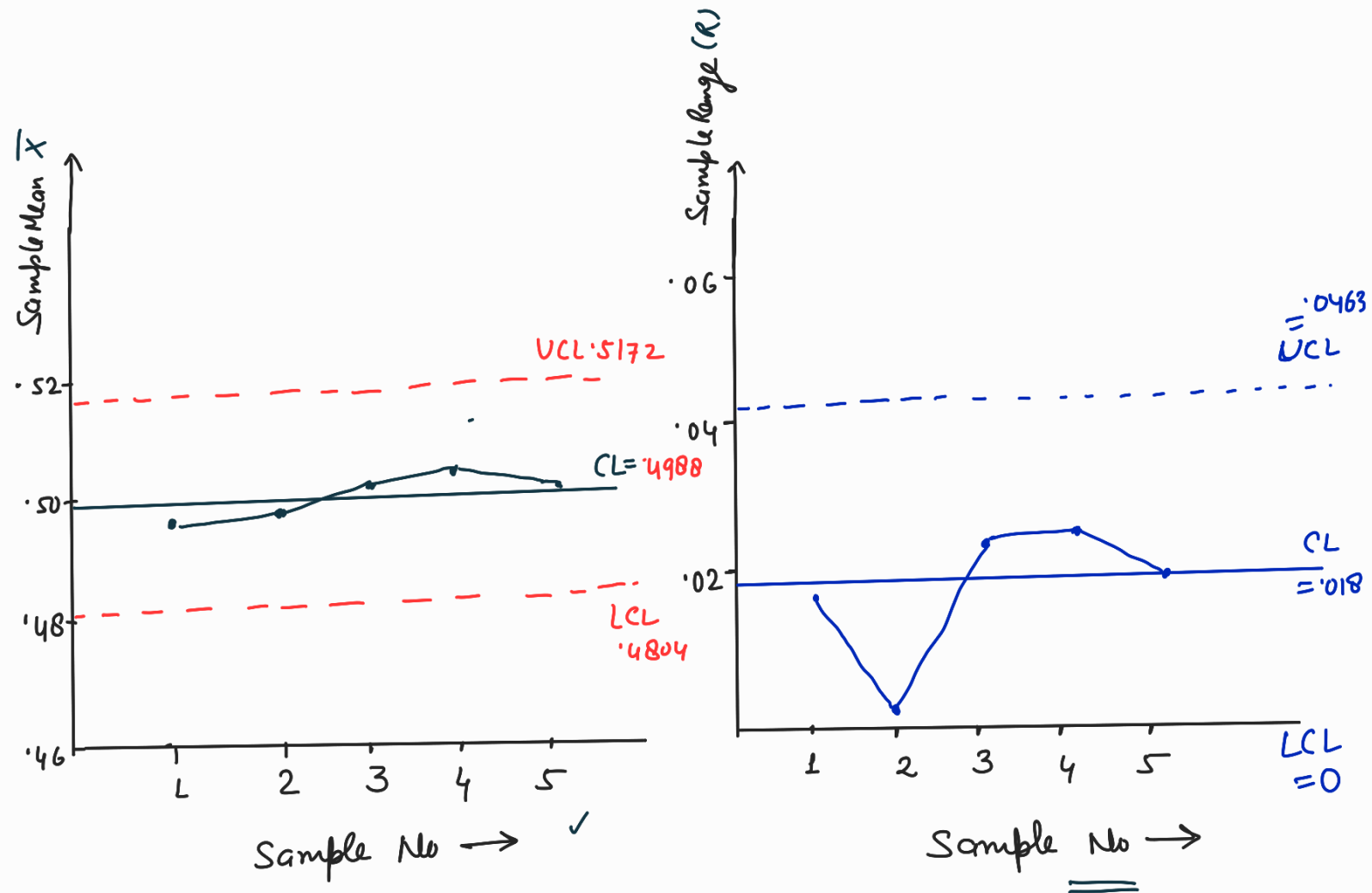
$$LCL = \bar{\bar{x}} - A_2 \bar{R} = .4988 - (1.02) \times .018 = .4804$$

Control limits for R-charts

$$CL = \bar{R} = .018$$

$$UCL = D_4 \bar{R} = 2.57 \times .018 = .0463$$

$$LCL = D_3 \bar{R} = 0 \times .018 = 0$$



Process is in statistical control

A. drilling machine bores holes with a mean diameter of .5230 cm and a standard deviation of .0032 cm. Calculate 2σ and 3σ upper and lower control limit for means of sample of 4

[Hint: $\rightarrow 2\sigma$ limit

$$\left. \begin{aligned} CL &= \bar{\bar{x}} = .5230 \\ UCL &= \bar{\bar{x}} + 2 \frac{\sigma}{\sqrt{n}} = .5262 \\ LCL &= \bar{\bar{x}} - 2 \frac{\sigma}{\sqrt{n}} = .5198 \end{aligned} \right\}$$

3 σ limit

$$\left. \begin{aligned} CL &= \bar{\bar{x}} = .5230 \\ UCL &= \bar{\bar{x}} + 3 \frac{\sigma}{\sqrt{n}} = .5278 \\ LCL &= \bar{\bar{x}} - 3 \frac{\sigma}{\sqrt{n}} = .5182 \end{aligned} \right\}$$

Topic: **Statistical Quality Control (SQC)** ✓**Control charts for Attributes** ✓✓ (p-chart, np-chart, C-chart) ✓✓✓

C-chart → C-chart is used when no. of defects per unit are counted. They are not classified as 'defective' or 'Not defective'.

C = No. of defects per Unit ←

C follows Poisson distribution

$$✓ \bar{C} = \text{Mean of } C = \frac{\text{No of defects in all samples}}{\text{Total No of samples}} ✓$$

$$\text{S.D } \sigma_C = \sqrt{\bar{C}} ←$$

Thus 3σ limits for C-charts

$$✓ UCL_C = \bar{C} + 3\sqrt{\bar{C}} \quad \bar{C} + 3\sigma$$

$$✓ LCL_C = \bar{C} - 3\sqrt{\bar{C}} \quad \bar{C} - 3\sigma$$

$$✓ CL = \underline{\bar{C}} ←$$

Ex. Draw a C-chart for following data pertaining to the number of foreign coloured threads (Considered as defects) in 15 piece of cloths of $2m \times 2m$ in a certain make of synthetic fibre and state your conclusion

(7), (12), 3, (20), (21), 5, 4, 3, 10, 8, 0, 9, 6, 7, (20) ←

Sol No of cloth piece = 15 ←

$$\text{Total No of defects} = 7 + 12 + 3 + 20 + 21 + 5 + 4 + 3 + 10 + 8 + 0 + 9 + 6 + 7 + 20 = \underline{135} ✓$$

$$\bar{C} = \frac{\text{Total defects}}{\text{Total samples}} = \frac{135}{15} = \underline{9} ✓$$

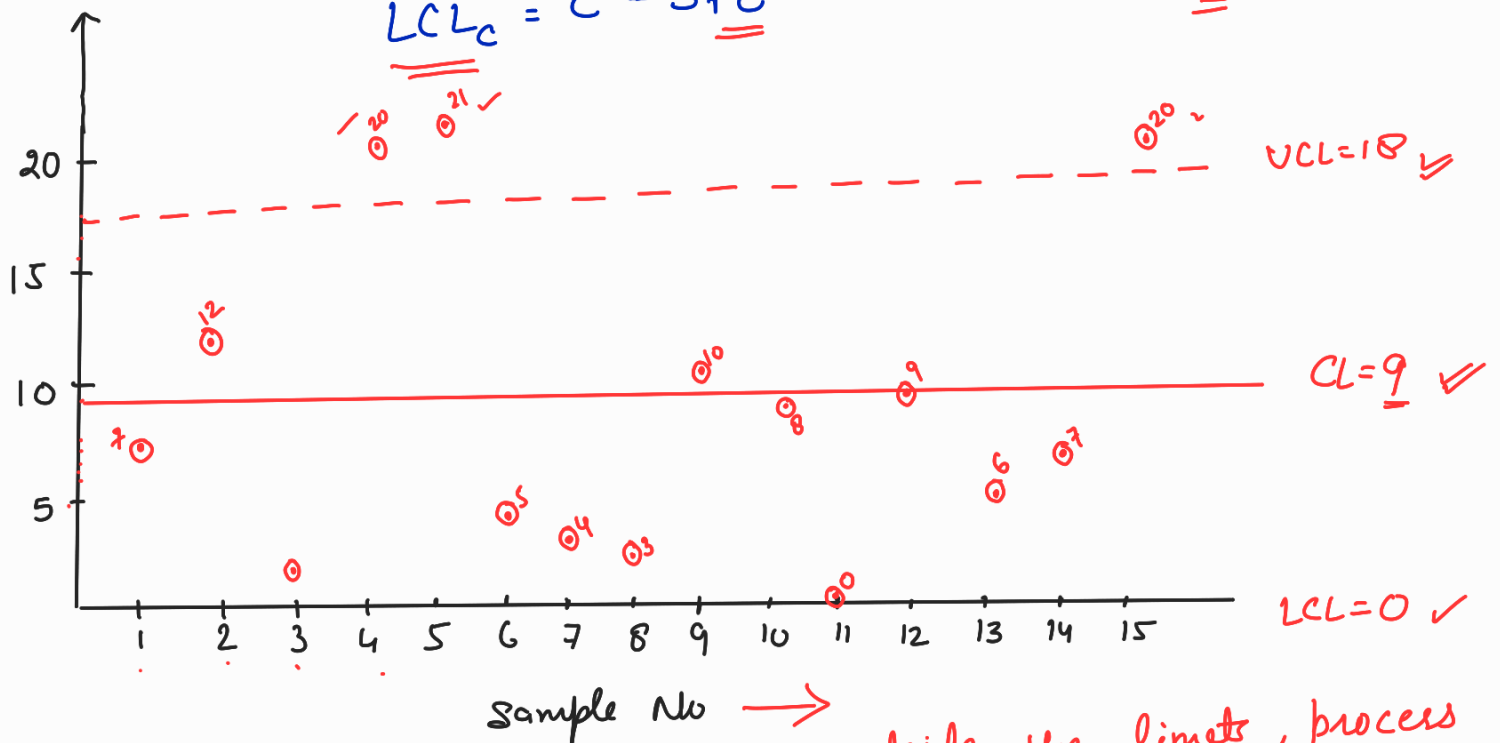
3 σ control limits

$$\sigma = \sqrt{\bar{c}}$$

$$CL = \bar{c} = 9$$

$$UCL_c = \bar{c} + 3\sqrt{\bar{c}} = 9 + 3\sqrt{9} = 18$$

$$LCL_c = \bar{c} - 3\sqrt{\bar{c}} = 9 - 3\sqrt{9} = 0$$



Three sample points are outside the limits, process is not under statistical control.

Fraction defective chart (P-chart)

This is designed to control the proportion (p) percentage (100p)

Or defective per sample

$$\text{Fraction defective } (\bar{p}) = \frac{\text{No. of defective item}}{\text{Sample size}} = d/n$$

$$\text{Mean fraction defective} = \bar{p}$$

$$\bar{p} = \frac{\text{Total No of defectives in all sample}}{\text{Total no of items in all samples}}$$

The statistical theory Binomial distribution is used to construct p-chart.

$$\text{Mean fraction defective} = \bar{p}$$

Mean value of fraction defective = $\bar{p}$

" " S.D = $\sqrt{\frac{\bar{p}(1-\bar{p})}{n}}$ ✓ = σ_p

$$\checkmark UCL_p = \bar{p} + 3\sigma_p = \bar{p} + 3\sqrt{\frac{\bar{p}(1-\bar{p})}{n}} \quad \bar{p} + 3\sigma_p$$

$$\checkmark LCL_p = \bar{p} - 3\sigma_p = \bar{p} - 3\sqrt{\frac{\bar{p}(1-\bar{p})}{n}} \quad \bar{p} - 3\sigma_p$$

Note → In sometime, LCL comes out negative, it is taken as zero. ✓✓

✱ It is used to control fraction defectives when sample size does not remain uniform or it varies.

$n\bar{p}$ -chart → Use when sample size is constant.

$$\bar{p} = \frac{\sum np}{\sum n} \checkmark$$

$$\checkmark CL = n\bar{p}$$

$$\checkmark \sigma_{np} = n\sigma_p = n\sqrt{\frac{\bar{p}(1-\bar{p})}{n}} = \sqrt{n\bar{p}(1-\bar{p})} \checkmark$$

$$\checkmark UCL_{np} = n\bar{p} + 3\sigma_{np} = n\bar{p} + 3\sqrt{n\bar{p}(1-\bar{p})} \checkmark$$

$$\checkmark LCL_{np} = n\bar{p} - 3\sigma_{np} = n\bar{p} - 3\sqrt{n\bar{p}(1-\bar{p})} \checkmark$$

Note → if LCL comes out negative, it is taken as zero.

Ex. The average percentage of defectives in 27 samples of size 1500 each was found to be 13.7%.

Construct a suitable control chart for this situation.
Discuss how the control chart can be used to control quality.

$$k = 27$$

$$n = 1500 \checkmark$$

$$\bar{p} = .137$$

$$\sigma_p = \sqrt{\frac{\bar{p}(1-\bar{p})}{n}}$$

$$CL = \bar{p} = .137$$

$$UCL = \bar{p} + 3\sigma_p = .137 + 3\sqrt{\frac{.137(1-.137)}{1500}}$$

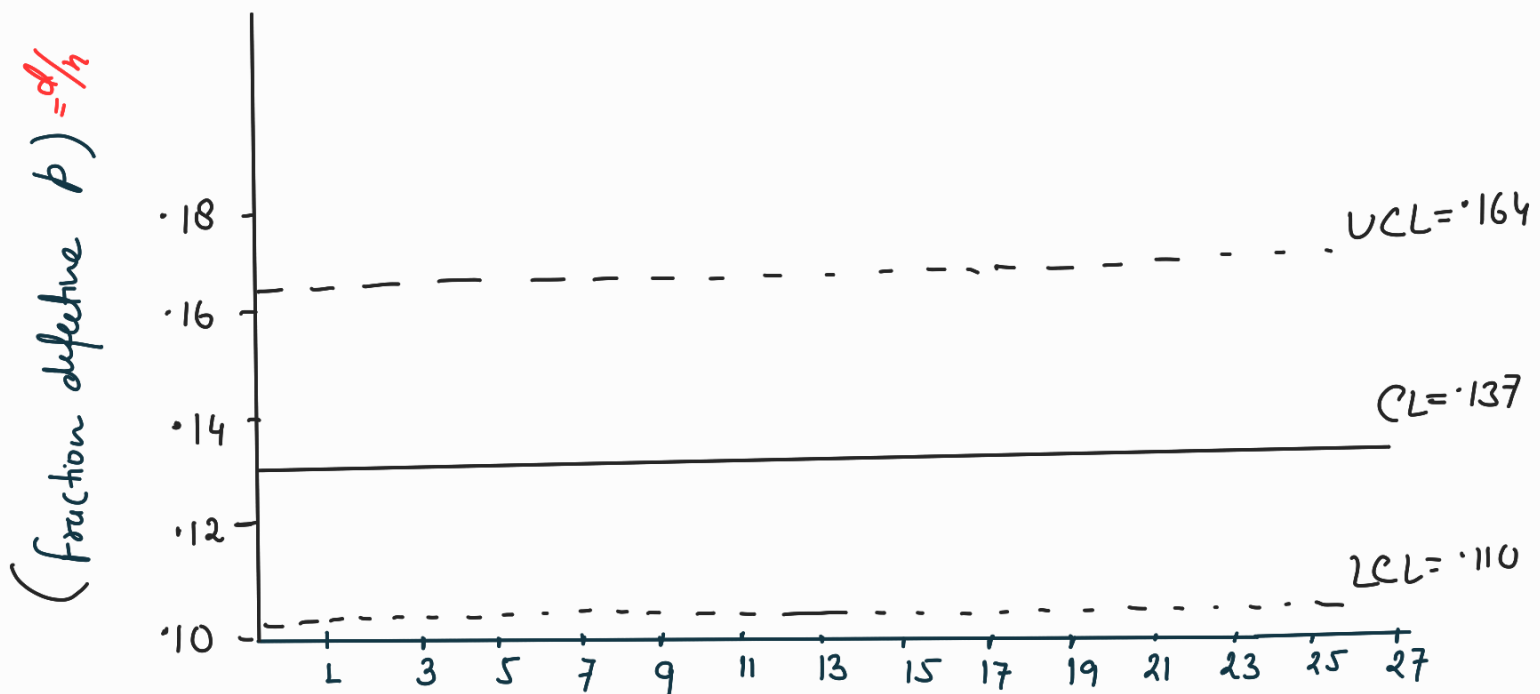
$$= .137 + .027$$

$$UCL = .164$$

$$LCL = \bar{p} - 3\sigma_p = .137 - 3\sqrt{\frac{.137(1-.137)}{1500}}$$

$$= .137 - .027$$

$$LCL = .110$$



Sample No $\rightarrow$

Ex. In a blade manufacturing factory, 1000 blades are examined daily. Draw the np chart for the following table and examine whether the

the following table and examine whether the process is under control? [2010]

Date : $\rightarrow \rightarrow$

No of defective blades : $\rightarrow$

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
9	10	12	8	7	15	10	12	10	8	7	13	14	15	16

Solⁿ

$n = 1000$ ✓

Total No of items = $1000 \times 15 = 15000$ ✓

Total No of defective = $\sum d$

$= 9 + 10 + 12 + 8 + 7 + 15 + 10 + 12 + 10 + 8 + 7 + 13 + 14 + 15 + 16$

$= 166$ ✓

$\bar{p} = \frac{\sum d}{N} = \frac{166}{15000} = 0.011$

$n\bar{p} = 1000 \times 0.011 = 11$

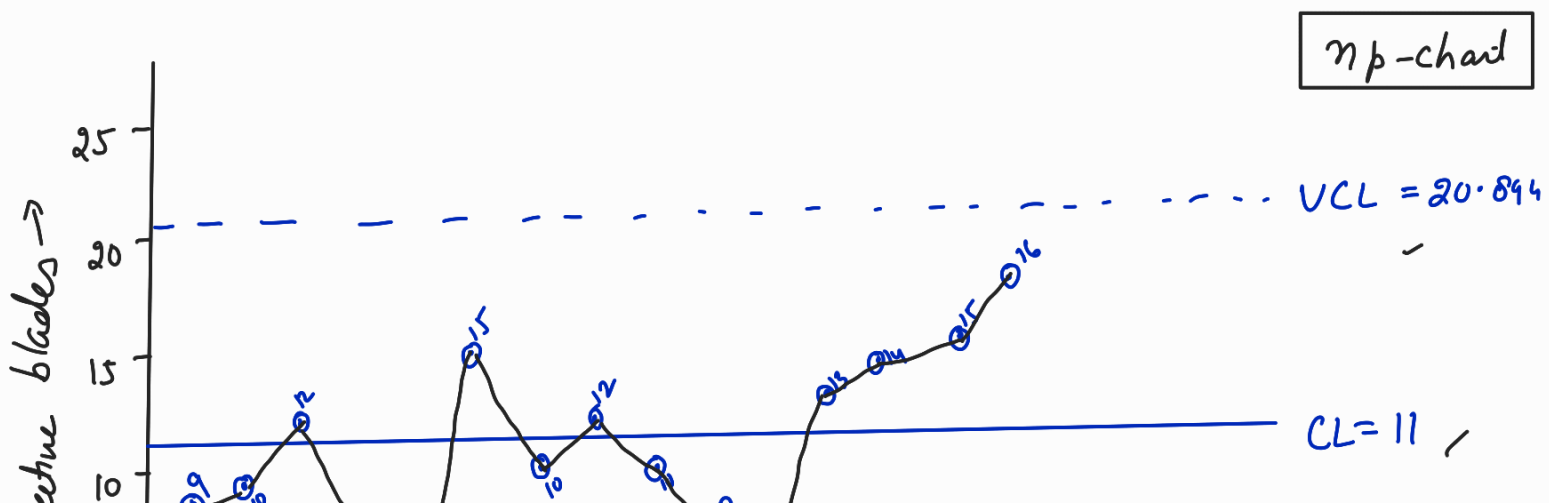
$CL = n\bar{p} = 11$ ✓

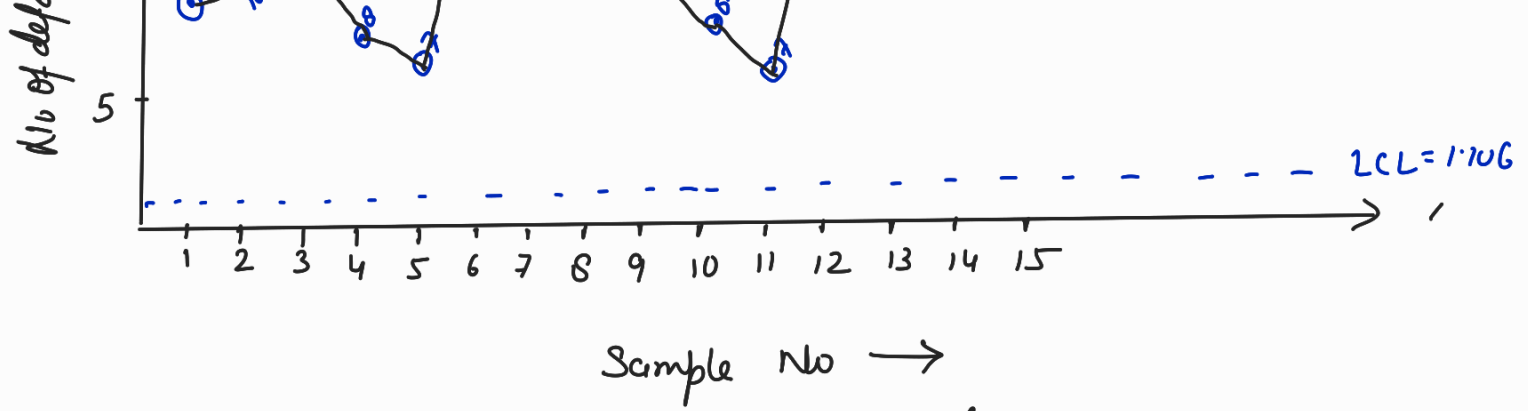
$VCL = n\bar{p} + 3\sqrt{n\bar{p}(1-\bar{p})} = n\bar{p} + 3\sigma_{np} = 11 + 3\sqrt{11(1-0.011)}$

$= 20.894$

$LCL = n\bar{p} - 3\sqrt{n\bar{p}(1-\bar{p})} = n\bar{p} - 3\sigma_{np} = 11 - 3\sqrt{11(1-0.011)}$

$= 1.106$





process is under control

Practice Question

Q $\rightarrow$ Distinguish between p-chart and np-chart. Following is the data of defective of 10 samples of size 100 each. Construct np-chart and give your comment

Sample No	1	2	3	4	5	6	7	8	9	10
No of defectives	6	9	12	5	12	8	8	16	13	7

[GBTU 2010, 2020-21]

[Hint $n = 100$, $\bar{p} = .096$, $n\bar{p} = 9.6$
 $CL = 9.6$, $UCL = 18.44$, $LCL = .762$
 process is in statistical control]