

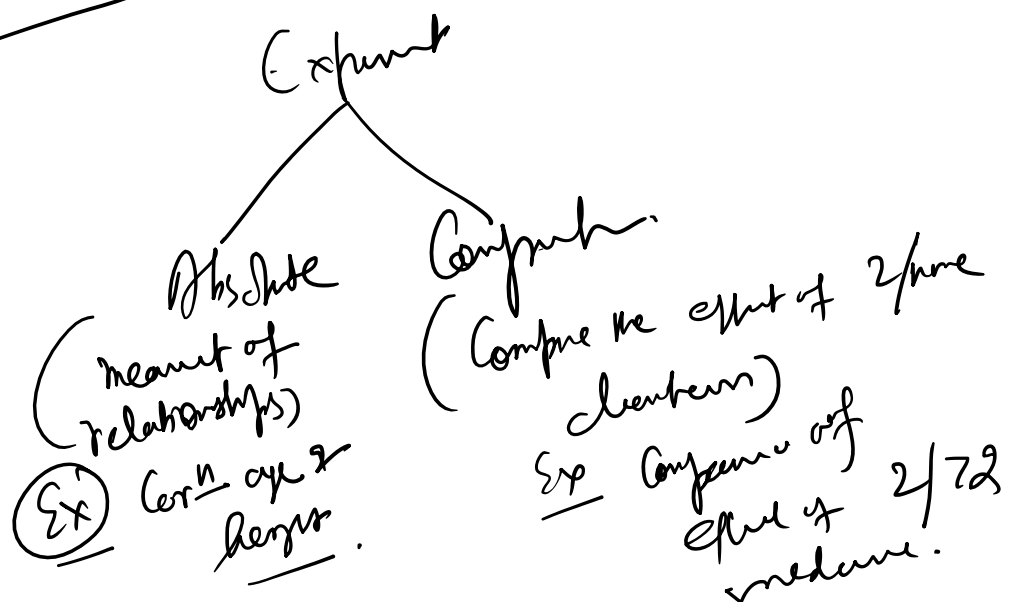
Random Data

RBD / D of Exp

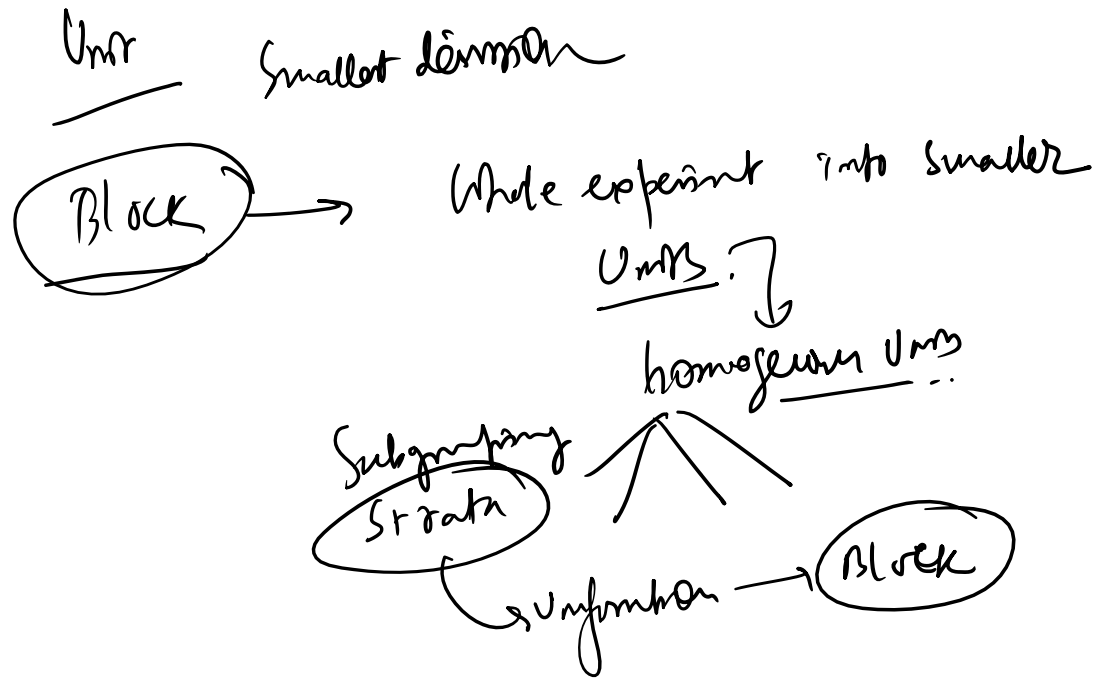
9062395123

well designed ??
 quickly & properly when answer can be taken

Experiment



2020 ⁴⁰ medane.
 cond
 samples



Precision Degree of Uncertainty

$$= \frac{1}{\sqrt{r}} = \frac{1}{\sigma^2/r} = \frac{r}{\sigma^2}$$

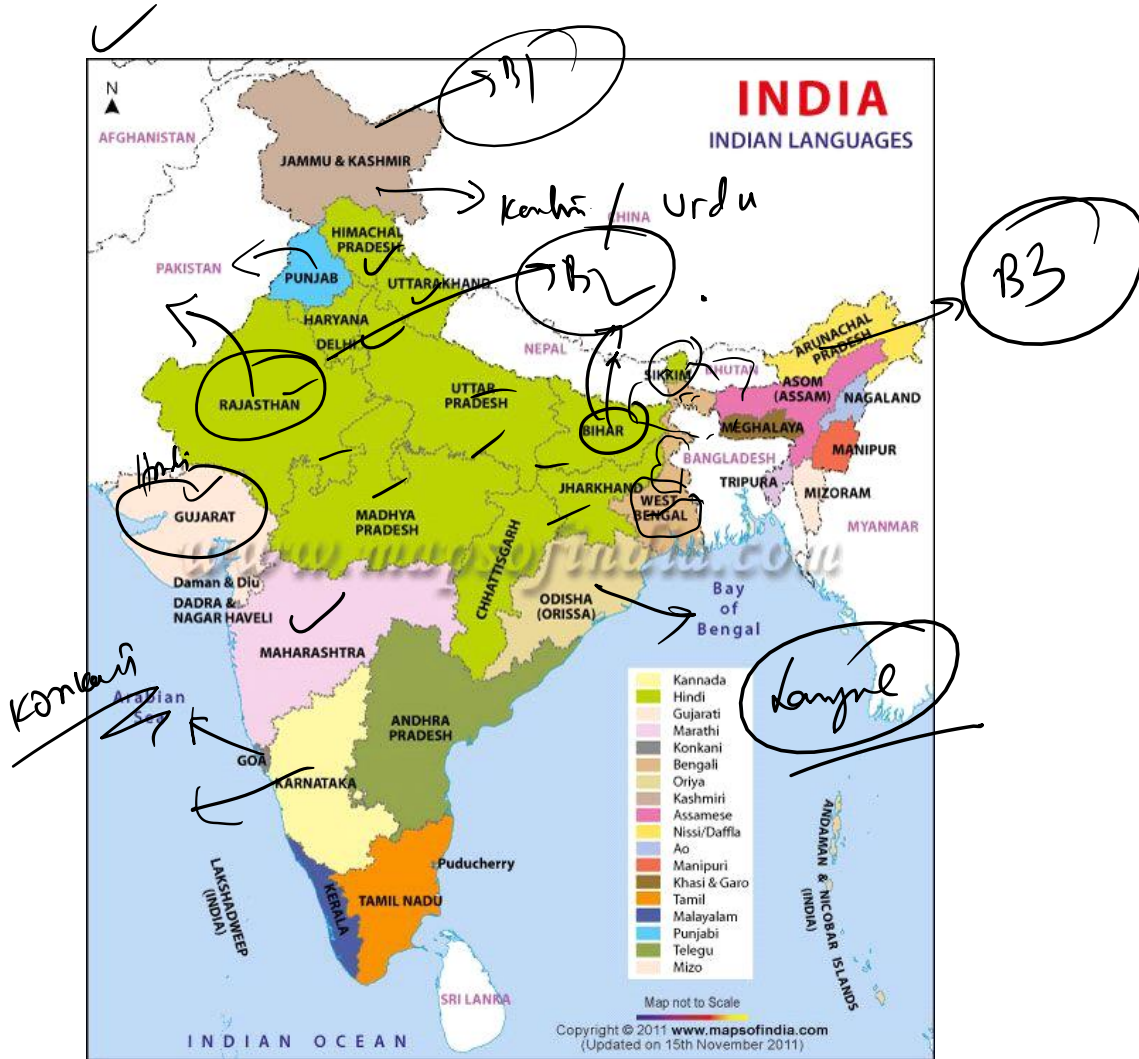
2003

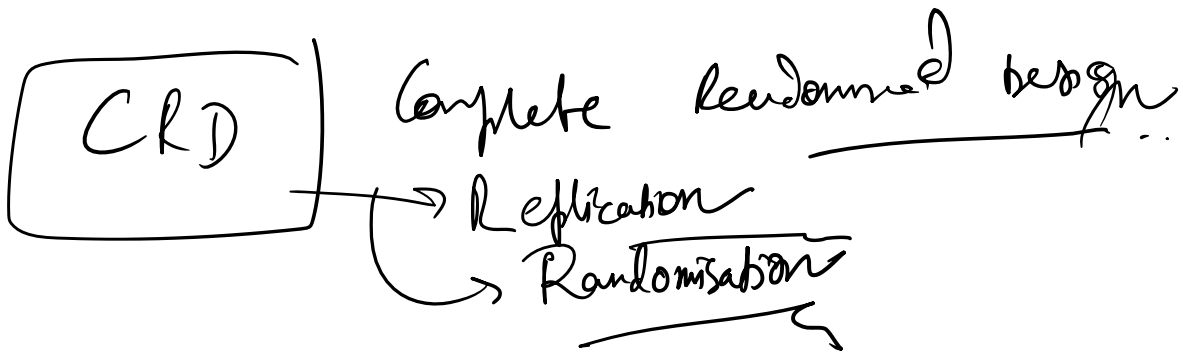
$r \rightarrow$ treatment reproducibility
 $\sigma^2 \rightarrow$ variance ...

✓

असल लेख

Experimental error ↓
by diving the homogenous data into
Blocus ..





Layout

Let us suppose that we have p treatments and i^{th} treatment is replicated n_i times ($i = 1, 2, \dots, p$).

The layout is given as:

Treatments	Replications				Total ($y_{i.}$)
✓ 1	y_{11}	y_{12}	...	y_{1n_1}	$y_{1.}$
✓ 2	y_{21}	y_{22}	y_{23}	y_{2n_2}	$y_{2.}$
...
✓ p	y_{p1}	y_{p2}	...	y_{pn_p}	$y_{p.}$

equal prob for all experiments

1 2 3 4 5

ANOVA

S_D^2

1 way ANOVA

Model

$$y_{ij} = \mu + \alpha_i + \epsilon_{ij}$$

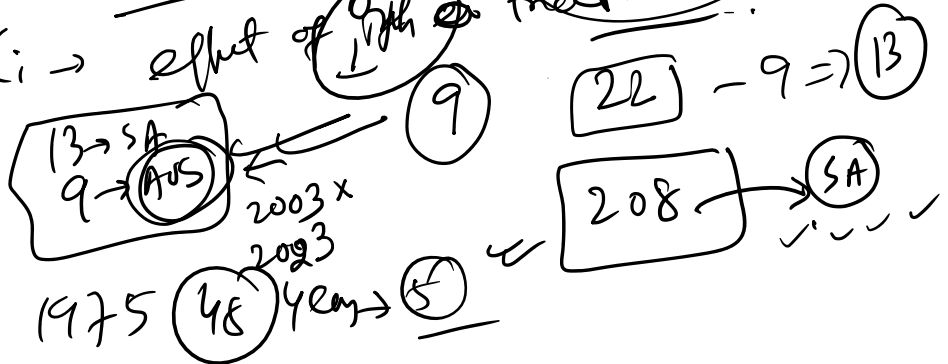
$y_{ij} \rightarrow j^{th}$ obsⁿ of i^{th} treatment

$i = 1, 2, \dots, n$
 $j = 1, 2, \dots, n_i^0$

$\mu =$ Grand mean effect

$\alpha_i \rightarrow$ effect of i^{th} treatment

$\epsilon_{ij} =$ Error



Model assumption $\sum_{i=1}^p \alpha_i = 0$ one $\hat{\sigma}^2$
 $N(0, \sigma^2)$

ANOVA Table for CRD

Source of variation	Degree of Freedom (d.f.)	Sum of squares (SS)	Mean sum of squares (MSS)	F-ratio
Treatment	$p - 1$	SST	$MST = \frac{SST}{p-1}$	$F_T = \frac{MST}{MSE}$
Error	$n - p$	SSE	$MSE = \frac{SSE}{n-p}$	
Total	$n - 1$	TSS		

eq - unu = df

$$df \geq 0$$

where

- $TSS = \sum_{i=1}^p \sum_{j=1}^{n_i} y_{ij}^2 - \frac{G^2}{n}$; $SST = \sum_{i=1}^p \frac{y_i^2}{n_i} - \frac{G^2}{n}$
- $SSE = TSS - SST$
- here y_i = total for the i^{th} treatment.
- $G = \sum_{i=1}^p \sum_{j=1}^{n_i} y_{ij} = \text{Grand total}$

$H_0 \rightarrow$

$$\alpha_1 = \alpha_2 = \dots = \alpha_p = 0$$

$$\mu_1 = \mu_2 = \dots = \mu_p$$

H_1 : effects of treatments are not same

$$F_T = \frac{MST}{MSE}$$

Advantages of CRD

The layout of design is easy.

Very useful to conduct small experiments.

The no. of replications need not to be same for each treatment.

The CRD provides maximum d.f. for estimation of error variance, which increases the sensitivity or the precision of the experiments when the number of treatments are small.

The statistical analysis remains simple even if some or all observations for any treatment are rejected or lost or missing for some purely accidental reasons.

CRD results in the maximum use of the experimental units since all the experimental material can be used.

Disadvantages of CRD

If experimental materials are not homogenous, the design suffers from the disadvantage of being inherently less informative than other more sophisticated designs.

Not suited for a large number of treatments. (very large) $n \rightarrow \infty$.

Uses of CRD

- Under conditions where the experimental material is homogenous e.g. in physics, chemistry and biological experiment for some green house studies.
- CRD may be used in a chemical or baking experiment where the experimental units are the part of the thoroughly mixed chemical or powder.



A company is considering three different covers for boxes of a brand of cereal. Box type A has picture of a sports hero eating the cereal, type B has a picture of a child eating the cereal, and type C has a picture of a bowl of the cereal. The company wants to determine which cereal box type provides for the most sales. Eighteen test markets were selected by the company and each box type was randomly assigned to six markets. The number of boxes of this cereal sold per 10000 population in a specified period is recorded for each test market. The data are as follows:

Type A	52.4	47.8	52.4	51.3	50	52.1
Type B	50.1	45.2	46	46.5	47.4	46.2
Type C	49.2	48.3	49	47.2	48.6	48.2



H_0 : no difference

H_1 : difference is there ..

$p = 3$ $n_i = 6$ $(i=1,2,3)$
 $n = 18$ let $n_i = 9$

Test whether there is any significant difference in the mean of three means.

$$TSS = \sum_{i=1}^p \sum_{j=1}^{n_i} y_{ij}^2 - \frac{G^2}{n} = (52.4^2 + 47.8^2 + \dots + 48.2^2) - \frac{8779^2}{18}$$

$$SST = \frac{1}{6} (306^2 + 281.4^2 + 290.5^2) - \frac{8779^2}{18} = 85.40 - \frac{8779^2}{18} = 57.57$$

$$SSE = TSS - SST = 85.40 - 57.57 = 33.83$$



ANOVA Table

Source of variation	d.f.	Sum of squares	Mean sum of squares	F-ratio
Treatment	3-1=2	51.57	25.79	11.41
Error	18-1=15	33.83	2.26	
Total	17	85.40		

Calc value > table value
1 to X

- The tabulated value ($F_{(2,15)}(0.05)$) = 3.86
- Hence F-ratio > tabulated (11.41 > 3.86), so we reject null hypothesis.
- Thus, there is a significant difference between the means of three treatments at the 5% level of significance.

RBD Larger number of treatments
Volume Comparison

(directional) → blocks..

Layout

← RBD

Let us suppose that we have p treatments and q blocks.

The layout is as follows:

Blocks \ Treatments	1	2	...	q	Total ($y_{i.}$)
1	y_{11}	y_{12}	...	y_{1q}	$y_{1.}$
2	y_{21}	y_{22}	...	y_{2q}	$y_{2.}$
...
p	y_{p1}	y_{p2}	...	y_{pq}	$y_{p.}$
Total ($y_{.j}$)	$y_{.1}$	$y_{.2}$...	$y_{.q}$	$y_{..}$

$$\sum y_{1.} + y_{2.} + \dots + y_{p.}$$

Model $y_{ij} = \mu + \alpha_i + \beta_j + \epsilon_{ij}$

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$\beta_j \rightarrow$ effect of j^{th} block
 $\alpha_i \rightarrow$ effect of i^{th} treatment

ANOVA Table for RBD

Source of variation	Degree of Freedom (d.f.)	Sum of squares (SS)	Mean sum of squares (MS)	F-ratio
Treatment	$p - 1$	SST	$MST = \frac{SST}{p-1}$	$F_T = \frac{MST}{MSE}$
Blocks	$q - 1$	SSB	$MSB = \frac{SSB}{q-1}$	$F_B = \frac{MSB}{MSE}$
Error	$(p - 1)(q - 1)$	SSE	$MSE = \frac{SSE}{(p-1)(q-1)}$	
Total	$n - 1$	TSS		

2 way ANOVA

where

$$TSS = \sum_{i=1}^p \sum_{j=1}^q y_{ij}^2 - \frac{G^2}{n}$$

$$SST = \frac{1}{q} \sum_{i=1}^p y_i^2 - \frac{G^2}{n}$$

$$SSB = \frac{1}{p} \sum_{j=1}^q y_j^2 - \frac{G^2}{n}$$

$$SSE = TSS - SST - SSB$$

y_i = total for the i^{th} treatment

y_j = total for the j^{th} block.

$$G = \sum_{i=1}^p \sum_{j=1}^q y_{ij} = \nu = \text{Grand total}$$

Treatment

$$H_{0T} : \alpha_1 = \alpha_2 = \dots = \alpha_p = 0$$

Block $H_{0B} : \beta_1 = \beta_2 = \dots = \beta_q = 0$

Same yields means of varieties

$$H_{0A} : \mu_{1A} = \mu_{2A} = \mu_{3A} = \mu_{4A}$$

$$H_{0B} : \text{Yields of fields}$$

A seed company performs an experiment to compare four varieties of rice. Five fields are available for the study and each field is subdivided

A seed company performs an experiment to compare four varieties of rice. Five fields are available for the study and each field is subdivided into four plots of equal size. Each variety is randomly assigned to a plot, and the yield in bushels is recorded as follows:

		Field				
		1	2	3	4	5
Variety of rice	1	45	37	41	48	32
	2	47	41	38	46	37
	3	53	47	50	56	45
	4	38	32	40	43	29

Test whether there is any significant difference in the yield mean according to variety of rice and fields.

H₀: (Yield) ≠ fields

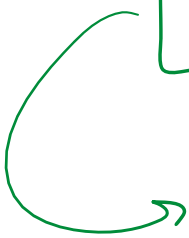
$p = 4, q = 5$

$n = pq = 20$

(F_{test})

LSD

→ Latin Square Design



RBD → that reduces the residual error in an experiment

↓
 Reducing the variability due to a controllable variable

$\frac{35}{2} \quad \frac{245}{2}$
210

Model

$$y_{ijk} = \mu + \alpha_i + \beta_j + \gamma_k + \epsilon_{ijk}$$

(CRD → RDB → LSD)

CRD \rightarrow ROB \rightarrow LSD

Only treatment \rightarrow CRD

treat + block \rightarrow RBD

treat + row + col \rightarrow LSD