

Completely Randomized Design (CRD) :-

The CRD is the simplest of all designs based on the principles of randomisation and replication. In this design  $t$  treatments are allocated at random to the experimental units over the entire experimental material.

Let us suppose we have  $t$  treatments, the  $i^{\text{th}}$  treatment being replicated  $n_i$  times  $i=1, 2, \dots, t$ . Then the whole experimental material is divided into  $n = \sum n_i$  experimental units and the treatments are distributed completely at random over the units subject to the condition that the  $i^{\text{th}}$  treatment occurs  $n_i$  times. In particular case if

$$n_i = r \quad \forall i=1, 2, \dots, t$$

i.e. if each treatment is repeated an equal no. of times  $r$ , then  $n = rt$  and randomisation gives every group of  $r$  units an equal chance of receiving the treatments.

Statistical analysis of CRD :- Statistical analysis of CRD is analogous to the ANOVA for a one-way classified data, the linear model becomes

$$Y_{ij} = \mu + \tau_i + \varepsilon_{ij} \quad \left( \begin{array}{l} i=1, 2, \dots, t \\ j=1, 2, \dots, r_i \end{array} \right)$$

where  $y_{ij}$  is the yield or response from the  $j$ th unit, receiving the  $i$ th treatment,  $\mu$  is the general mean effect,  $\tau_i$  is the effect due to the  $i$ th treatment and  $e_{ij}$  is error effect due to chance s.t.  $e_{ij}$  is identically and independently distributed (i.e.d.)  $N(0, \sigma_e^2)$  then

$n = \sum x_i$  is the total no. of experimental units. If we write

$$\sum_i \sum_j y_{ij} = Y_{..} = G = \text{Grand total of all the } n \text{ obs.}$$

$$\sum_j y_{ij} = y_{i.} = T_i = \text{Total response of the units receiving the } i\text{th treatment.}$$

Then as in ANOVA (one way classified data)

$$\sum_i \sum_j (y_{ij} - \bar{y}_{..})^2 = \sum_i \sum_j (y_{ij} - \bar{y}_{i.})^2 + \sum_i x_i (\bar{y}_{i.} - \bar{y}_{..})^2$$

$$\text{i.e. } T.S.S. = S.S.E. + S.S.T.$$

where T.S.S., S.S.T and S.S.E. are the total sum of squares, sum of squares due to treatments (betw. treatment S.S.) and sum of square due to error (i.e. within treatment S.S.) given respectively by

$$T.S.S. = \sum_i \sum_j (y_{ij} - \bar{y}_{..})^2$$

$$S.S.T. = \sum_i x_i (\bar{y}_{i.} - \bar{y}_{..})^2 = S_T^2 \text{ (say)}$$

$$S.S.E. = \sum_i \sum_j (y_{ij} - \bar{y}_{i.})^2 = S_E^2 \text{ (say)}$$

## ANOVA TABLE FOR CRD

| Source of variation | d.f.  | S.S.            | M.S.S.                           | Variance ratio (F)                    |
|---------------------|-------|-----------------|----------------------------------|---------------------------------------|
| Treatments          | $v-1$ | $S_T^2$         | $\Delta_T^2 = \frac{S_T^2}{v-1}$ | $F_T = \frac{\Delta_T^2}{\Delta_E^2}$ |
| Error               | $n-v$ | $S_E^2$         | $\Delta_E^2 = \frac{S_E^2}{n-v}$ |                                       |
| Total               | $n-1$ | $S_T^2 + S_E^2$ |                                  |                                       |

under the null hypothesis  $H_0: \tau_1 = \tau_2 = \dots = \tau_v$  against the alternative that all  $\tau$ 's are not equal. the statistic

$$F_T = \frac{\Delta_T^2}{\Delta_E^2} \sim F(v-1, n-v)$$

i.e.  $F_T$  follows  $F$  (central) distribution with  $(v-1, n-v)$  d.f.

If  $F_T > F_{\alpha}(v-1, n-v)$  then  $H_0$  is refuted at  $\alpha\%$

level of significance and we conclude that treatment differ significantly. If  $F_T < F_{\alpha}(v-1, n-v)$   $H_0$  may be accepted i.e. the data do not differ significantly and we conclude all treatments alike.

**Remark** - The following formulae for the calculation of various S.S. are much convenient to use from practical point of view

$$T.S.S. = \sum_j (y_{.j} - \bar{y}_{..})^2 = \sum_j y_{.j}^2 - n\bar{y}_{..}^2$$

- ① flexible ② easy analysis ③ all experimental material is used  
 ④ provides max<sup>m</sup> no of degrees of freedom increase precision of sensitivity.  
 ⑤ Inherently less informative ⑥ no restriction on randomisation

$$= \sum \sum y_{ij}^2 - n \left( \frac{\sum \sum y_{ij}}{n} \right)^2$$

$$= \sum \sum y_{ij}^2 - \frac{(\sum y_i)^2}{n} ; \quad n = \sum 1$$

$$\therefore \text{T.S.S.} = \text{Raw S.S.} - \text{Correction factor} \\ = \text{R.S.S.} - \text{C.F.}$$

$$\begin{aligned} \text{S.S.T.} &= \sum n_i (y_i - \bar{y})^2 \\ &= \sum n_i \bar{y}_i^2 - n \bar{y}^2 \\ &= \sum \left( \frac{T_i^2}{n_i} \right) - \text{C.F.} \end{aligned}$$

$$\begin{aligned} \therefore \text{S.S.P.} &= \text{T.S.S.} - \text{S.S.T.} \\ &= \sum \sum y_{ij}^2 - \sum \left( \frac{T_i^2}{n_i} \right) \\ &= \sum \sum y_{ij}^2 - \sum n_i \bar{y}_i^2 \end{aligned}$$

### SE of CRD

S.E. of the difference bet<sup>n</sup> any two treatment means is

$$\text{S.E.} = \sqrt{\frac{2}{r}} \\ \text{C.D.} = \sqrt{\frac{2}{r}} \times t_{0.05} \times \text{S.E.} \\ \text{Critical Difference (C.D.)}$$

is measure to find that when  $H_0$  is rejected which treatment mean differ significantly. i.e. the least difference bet<sup>n</sup> any two means to be significant.

Application : — CRD is mostly applicable in laboratory techniques, in chemical & biological methodology studies of cooked physio & chemical

### Randomised Block Design (R.B.D.)

This is a two-way design. In field experiment if the whole of the experimental area is not