

6-13. BALANCED INCOMPLETE BLOCK DESIGNS

The precision of the estimate of a treatment effect depends on the number of replications of the treatment *i.e.*, the larger is the number of replications, the more is the precision. A similar thing holds for the precision of the estimate of the difference between two treatment effects. If two treatments occur together in a block, then we can say that these are replicated once in that block. Similarly, if there are say, p blocks in a design in each of which the two treatments occur together, then the pair of treatments is said to be replicated p times in the design. The precision of the estimate of the difference between two treatments depends on the number of replications of the two treatments.

If in a block the number of experimental units or plots is smaller than the number of treatments, then the block is said to be incomplete and a design constituted of such blocks is called an '*incomplete block design*'.

As the name suggests, the balanced incomplete block designs are arranged in blocks or groups that are smaller than a complete replication in order to eliminate heterogeneity to a greater extent than is possible with randomised block design and Latin square design. These designs were introduced by *F. Yates* in a paper "A new method of arranging variety trials involving a large number of varieties" *Journal Agr. Sci.* 26, 424-455, 1936.

In factorial experiments confounding enables us to reduce the size of the block at the cost of information on certain treatment comparisons which may be relatively of less importance. But in Balanced Incomplete block designs (BIBD) which were developed for experiments in plant breeding and agriculture selection of all comparisons among pairs of treatments is made with equal precision.

Incomplete Block Design (I.B.D.). Definition. An incomplete block design is one having v treatments and b blocks each of size k such that each of the treatments is replicated r times and each pair of treatments occurs once and only once in the same block. v , b , r and k are known as the parameters of the I.B.D.

Balanced Incomplete Block Design (BIBD). Definition. An arrangement of v treatments in b blocks of k plots each ($k < v$) is known as BIBD, if

(i) each treatment occurs once and only once in r blocks and

(ii) each pair of treatments occurs together in λ blocks.

BIBD is used when all treatment comparisons are equally important as it ensures equal precisions of the estimates of all pairs of treatment effects. [Apart from ensuring equal precisions all the treatments, the variance of the estimate of any treatment variety mean, is σ^2/r , where σ^2 is the error variance and r the number of replicates, which is same for each treatment.]

6-13-1. Parameters of B.I.B.D. The integers v , r , b , k and λ are called the parameters of the B.I.B.D., where

v = number of varieties or treatments, b = number of blocks

k = block size, r = number of replicates for each treatment

λ = number of blocks in which any pair of treatments occurs together or number of times any two treatments occur together in a block. The following parametric relations serve as a necessary condition for the existence of a B.I.B.D.

(i) $vr = bk$, (ii) $\lambda(v - 1) = r(k - 1)$, and (iii) $b \geq v$ (Fisher's Inequality.)

We shall now establish these results as theorems 6-1 to 6-3 on B.I.B.D.

Theorem 6.1.

$$vr = bk \quad \dots(6-261)$$

Proof. Since there are v treatments each replicated r times, total number of plots in the design is vr . Further since there are b blocks each of size k , there are bk plots in all.

Hence, $vr = bk$

6-13-2. Incidence Matrix. Associated with any design D is the *incidence matrix* $N = (n_{ij})$, ($i = 1, 2, \dots, v ; j = 1, 2, \dots, b$), where n_{ij} denotes the number of times the i th treatment occurs in the j th block. Thus by the definition of a BIBD,

$$N = \begin{bmatrix} n_{11} & n_{12} & \dots & n_{1b} \\ n_{21} & n_{22} & \dots & n_{2b} \\ \vdots & \vdots & \vdots & \vdots \\ n_{v1} & n_{v2} & \dots & n_{vb} \end{bmatrix} \quad \dots(6-262)$$

where $n_{ij} = 1$, if i th treatment occurs in the j th block. $\dots(6-262a)$
 $= 0$, otherwise

Remark. Since in case of BIBD, n_{ij} can take only two values 0 or 1, BIBD is sometimes called a *binary design*.

We also observe, by definition of BIBD :

$$\sum_{j=1}^b n_{ij} = \sum_{j=1}^b n_{ij}^2 = r ; (i = 1, 2, \dots, v) \quad \dots(6-263)$$

$$\sum_{i=1}^v n_{ij} = \sum_{i=1}^v n_{ij}^2 = k ; (j = 1, 2, \dots, b) \quad \dots(6-263a)$$

$$\sum_{j=1}^b n_{ij} n_{lj} = \lambda ; (i \neq l = 1, 2, \dots, v), \quad \dots(6-263b)$$

since $n_{ij} n_{lj} = 1$ if and only if i th and l th treatments occur together in the j th block otherwise it is zero and they occur together in λ blocks.

If N' denotes the transpose of N then

$$NN' = \begin{bmatrix} \sum_j n_{1j}^2 & \sum_j n_{1j} n_{2j} & \dots & \sum_j n_{1j} n_{vj} \\ \sum_j n_{2j} n_{1j} & \sum_j n_{2j}^2 & \dots & \sum_j n_{2j} n_{vj} \\ \vdots & \vdots & \vdots & \vdots \\ \sum_j n_{vj} n_{1j} & \sum_j n_{vj} n_{2j} & \dots & \sum_j n_{vj}^2 \end{bmatrix} = \begin{bmatrix} r & \lambda & \lambda & \dots & \lambda \\ \lambda & r & \lambda & \dots & \lambda \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \lambda & \lambda & \lambda & \dots & r \end{bmatrix}_{v \times v} \quad \dots(6-264)$$

[From 6-263 and (6-263b)]

Theorem 6.2.

$$\lambda(v - 1) = r(k - 1). \quad \dots(6-265)$$

Proof. Let us denote by E_{mn} the $m \times n$ matrix all of whose elements are unity. From (6-264), we get

$$NN' E_{v1} = \begin{bmatrix} r & \lambda & \lambda & \dots & \lambda \\ \lambda & r & \lambda & \dots & \lambda \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \lambda & \lambda & \lambda & \dots & r \end{bmatrix}_{v \times v} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}_{v \times 1} = [r + \lambda(v - 1)] \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

Also $= [r + \lambda(v - 1)] E_{v1} \quad \dots(6-266)$

$$\begin{aligned}
 NN' E_{v1} &= N(N' E_{v1}) \\
 &= N \begin{bmatrix} n_{11} & n_{21} & \dots & n_{v1} \\ n_{12} & n_{22} & \dots & n_{v2} \\ \vdots & \vdots & & \vdots \\ n_{1b} & n_{2b} & \dots & n_{vb} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} = N \begin{bmatrix} \sum_i n_{i1} \\ \sum_i n_{i2} \\ \vdots \\ \sum_i n_{ib} \end{bmatrix} = N \begin{bmatrix} k \\ k \\ \vdots \\ k \end{bmatrix} \quad \text{[From (6.263a)]} \\
 &= k \begin{bmatrix} n_{11} & n_{12} & \dots & n_{1b} \\ n_{21} & n_{22} & \dots & n_{2b} \\ \vdots & \vdots & & \vdots \\ n_{v1} & n_{v2} & \dots & n_{vb} \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} = k \begin{bmatrix} \sum_j n_{1j} \\ \sum_j n_{2j} \\ \vdots \\ \sum_j n_{vj} \end{bmatrix} = k \begin{bmatrix} r \\ r \\ \vdots \\ r \end{bmatrix} \quad \text{[From (6.263)]} \\
 &= kr E_{v1} \quad \dots \text{(6.266a)}
 \end{aligned}$$

From (6.266) and (6.266a), we get

$$[r + \lambda(v - 1)] E_{v1} = kr E_{v1} \Rightarrow r + \lambda(v - 1) = kr \text{ i.e., } \lambda(v - 1) = r(k - 1), \text{ as desired.}$$

Aliter. v treatments gives rise to ${}^v C_2$ pairs and since each pair occurs λ times, the total number of times all the pairs occur in the design is $\lambda {}^v C_2$

Further since the size of each block is k , each block gives rise to ${}^k C_2$ pairs and since there are b blocks in all, the total number of treatment pairs in all the blocks is $b {}^k C_2$. Hence, we get

$$\lambda {}^v C_2 = b {}^k C_2 \Rightarrow \lambda v(v - 1) = bk(k - 1) \text{ i.e., } \lambda(v - 1) = \frac{bk}{v}(k - 1)$$

$$\therefore \lambda(v - 1) = r(k - 1), \text{ as desired.} \quad [\because vr = bk]$$

Theorem 6.3. $b \geq v$ (Fisher's Inequality).

Proof. From (6.84), the determinant of the matrix NN' is given by

$$|NN'| = \begin{vmatrix} r & \lambda & \lambda & \dots & \lambda \\ \lambda & r & \lambda & \dots & \lambda \\ \vdots & \vdots & \vdots & & \vdots \\ \lambda & \lambda & \lambda & & r \end{vmatrix}_{v \times v}$$

Adding 2nd, 3rd, ..., v th columns to the first column and taking $[r + (v - 1)\lambda]$ common from the first column, we get

$$NN' = [r + (v - 1)\lambda] \begin{vmatrix} 1 & \lambda & \lambda & \dots & \lambda \\ 1 & r & \lambda & \dots & \lambda \\ \vdots & \vdots & \vdots & & \vdots \\ 1 & \lambda & \lambda & \dots & r \end{vmatrix} = [r + (v - 1)\lambda] \begin{vmatrix} 1 & \lambda & \lambda & \dots & \lambda \\ 0 & (r - \lambda) & 0 & \dots & 0 \\ 0 & 0 & (r - \lambda) & \dots & 0 \\ \vdots & \vdots & \vdots & & \vdots \\ 0 & 0 & 0 & \dots & (r - \lambda) \end{vmatrix}$$

(Subtracting first row from the 2nd, 3rd, ..., v th row)

$$= [r + (v - 1)\lambda] (r - \lambda)^{v-1} \quad \text{(Expanding by first column)}$$

$$= rk(r - \lambda)^{v-1} \quad \text{[Using (6.265)]}$$

... (6.268a)

Thus $|N N'| \neq 0$, for if $r = \lambda$ then from (6-265), we get

$$(v - 1) = (k - 1) \Rightarrow v = k$$

indicating that the design reduces to randomised block design. Hence, $N N'$ is non-singular and consequently $\text{Rank}(N N') = v$, ... (6-269)

since v is the order of the matrix $N N'$.

But $\text{Rank}(N N') = \text{Rank}(N)$

$$\therefore \text{Rank}(N) = v \quad [\text{From (6-269)}] \quad \dots(6-270)$$

But since N is a $v \times b$ matrix, its rank can be at most b .

$$v = \text{rank } N \leq b \Rightarrow b \geq v, \text{ as desired.} \quad \dots(6-270a)$$

Deductions. (i) $r \geq k$

Proof. We have $vr = bk \Rightarrow r = \frac{b}{v} \cdot k$

Since $b \geq v$, we get $r \geq k$

(ii) $b \geq v + r - k \quad \dots(6-272)$

Proof. We have $v - k \geq 0$ and $r - k \geq 0$ [From (6-271)]

$$\therefore (v - k)(r - k) \geq 0 \Rightarrow \left(\frac{v}{k} - 1\right)(r - k) \geq 0 \text{ i.e., } \frac{v}{k}(r - k) - (r - k) \geq 0$$

$$\therefore \frac{vr}{k} - v \geq r - k \Rightarrow b \geq v + r - k \quad [\because vr = bk]$$

6-13-3. Symmetric BIBD

Definition. A BIBD is said to be symmetric if $b = v$ and $r = k$.

In this case the incidence matrix N is a square matrix.

Substituting $r = k$ and $b = v$ in (6-88a), we get

$$|N N'| = r^2 (r - \lambda)^{v-1} \Rightarrow |N| |N'| = r^2 (r - \lambda)^{v-1} \text{ i.e., } |N|^2 = r^2 (r - \lambda)^{v-1} \quad \dots(6-273)$$

$$\therefore |N| = \pm r (r - \lambda)^{\frac{v-1}{2}} \quad [\because |N| = |N'|]^v \quad \dots(6-274)$$

Since the determinant of the incidence matrix N is an integer, hence when v is even, $(r - \lambda)$ must be a perfect square.

Remarks 1. Necessary condition for a symmetrical BIBD with v as even is that $(r - \lambda)$ must be a perfect square.

2. If N is the incidence matrix of BIBD, then

- (i) every row sum is r ,
- (ii) every column sum is k ,
- (iii) the inner product of any two rows of N is λ .

For example, let us consider the following BIBD with $v = 4, b = 6, k = 2, r = 3, \lambda = 1$

BIBD Treatments

	1	2
1	1	2
2	1	3
3	1	4
4	2	3
5	2	4
6	3	4

$$N_{4 \times 6} = \begin{bmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 & 1 \end{bmatrix}$$

We note that every row sum in N is $3 = r$, every column sum is $2 = k$ and the inner product of any two rows, e.g., say 2nd and 3rd rows is

$$[1 \ 0 \ 0 \ 1 \ 1 \ 0] \begin{bmatrix} 0 \\ 1 \\ 0 \\ 1 \\ 0 \\ 1 \end{bmatrix} = 1 = \lambda$$

Theorem 6.4. In a symmetric BIBD, the number of treatments common between any two blocks is λ .

Proof. We have already proved in (6.84) that

$$N N' = \begin{bmatrix} r & \lambda & \lambda & \dots & \lambda \\ \lambda & r & \lambda & \dots & \lambda \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \lambda & \lambda & \lambda & \dots & r \end{bmatrix}_{v \times v} = \begin{bmatrix} r-\lambda & 0 & 0 & \dots & 0 \\ 0 & r-\lambda & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & r-\lambda \end{bmatrix} + \begin{bmatrix} \lambda & \lambda & \dots & \lambda \\ \lambda & \lambda & \dots & \lambda \\ \vdots & \vdots & \ddots & \vdots \\ \lambda & \lambda & \dots & \lambda \end{bmatrix} \dots(6-275)$$

where I_v is a unit matrix of order v .

Also for a symmetric BIBD, we have (for proof, see remark below) :

$$(N N')^{-1} = \frac{1}{r-\lambda} \left[I_v - \frac{\lambda}{r^2} E_{vv} \right] \Rightarrow (N')^{-1} N^{-1} = \frac{1}{r-\lambda} \left[I_v - \frac{\lambda}{r^2} E_{vv} \right] \dots(6-276)$$

Premultiplying by (N') , we get $N^{-1} = \frac{1}{r-\lambda} \left[N' - \frac{\lambda}{r^2} N' E_{vv} \right] \dots(6-277)$

But it can be easily verified that for symmetric BIBD

$$N' E_{vv} = N E_{vv} = r E_{vv} = k E_{vv} \Rightarrow \frac{N'}{r} E_{vv} = E_{vv} \quad (\because r=k) \dots(6-278)$$

Substituting in (6-277), we get $N^{-1} = \frac{1}{r-\lambda} \left[N' - \frac{\lambda}{r} E_{vv} \right]$

Post multiplying by N , we have

$$I_v = \frac{1}{r-\lambda} \left[N' N - \frac{\lambda}{r} N E_{vv} \right] = \frac{1}{r-\lambda} \left[N' N - \lambda E_{vv} \right] \quad \text{[From (6-278)]}$$

$$\Rightarrow N' N = (r-\lambda) I_v + \lambda E_{vv} \dots(6-279)$$

From (6-275) and (6-279), we get for a symmetric BIBD

$$N' N = N' N \quad \dots(6-280)$$

Thus, the inner product of any two rows of N is equal to the inner product of any two columns of N , i.e., λ .

Hence, in case of a symmetric BIBD, any two blocks have λ treatments in common.

Remark. $(N N')^{-1} = \frac{1}{|N N'|} \text{Adj}(N N')$... (1)

The cofactor of diagonal element of $N N'$ =
$$\begin{vmatrix} r & \lambda & \dots & \lambda \\ \lambda & r & \dots & \lambda \\ \vdots & \vdots & \ddots & \vdots \\ \lambda & \lambda & \dots & r \end{vmatrix}_{(v-1) \times (v-1)}$$

$= (r - \lambda)^{v-2} [r + (v - 2)\lambda] = (r - \lambda)^{v-2} [r + (v - 1)\lambda - \lambda]$ [c.f. (6-269)]
 $= (r - \lambda)^{v-2} [rk - \lambda]$ [$\because (v - 1)\lambda = r(k - 1)$]
 $= r^2 (r - \lambda)^{v-2} - \lambda (r - \lambda)^{v-2} = A$, (say). ($\because r = k$ for symmetric BIBD) ... (2)

The co-factor of off-diagonal element of $N N'$

$$= - \begin{vmatrix} \lambda & \lambda & \dots & \lambda \\ \lambda & r & \dots & \lambda \\ \vdots & \vdots & \ddots & \vdots \\ \lambda & \lambda & \dots & r \end{vmatrix}_{(v-1) \times (v-1)} = - \begin{vmatrix} \lambda & 0 & \dots & 0 \\ \lambda & (r - \lambda) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ \lambda & 0 & \dots & (r - \lambda) \end{vmatrix}_{(v-1) \times (v-1)}$$

$= -\lambda (r - \lambda)^{v-2} = B$, (say). ... (3)

Since $N N'$ is symmetric i.e., $N N' = N' N$, the transpose of the matrix of cofactors of elements of $N N'$ is same as the matrix of cofactors of $N N'$,

Also, for a symmetric BIBD,

$$|N N'| = r^2 (r - \lambda)^{v-1} = C$$
, (say). [From (6-273)] ... (4)

Substituting from (2), (3) and (4) in (1), we get

$$(N N')^{-1} = \begin{bmatrix} A/C & B/C & \dots & B/C \\ B/C & A/C & \dots & B/C \\ \vdots & \vdots & \ddots & \vdots \\ B/C & B/C & \dots & A/C \end{bmatrix}_{v \times v} = \frac{1}{(r - \lambda)} \begin{bmatrix} 1 - \frac{\lambda}{r^2} & -\frac{\lambda}{r^2} & \dots & -\frac{\lambda}{r^2} \\ -\frac{\lambda}{r^2} & 1 - \frac{\lambda}{r^2} & \dots & -\frac{\lambda}{r^2} \\ \vdots & \vdots & \ddots & \vdots \\ -\frac{\lambda}{r^2} & -\frac{\lambda}{r^2} & \dots & 1 - \frac{\lambda}{r^2} \end{bmatrix}_{v \times v}$$

$$= \frac{1}{(r - \lambda)} \left\{ \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix} + \begin{bmatrix} -\frac{\lambda}{r^2} & -\frac{\lambda}{r^2} & \dots & -\frac{\lambda}{r^2} \\ -\frac{\lambda}{r^2} & -\frac{\lambda}{r^2} & \dots & -\frac{\lambda}{r^2} \\ \vdots & \vdots & \ddots & \vdots \\ -\frac{\lambda}{r^2} & -\frac{\lambda}{r^2} & \dots & -\frac{\lambda}{r^2} \end{bmatrix} \right\}$$

$$\Rightarrow (N N')^{-1} = \frac{1}{r - \lambda} \left(\mathbf{1}_v - \frac{\lambda}{r} \mathbf{E}_{vv} \right) \quad \dots(6-281)$$

Theorem 6.5. If N is the incidence matrix of a symmetric BIBD, then

$$(N N') (N' N) = (r - \lambda) N' N + k^2 \lambda \mathbf{E}_{vv} \quad \dots(6-282)$$

Proof. For a symmetric BIBD, we have

$$N' N = (r - \lambda) \mathbf{I}_v + \lambda \mathbf{E}_{vv} \quad \text{[From (6-279)]}$$

$$\begin{aligned} \Rightarrow (N' N) (N' N) &= N' N [(r - \lambda) \mathbf{I}_v + \lambda \mathbf{E}_{vv}] = (r - \lambda) N' N + \lambda (N' N) \mathbf{E}_{vv} \\ &= (r - \lambda) N' N + \lambda N' (N \mathbf{E}_{vv}) = (r - \lambda) N' N + \lambda N' (k \mathbf{E}_{vv}) \quad \text{[From (6-278)]} \\ &= (r - \lambda) N' N + \lambda k (N' \mathbf{E}_{vv}) = (r - \lambda) N' N + \lambda k \cdot k \mathbf{E}_{vv} \quad \text{[From (6-278)]} \\ &= (r - \lambda) N' N + \lambda k^2 \mathbf{E}_{vv} \end{aligned}$$

6-13.4. Resolvable Design. Definition. A BIBD with parameters v, r, b, k and λ is said to be resolvable if the b blocks can be divided into r groups or sets of b/r blocks each, b/r being an integer, such that b/r blocks forming any of these sets give a complete replication of all the v treatments.

TABLE 6-76 : BIBD

Blocks	Treatments		
1	1	2	First set
2	3	4	
3	1	3	Second set
4	2	4	
5	1	4	Third set
6	2	3	

For example, let us consider the BIBD with parameters $v = 4, b = 6, r = 3, k = 2$ and $\lambda = 1$, as given in Table 6-76.

Here $b = 6$ blocks are divided into $r = 3$ sets each of $b/r = 6/3 = 2$ (integer) blocks. Moreover, each set contains each of the treatments occurring once and only once.

Also $\lambda = 1$.

Hence, the above design is resolvable BIBD.

Theorem 6.6. For a resolvable BIBD with parameters v, b, r, k, λ

$$b \geq v + r - 1 \quad \dots(6-283)$$

Proof. Since the design is resolvable, b/r must be an integer, equal to n (say), i.e.,

$$b/r = n \Rightarrow b = rn \quad \dots(6-284)$$

But for a BIBD

$$vr = bk \Rightarrow vr = rnk \quad \text{i.e.,} \quad v = nk \quad \dots(6-284a)$$

Also for a BIBD

$$\begin{aligned} r(k - 1) &= \lambda (v - 1) = \lambda (nk - 1) \quad \text{[From (6-284a)]} \\ \Rightarrow r &= \frac{\lambda (nk - 1)}{k - 1} = \frac{\lambda n (k - 1) + \lambda n - \lambda}{k - 1} = \lambda n + \frac{\lambda (n - 1)}{(k - 1)} \\ \therefore r - \lambda n &= \frac{\lambda (n - 1)}{(k - 1)} \quad \dots(6-285) \end{aligned}$$

Since r is an integer, and λ and n are also integers, from (6-285), we conclude that $\lambda(n - 1)/(k - 1)$ must be an integer.

Now, if possible let

$$\begin{aligned} & b < v+r-1, \quad \text{i.e.,} \quad b-r < r-1 \quad \dots(*) \\ \Rightarrow & r(n-1) < (v-1) \quad \text{[From (6-284)]} \\ \Rightarrow & r(n-1) < \frac{r(k-1)}{\lambda} \quad [\because r(k-1) = \lambda(v-1)] \\ \Rightarrow & \frac{\lambda(n-1)}{k-1} < 1, \end{aligned}$$

which is a contradiction of the fact that $\lambda(n-1)/(k-1)$ is an integer. Hence, our assumption (*) is wrong and we must have

$$b \geq v+r-1$$

Aliter. The incidence matrix N of the design consists of r sets of b/r rows each, where any set of rows is such that it occurs once and only once in each column of the set. By adding the 1st, 2nd, ..., $((b/r) - 1)$ th row, to the (b/r) th row of a set we obtain a row consisting of ones only. Moreover, since there are r sets, for each of these sets the rows add up to the same vector $(1, 1, \dots, 1)$. We know that if any elementary row transformation makes r row vectors identical then the rank is at least reduced by $(r-1)$. Hence using (6-270), we have

$$v = \text{Rank } N \leq b - (r-1) \Rightarrow b \geq v+r-1$$

6-13-5. Affine Resolvable Design.

Definition. A resolvable design is said to be affine resolvable if $b = r + v - 1$ and any two blocks from different sets have k^2/v treatments common where k^2/v is an integer.

For example, let us consider the resolvable design of § 6-10-3 with parameters $v = 4$, $b = 6$, $r = 3$, $k = 2$ and $\lambda = 1$.

We observe that the condition $b = v + r - 1$ is satisfied. Also $(k^2/v) = 1$ (integer) and any two blocks from different sets have only one treatment ($\lambda = 1$) common. Hence, the design is affine resolvable.

6-13-6. Analysis of Balanced Incomplete Block Design. (Intra Block Analysis of BIBD). This method of analysis developed by F. Yates by the use of standard Least Square Technique is sometimes referred to as the 'intra-block' analysis, i.e., the analysis without recovery of interblock information.

Consider 'a' units of material comprising b blocks to which v treatments are applied such that each treatment is replicated r times subject to the conditions of the BIBD.

Let $N = (n_{ij})$ be the incidence matrix of the BIBD, where

$$\begin{aligned} n_{ij} &= 1, \text{ if } i\text{th treatment occurs in the } j\text{th block} \\ &= 0, \text{ otherwise} \end{aligned}$$

Then as given in (6-263), (6-263a) and (6-263b), we have

$$\sum_{j=1}^b n_{ij} = \sum_{j=1}^b n_{ij}^2 = r \quad ; \quad \sum_{i=1}^v n_{ij} = \sum_{i=1}^v n_{ij}^2 = k \quad \text{and} \quad \sum_{j=1}^b n_{ij} n_{il} = \lambda, (i \neq l) \quad \dots(6-286)$$

Let y_{ij} ($i = 1, 2, \dots, v$; $j = 1, 2, \dots, b$) be the observation recorded on a unit of the material for the i th treatment which we suppose to be in the j th block where (i, j) runs through the set D for which $n_{ij} = 1$. Here the mathematical model is:

$$y_{ij} = \mu + t_i + b_j + \epsilon_{ij} \text{ for } (i, j) \in D \quad \dots(6-287)$$

where μ is the general mean effect, t_i ($i = 1, 2, \dots, v$) is the effect of the i th treatment, b_j ($j = 1, 2, \dots, b$) is the effect of the j th block and ϵ_{ij} are the $v \times r$ intra-block errors which are assumed to be independently normally distributed with mean zero and common variance σ_e^2 , i.e., ϵ_{ij} are i.i.d. $N(0, \sigma_e^2)$. In the *intra block analysis* we assume that the treatment effects and block effects are fixed, though unknown.

According to the principle of least squares, the normal equations for estimating the $(v + b + 1)$ parameters t_i , ($i = 1, 2, \dots, v$); b_j , ($j = 1, 2, \dots, b$) and μ are obtained on minimising the error or residual sum of squares.

$$E = \sum_{(i,j) \in D} \epsilon_{ij}^2 = \sum_{(i,j) \in D} (y_{ij} - \mu - t_i - b_j)^2$$

Equating to zero the partial derivatives of E w.r.t., μ , t_i and b_j , we get the normal equations as :

$$\frac{\partial E}{\partial \mu} = -2 \sum_{(i,j) \in D} (y_{ij} - \hat{\mu} - \tau_i - \beta_j) = 0 \quad \dots(6-287a)$$

$$\frac{\partial E}{\partial t_i} = -2 \sum_{j \in D_i} (y_{ij} - \hat{\mu} - \tau_i - \beta_j) = 0 \quad \dots(6-287b)$$

$$\frac{\partial E}{\partial b_j} = -2 \sum_{i \in D_j} (y_{ij} - \hat{\mu} - \tau_i - \beta_j) = 0 \quad \dots(6-287c)$$

where μ , τ_i and β_j are the least square estimates of μ , t_i and b_j respectively, D_i is the set of values for which $(i, j) \in D$ for given i and D_j is the set of k values for which $(i, j) \in D$ for given j .

In order that the set of equations (6-287a), (6-287b) and (6-287c) has an unique solution, we must impose the restrictions

$$\sum_{i=1}^v \tau_i = 0, \quad \sum_{j=1}^b \beta_j = 0 \quad \dots(6-288)$$

$$(6-287a) \text{ gives : } \sum_{(i,j) \in D} y_{ij} - rv \hat{\mu} - \sum_{(i,j) \in D} \tau_i - \sum_{(i,j) \in D} \beta_j = 0 \quad \dots(6-289)$$

$$\text{Now } \sum_{(i,j) \in D} \tau_i = \sum_i \left(\sum_{j \in D_i} \tau_i \right) = \sum_i (r \tau_i) = r \sum_i \tau_i = 0 \quad [\text{From (6-288)}] \quad \dots(6-290)$$

$$\text{Similarly, } \sum_{(i,j) \in D} \beta_j = \sum_j \left(\sum_{i \in D_j} \beta_j \right) = k \sum_j \beta_j = 0 \quad [\text{From (6-288)}] \quad \dots(6-290a)$$

Substituting in (6-289), we get

$$\hat{\mu} = \frac{\sum_{(i,j) \in D} y_{ij}}{rv} = \bar{y} \dots(6-291)$$

From (6-105b), we get

$$\sum_{j \in D_i} y_{ij} - r \hat{\mu} - r \tau_i - \sum_{j \in D_i} \beta_j = 0 \quad \Rightarrow \quad T_i = r y_{..} + r \tau_i + \sum_{j \in D_i} \beta_j$$

where $T_i = \sum_{j \in D_i} y_{ij}$ is the total yield for the i th treatment. Further, since D_i is the set of values for which $(i, j) \in D$ for given i , i.e., $n_{ij} = 1$ and i has a fixed value, we write

$$T_i = r \bar{y}_{..} + r \tau_i + \sum_{j=1}^b n_{ij} \beta_j \quad \dots(6-292)$$

From (6-292) we observe that the estimate of a treatment effect is no longer of the form of the observed treatment mean minus the grandmean, since the block effects do not enter in the same way into all the observed treatment means. For example, a treatment may be favoured by occurring only in blocks with high block effects.

Similarly from (6-287c), we shall get

$$B_j = k \bar{y}_{..} + k \beta_j + \sum_{i=1}^v n_{ij} \tau_i \quad \dots(6-293)$$

where $B_j = \sum_{i \in D_j} y_{ij}$ is the total yield from the j th block.

The quantity :
$$\frac{B_j}{k} - \bar{y}_{..} = \beta_j + \frac{1}{k} \sum_{i=1}^v n_{ij} \tau_i \quad \dots(6-294)$$

may be called the estimate of the j th block effect ignoring treatments.

Substituting the value of β_j from (6-293) in (6-292), we get

$$\begin{aligned} T_i &= r \bar{y}_{..} + r \tau_i + \frac{1}{k} \sum_{j=1}^b \left[n_{ij} \left(B_j - k \bar{y}_{..} - \sum_{p=1}^v n_{pj} \tau_p \right) \right] \\ \Rightarrow T_i - \frac{1}{k} \sum_{j=1}^b n_{ij} B_j &= r \bar{y}_{..} + r \tau_i - \bar{y}_{..} \sum_{j=1}^b n_{ij} - \frac{1}{k} \sum_{j=1}^b \sum_{p=1}^v n_{ij} n_{pj} \tau_p \\ &= r \tau_i - \frac{1}{k} \sum_{j=1}^b \sum_{p=1}^v n_{ij} n_{pj} \tau_p \quad \dots(6-295) \end{aligned}$$

The quantity :
$$Q_i = T_i - \sum_{j=1}^b \frac{n_{ij} B_j}{k} \quad \dots(6-296)$$

is called the i th adjusted treatment total or the adjusted total yield for the i th treatment. The adjustment consists in subtracting from the treatment total T_i the sum of the j th block average B_j/k (average yield per plot for the j th block) for the blocks in which the i th treatment occurs.

Thus from (6-294) and (6-295), we have

$$\begin{aligned} Q_i &= r \tau_i - \frac{1}{k} \sum_{j=1}^b \sum_{p=1}^v n_{ij} n_{pj} \tau_p \\ &= -\frac{\tau_1}{k} \sum_{j=1}^b n_{ij} n_{1j} - \frac{\tau_2}{k} \sum_{j=1}^b n_{ij} n_{2j} - \dots + \tau_i \left(r - \frac{1}{k} \sum_{j=1}^b n_{ij}^2 \right) - \dots - \frac{\tau_v}{k} \sum_{j=1}^b n_{ij} n_{vj} \\ &= -\frac{\lambda}{k} (\tau_1 + \tau_2 + \dots + \tau_v) + \left(r - \frac{r}{k} + \frac{\lambda}{k} \right) \tau_i \quad \text{[Using (6-286)]} \\ &= \left(r - \frac{r}{k} + \frac{\lambda}{k} \right) \tau_i \quad \left(\because \sum_{i=1}^v \tau_i = 0 \right) \\ &= \left[\frac{r(k-1) + \lambda}{k} \right] \tau_i = \left[\frac{\lambda(v-1) + \lambda}{k} \right] \tau_i \quad \text{[From (6-265)]} \end{aligned}$$

$$\therefore Q_i = \frac{\lambda v}{k} \tau_i \Rightarrow \tau_i = Q_i - \frac{k}{\lambda v}; (i = 1, 2, \dots, v) \quad \dots(6-297)$$

If we write : $E = \frac{\lambda v}{kr} \quad \dots(6-297a)$

Then, $\tau_i = \frac{Q_i}{rE} \quad \dots(6-297b)$

It may be remarked that $E < 1$, since

$$E = \frac{v}{k} \cdot \frac{\lambda}{r} = \frac{v(k-1)}{k(v-1)} = \frac{vk-v}{vk-k} < 1 \quad [\because \lambda(v-1) = r(k-1) \text{ and } v > k] \quad \dots(6-298)$$

The quantity $\left(\sum_{i=1}^v \tau_i Q_i \right)$ is the sum of squares due to treatments after adjusting for block effects. Thus

$$\text{S.S. Treatments (adjusted)} = \sum_{i=1}^v \tau_i Q_i = \sum_{i=1}^v \frac{Q_i^2}{rE} \quad \dots(6-299)$$

The quantity

$$\sum_{j=1}^b (B_j^2/k) - n \hat{\mu} \bar{y}_{..} = \sum (B_j^2/k) - \frac{1}{n} \left[\sum_{(i,j) \in D} y_{ij} \right]^2 = \sum_{j=1}^b (B_j^2/k) - \text{C.F.} \quad \dots(6-300)$$

Under the model (6-287), the residual sum of squares is given by :

$$\sum_{(i,j) \in D} (y_{ij} - \hat{\mu} - \tau_i - \beta_j)^2 = \sum_{(i,j) \in D} y_{ij}^2 - n \hat{\mu} \bar{y}_{..} - \sum_{i=1}^v T_i \tau_i - \sum_{j=1}^b B_j \beta_j$$

The sum of the squares due to $\hat{\mu}$, τ_i and β_j being given by ($n = vr = bk$)

$$\begin{aligned} S(\mu, b, \tau) &= n \hat{\mu} \bar{y}_{..} + \sum_{i=1}^v T_i \tau_i + \sum_{j=1}^b B_j \beta_j = \hat{\mu} \sum_{(i,j) \in D} y_{ij} + \sum_{j=1}^b B_j \beta_j + \sum_{i=1}^v T_i \tau_i \\ &= \hat{\mu} \sum_j \left(\sum_{i \in D_i} y_{ij} \right) + \sum_j B_j \beta_j + \sum_i T_i \tau_i = \hat{\mu} \sum_j B_j + \sum_j B_j \beta_j + \sum_i T_i \tau_i \\ &= \sum_j (\hat{\mu} + \beta_j) B_j + \sum_i T_i \tau_i \end{aligned}$$

Substituting for $\hat{\mu} + \beta_j$ from (6-293), we get

$$\begin{aligned} S(\mu, b, \tau) &= \sum_{j=1}^b \left[\frac{1}{k} \left(B_j - \sum_{i=1}^v n_{ij} \tau_i \right) B_j \right] + \sum_{i=1}^v T_i \tau_i \\ &= \sum_{j=1}^b (B_j^2/k) + \sum_{i=1}^v \tau_i \left(T_i - \sum_j n_{ij} B_j/k \right) = \sum_{j=1}^b (B_j^2/k) + \sum_{i=1}^v \tau_i Q_i \\ &= n \hat{\mu} \bar{y}_{..} + \left(\sum_{j=1}^b \frac{B_j^2}{k} - n \hat{\mu} \bar{y}_{..} \right) + \sum_{i=1}^v \tau_i Q_i \end{aligned}$$

is the sum of squares due to block ignoring treatments, i.e., it is (unadjusted) block sum of squares.

The analysis of variance table for Intrablock analysis of BIBD is given in Table 6-77.

TABLE 6.77 : ANOVA FOR B.I.B.D. (INTRABLOCK ANALYSIS)

Source	d.f.	Sum of Squares	E (M.S.S.)
Between blocks (unadjusted)	$b - 1$	$\frac{1}{k} \sum_{j=1}^b B_j^2 - (G^2/bk)$	
Between treatments (adjusted)	$v - 1$	$\sum_{i=1}^v (Q_i^2/rE)$	$\sigma_e^2 + rE \sum_{i=1}^v \frac{\tau_i^2}{v-1}$
Intra block error	$bk - b - v + 1$	By subtraction	σ_e^2
Total	$bk - 1$	$\sum_{(i,j) \in D} y_{ij}^2 - (G^2/bk)$ = R.S.S. - C.F.	

In the above table $G = \sum_{(i,j) \in D} y_{ij}$ is the grand total of all the observations and $\sum_{(i,j) \in D} y_{ij}^2$ is

the raw sum of squares (R.S.S.).

For testing the null hypothesis : $H_0 : t_1 = t_2 = \dots = t_v$, the treatments (adjusted) mean square is compared with the intrablock error mean square as usual.

If we are interested to make inferences about the block effects, the (adjusted) block mean sum of squares is to be compared with error mean square. The adjusted block sum of squares can be computed from the following identity :

$$\text{Block S.S. (adjusted) + Treatment S.S. (unadjusted)} = \text{Block S.S. (unadjusted) + Treatment S.S. (adjusted)} \quad \dots(6-301)$$

where,
$$\text{Treatment S.S. (unadjusted)} = \frac{1}{r} \sum_{i=1}^v T_i^2 - \text{C.F.} \quad \dots(6-301a)$$

6-13.7. Comparison of Two Treatments. The treatment contrast $t_i - t_p$ is estimated by :

$$\tau_i - \tau_p = \frac{1}{rE} (Q_i - Q_p) \quad [\text{From (6-297b)}]$$

$$\therefore \text{Var} (\tau_i - \tau_p) = \frac{1}{r^2 E^2} [\text{Var} (Q_i) + \text{Var} (Q_p) - 2 \text{Cov} (Q_i, Q_p)] \quad \dots(6-302)$$

$$\text{Var} (Q_i) = \text{Var} (T_i - \sum_j n_{ij} B_j/k)$$

$$= \text{Var} (T_i) + \sum_j \frac{n_{ij}^2}{k^2} \text{Var} (B_j) - 2 \sum_j \frac{n_{ij}}{k} \text{Cov} (T_i, B_j). \text{ But}$$

$$\text{Var} (T_i) = \text{Var} \left(\sum_{j \in D_i} y_{ij} \right) = r \sigma_e^2; \text{Var} (B_j) = \text{Var} \left(\sum_{i \in D_j} y_{ij} \right) = k \sigma_e^2 \text{ and Cov} (T_i, B_j) = n_{ij} \sigma_e^2 \dots(*)$$

$$\therefore \text{Var} (Q_i) = r \sigma_e^2 + \frac{\sigma_e^2}{k} \sum_j n_{ij}^2 - \frac{2 \sigma_e^2}{k} \sum_j n_{ij}^2$$

$$= r \left(1 + \frac{1}{k} - \frac{2}{k} \right) \sigma_e^2 = \frac{r(k-1)}{k} \sigma_e^2 \quad [\text{Using (6-286)}] \quad \dots(6-302a)$$

$$\text{Cov} (Q_i, Q_p) = \text{Cov} \left[\left\{ T_i - \left(\sum_j n_{ij} B_j/k \right) \right\} \left\{ T_p - \left(\sum_j n_{pj} B_j/k \right) \right\} \right]$$

$$\begin{aligned}
 &= \text{Cov}(T_i, T_p) - \text{Cov}\left[T_i, \left(\sum_j n_{pj} B_j/k\right)\right] \\
 &\quad - \text{Cov}\left[T_p, \left(\sum_j n_{ij} B_j/k\right)\right] + \text{Cov}\left[\left(\sum_j n_{ij} B_j/k\right) \left(\sum_j n_{pj} B_j/k\right)\right] \\
 &= 0 - \frac{1}{k} \sum_j n_{pj} \text{Cov}(T_i, B_j) - \frac{1}{k} \sum_j n_{ij} \text{Cov}(T_p, B_j) + \frac{1}{k^2} \sum_{j,j'} n_{ij} n_{pj'} \text{Cov}(B_j, B_{j'}) \\
 &= -\frac{\sigma_e^2}{k} \sum_j n_{pj} n_{ij} - \frac{\sigma_e^2}{k} \sum_j n_{ij} n'_{pj} + \frac{1}{k^2} \sum_{j,j'} n_{ij} n_{pj'} \text{Cov}(B_j, B_{j'}) \quad \text{[Using (*)]}
 \end{aligned}$$

But
$$\left. \begin{aligned}
 \text{Cov}(B_j, B_{j'}) &= 0, (j \neq j') \\
 &= k \sigma_e^2, (j = j')
 \end{aligned} \right\} \dots (**)$$

Hence, on using (**) and (6.104), we get

$$\text{Cov}(Q_i, Q_p) = -2 \frac{\sigma_e^2}{k} \cdot \lambda + \frac{1}{k^2} \cdot k \sigma_e^2 \lambda = -\frac{\lambda}{k} \sigma_e^2 \dots (6-302b)$$

Substituting from (6-302a) and (6-302b) in (6-302), we have

$$\begin{aligned}
 \text{Var}(\tau_i - \tau_p) &= \frac{1}{r^2 E^2} \left[\frac{2 \sigma_e^2 \cdot r(k-1)}{k} + \frac{2\lambda \sigma_e^2}{k} \right] \\
 &= \frac{2\lambda \sigma_e^2}{r^2 E^2 k} [v - 1 + 1] = \frac{2\lambda v \sigma_e^2}{k r^2 E^2} \quad [\because \lambda(v-1) = r(k-1)] \\
 &= \frac{2\sigma_e^2}{rE} \quad \left[\because E = \frac{\lambda v}{rk} \right] \dots (6-302)
 \end{aligned}$$

Thus, the test statistic for testing the hypothesis of equality of two treatments, i.e., $H_0: t_i = t_p$ becomes

$$t = \frac{|\tau_i - \tau_p|}{\left[2 s_e^2 / (rE)\right]^{1/2}} \dots (6-304)$$

where s_e^2 , the error mean (sum of) square is an unbiased estimate of σ_e^2 . The statistic t follows students' t -distribution with error $d.f.$

6-13-8. Efficiency of Balanced Incomplete Block Design Relative to Randomised Block Design. So far, as the experimental operations are concerned, incomplete block designs are no more difficult than randomised blocks. some extra planning is involved in drawing up and randomising the experimental plan, especially if care is to be taken to make the best possible grouping of the experimental units. The gain in accuracy over randomised blocks depends upon the type of the experimental material and may be expected to increase as the number of treatments is increased.

In case of a BIBD :
$$\text{Var}(\tau_i - \tau_p) = \frac{2\sigma_e^2}{rE} = \frac{2s_e^2}{rE} \dots (6-305)$$
 where s_e^2 is error mean square for BIBD.

In case of randomised block design :
$$\text{Var}(\tau_i - \tau_p) = \frac{2\sigma_e^2}{r} = \frac{2s_e^2}{r}$$

where σ_e^2 is the variance per unit in R.B.D. and s_e^2 is the unbiased estimate of σ_e^2 , given by error mean square for R.B.D.

Hence, the efficiency E' of BIBD w.r.t. R.B.D. is given by :

$$E' = \frac{s_e'^2}{r} \cdot \frac{rE}{2s_e^2} = E \left(\frac{s_e'^2}{s_e^2} \right) ; E < 1 \quad \dots(6-306)$$

Thus, for a homogeneous set of material (i.e., $s_e'^2 = s_e^2$), the BIBD is less efficient than the R.B.D. If, however, by taking small block size, the reduction in variance is larger enough than the one to compensate the loss in accuracy due to the use of BIBD, the BIBD is preferred. If the reduction in variance is just enough (i.e., $E \cdot \frac{s_e'^2}{s_e^2} \approx 1$), the R.B.D. should be preferred.

Example 6-21. The data in Table 6-78 gives the results of an experiment for comparing 7 treatments in 7 blocks of 3 units each, there thus being 3 replications of each treatment. Analyse the data.

Treatments	Blocks						
	1	2	3	4	5	6	7
1	50	42	91	—	—	—	—
2	—	—	118	94	94	—	—
3	76	—	—	64	—	80	—
4	—	—	72	—	—	53	31
5	44	—	—	—	65	—	54
6	—	102	—	—	119	92	—
7	—	38	—	38	—	—	37

Solution. The above design is a BIBD with parameters : $v = 7, b = 7, r = 3 = k, \lambda = 1$

To carry out the analysis of the design we compute the following quantities :

T_i = Total yield for the i th treatment from all the blocks, ($i = 1, 2, \dots, 7$)

B_j = Total yield from the j th block, ($j = 1, 2, \dots, 7$)

$\sum_j n_{ij} B_j$ = Total yield from all the blocks in which i th treatment occurs [($i = 1, 2, \dots, 7$)]

For example,

$\sum n_{1j} B_j$ = Total yield from all the blocks in which 1st treatment occurs.

$$= B_1 + B_2 + B_3 = 170 + 182 + 281 = 633$$

$$\sum n_{3j} B_j = B_1 + B_4 + B_6 = 170 + 196 + 225 = 591 ; \text{ and so on}$$

$$Q_i = T_i - \sum_{j=1}^7 n_{ij} B_j / k ; (i = 1, 2, \dots, 7)$$

TABLE 6.78A : CALCULATIONS FOR VARIOUS S.S.

	Treatment Totals' T_i	B_j	B_j^2	$\sum_j n_{ij} B_j$	$\sum_j n_{ij} B_j/k$	Q_i	Q_i^2	T_i^2
(1)	(2)	(3)	(4)	(5)	(6)	(7) = (2) - (6)	(8)	(9)
1	183	170	28,900	633	211.00	-28.00	781.00	33,489
2	306	182	33,124	755	251.67	54.33	2,951.75	93,636
3	220	281	78,961	591	197.00	23.00	529.00	48,400
4	156	196	38,416	628	209.33	-53.33	2,844.09	24,336
5	163	278	77,284	570	190.00	-27.00	729.00	26,569
6	313	225	50,625	685	228.33	84.67	7,169.00	97,969
7	113	122	14,844	500	166.67	-53.67	2,880.47	12,769
Total	1,454	1,454	3,22,194				17,887.31	3,37,168

If y_{ij} denotes the yield from the experimental unit receiving the i th treatment in the j th block,

$$G = \sum_{(i,j) \in D} y_{ij} = 1,454 \quad ; \quad C.F. = \frac{G^2}{21} = \frac{(1,454)^2}{21} = \frac{21,14,116}{21} = 1,00,672.19$$

$$\text{Raw S.S.} = \sum_{(i,j) \in D} y_{ij}^2 = 1,15,730 \quad ; \quad \text{Total S.S.} = \text{Raw S.S.} - C.F. = 15,057.81$$

$$\text{Blocks (unadjusted) S.S.} = \frac{1}{3} \sum_{j=1}^7 B_j^2 - C.F. = 1,07,398.00 - 1,00,672.19 = 6,725.81$$

$$\text{Treatments (adjusted) S.S.} = \sum_{i=1}^7 Q_i^2 / rE_i \quad \quad \quad [(c.f. (6.299))]$$

where $E = \frac{\lambda v}{kr} \Rightarrow rE = \frac{\lambda v}{k} = \frac{7}{3}$

$$\text{Treatments (adjusted) S.S.} = \frac{17,887.31 \times 3}{7} = 7,665.99$$

$$\text{Error (Intrablock) S.S.} = \text{Total S.S.} - \text{Blocks (unadjusted) S.S.} - \text{Treatments (adjusted) S.S.}$$

$$= 15,057.81 - 6,725.81 - 7,665.99 = 666.01$$

TABLE 6.79 : ANOVA TABLE FOR BIBD (INTRA-BLOCK ANALYSIS)

Source of Variation	S.S.	d.f.	Mean S.S.	Variance Ratio
Blocks (Unadjusted)	6,725.81	6	1,120.97	$\frac{1,120.97}{83.25} = 13.46$
Treatments (Adjusted)	7,665.99	6	1,277.67	$\frac{1,277.67}{83.25} = 15.347^*$
Error	666.01	8	83.25	—
Total	15,057.80	20		

Since tabulated $F_{6,8} (0.05) = 3.58$, the variance ratio due to treatments is highly significant and hence the null hypothesis

$$H_0 : t_1 = t_2 = \dots = t_7$$

i.e., the treatment effects are equal is rejected at 5% level of significance. Hence, we conclude that the treatments differ significantly from each other.

To find out which pair of treatments differ significantly we compute the critical difference between the estimates of $t_i - y_p$, i.e., between $\tau_i - \tau_p$.

$$C.D. \text{ for } (\tau_i - \tau_p) = t_{error \text{ or } d.f.} (0.025) S.E. (\tau_i - \tau_p) = t_8 (0.025) \cdot \sqrt{\frac{2s^2}{rE}}$$

$$= 2.306 \times \sqrt{\frac{2 \times 3 \times 83.25}{7}} = 2.306 \times 8.447 = 19.4787 \quad \dots(*)$$

Adjusted treatment means, say \bar{T}_i , ($i = 1, 2, \dots, 7$) are given by

$$\bar{T}_i = m + \tau_i = m + \frac{k}{\lambda v} Q_i = m + \frac{3}{7} Q_i \quad [c.f. (6-297)] \quad \dots(**)$$

where
$$m = \frac{G}{bk} = \frac{1,454}{21} = 69.24$$

TABLE 6-80 : ADJUSTED TREATMENTS

	Q_i	$\bar{T}_i = 69.24 + \frac{3}{7} Q_i$
1	-28.00	57.24
2	54.33	92.53
3	23.00	79.10
4	-53.33	46.38
5	-27.00	57.67
6	84.67	105.52
7	-53.67	46.24

$\therefore \bar{T}_i = 69.24 + \frac{3}{7} Q_i$, which are calculated in Table 6-80.

$$C.D. \text{ for } (\bar{T}_i - \bar{T}_p) = t_8 (0.025) S.E. (\bar{T}_i - \bar{T}_p)$$

$$= t_8 (0.025) S.E. [(m + \tau_i) - (m + \tau_p)]$$

$$= t_8 (0.025) S.E. (\tau_i - \tau_p) = 19.4784 \quad [\text{From } (*)]$$

Treatments (adjusted) in descending order are given in Table 6-81.

TABLE 6-81 : ADJUSTED TREATMENTS

		$ \Delta $
\bar{T}_6	105.52	12.99
\bar{T}_2	92.53	13.43
\bar{T}_3	79.10	21.43
\bar{T}_5	57.67	0.43
\bar{T}_1	57.24	0.86
\bar{T}_4	46.38	0.14
\bar{T}_7	46.24	

Conclusions. (i) Treatment 6 differs significantly from all the treatments except treatment 2.

(ii) Treatment 2 differs significantly from all the treatments except 6th and 3rd.

(iii) Treatment 3 differs significantly from all the treatments except treatment 2.

(iv) Treatment 5 differs significantly from all the treatments except 1st, 4th and 7th.

(v) Treatment 1 differs significantly from all the treatments except 4th, 5th and 7th.

(vi) Treatment 4 differs significantly from all the treatments except 4th, 5th and 7th.

(vii) Treatment 7 differs significantly from all the treatments except 1st, 4th and 5th.

(viii) Treatments 1, 4, 5, 7 do not differ significantly among themselves. Treatment 2 and 6 are also homogeneous.

In order to test hypothesis H_b about the homogeneity of blocks, i.e.,

$$H_b : b_1 = b_2 = \dots = b_7$$

i.e., all the block effects are equal, we have to calculate the (adjusted) block S.S.

$$\text{(Unadjusted) Treatment S.S.} = \frac{1}{3} \sum_{i=1}^7 T_i^2 - \text{C.F.} = \frac{1}{3}(3,37,168) - 10,06,772 \cdot 19 = 11,717 \cdot 14$$

Hence, on using (6.301),

$$\text{(Adjusted) Block S.S.} = \text{Block S.S. (unadjusted)} + \text{Treatment S.S. (adjusted)}$$

$$-\text{Treatment S.S. (unadjusted)} = 6,725 \cdot 81 + 7,665 \cdot 99 - 11,717 \cdot 14 = 2,674 \cdot 66$$

In order to test $H_b : b_1 = b_2 = \dots = b_7$, we will compare (adjusted) Block mean S.S. with error mean S.S. and hence the test statistic is given by

$$F = \frac{(2,674 \cdot 66)/6}{83 \cdot 25} = 5 \cdot 355 \sim F(6, 8)$$

Since $F \geq 3 \cdot 58$, H_b is refuted or rejected

Remark. In order to test the significance of the different treatment pairs, we need not calculate the adjusted treatment means \bar{T}_i , ($i = 1, 2, \dots, 7$). The test can as well be carried out with τ_i 's, using the test statistic (6.304.)

DISCUSSION AND REVIEW QUESTIONS

1. (a) What do you understand by "Design of an Experiment". Explain with examples the need of a design of experiment to draw sound conclusion.
(b) Explain the purpose of design of experiments, and indicate the characteristics of a good experimental design.

2. (a) Describe the following three fundamental principles :

(i) randomisation, (ii) replication, and (iii) local control.

Explain the importance of these principles with respect to designing statistical experiments. Give suitable illustrations.

(b) Explain, with illustrations, the different methods of controlling heterogeneity in field experiments.

(c) Discuss how the efficiency of an experiment can be increased by increased replication and local control.

(d) Define 'experimental error'. What are its main sources? What methods are adopted to increase the accuracy of an experiment?

3. (a) Write critical notes on replication and local control, as means of increasing the efficiency of a design.

(b) Show, by suitable examples, how randomisation enables us to apply statistical methods.

COMPLETELY RANDOMISE DESIGN AND RANDOMISED BLOCK DESIGN

4. (a) Give the layout of a completely randomised block design and explain the situations when it is used. Discuss its merits and demerits.

(b) Give the complete statistical analysis of C.R.D.

5. (a) What is meant by a Randomised Block Design?

(b) Give the analysis of variance for the design, stating clearly the mathematical model and the underlying assumptions.

Obtain the efficiency of this design compared to completely randomised design.

Source :

1. S.C. Gupta and V.K. Kapoor : Fundamental of Applied Statistics – Sultan Chand & Sons, Fourth Edition, 2015.
2. Panneer Selvam: Design And Analysis of Experiments, Prentice Hall.