

# Partially Balanced Resolution $\mathbb{N}^*$ Designs in a $2^m$ -Factorial

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## ABSTRACT

Srivastava and Anderson (1970) illustrate a method of obtaining Balanced (but not orthogonal) Resolution  $\mathbb{N}^*$  designs starting with a BIB design. The incidence matrix of a BIB design with parameters  $(v, b, r, k, \text{ and } \lambda)$  is utilized to obtain Balanced Resolution  $\mathbb{N}^*$  designs with  $m$  factors and  $n=2b$  runs, where  $m \leq v$ . In this paper, the same idea is extended to the case of PBIB designs to obtain Partially Balanced Resolution  $\mathbb{N}^*$  designs. In the designs obtained here the variances are balanced and the covariances are partially balanced with respect to the main effects. A proof of this property of Partially Balanced Resolution  $\mathbb{N}^*$  designs is given. The efficiency of Partially Balanced Resolution  $\mathbb{N}^*$  designs is also considered and examples of Partially Balanced Resolution  $\mathbb{N}^*$  designs are included.

## 1. Introduction

Consider  $2^n$ -fractional factorial designs of Resolution  $\mathbb{N}$ , where the three and higher order interactions are assumed negligible and the main effects are parameters of interest. The mean and the two-factor interactions are not assumed to be negligible and they are not of interest with regard to estimation. A subclass of the above is known as Resolution  $\mathbb{N}^*$  designs if the main effects are estimable orthogonal to the general mean and two-factor interactions.

Using Box-Wilson fold-over principle, we can generate a Resolution  $\mathbb{N}^*$  design from a Resolution  $\mathbb{III}$  design. It is well known that Resolution  $\mathbb{III}$  designs for  $4\alpha-1$  factors in  $4\alpha$  runs can be obtained using Hadamard matrices. To each run then we adjoin a factor at level 1 and folding over the plan thus obtained we get a Resolution  $\mathbb{N}^*$  design for  $4\alpha$

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$(=m)$  factors in  $8\alpha(=n)$  runs. These designs are orthogonal and balanced. However, orthogonal designs, while a subclass of the class of balanced designs, are not available for every value of  $n$ .

Srivastava and Anderson (1970) illustrated a method of obtaining Balanced (but not orthogonal) Resolution  $IV^*$  designs with  $m$  factors and  $n=2b$  runs starting with a BIB design with parameters  $(v, b, r, k, \text{ and } \lambda)$ , where  $m \leq v$ .

The same idea could be extended to the case of PBIB designs to obtain Partially Balanced Resolution  $IV^*$  designs. In this paper, we deal with Partially Balanced Resolution  $IV^*$  designs. In Section 2, we illustrate a method of a Partially Balanced Resolution  $IV^*$  design using a PBIB design and the property of this design is investigated. In Section 3, we consider efficiency of Partially Balanced Resolution  $IV^*$  designs. Examples of Partially Balanced Resolution  $IV^*$  designs with  $m$  factors in  $2m$  runs are presented in Section 4. A result in Section 2 is proved in Appendix.

## 2. Partially Balanced Resolution $IV^*$ design in a $2^n$ -factorial

We can illustrate a method of obtaining a Partially Balanced Resolution  $IV^*$  design  $m$  factors in  $n=2b$  runs using a PBIB design with parameters  $(v=m, b, r, k; \lambda_1, \lambda_2, \dots, \lambda_t)$ . Write down the incidence matrix  $N(\text{m} \times \text{b})$ , and we get a design  $T$  given by

$$(2.1) \quad T = [N : \bar{N}],$$

where  $\bar{N}$  denotes the matrix by interchanging the symbols 0 and 1 in  $N$ . This is the well known Box-Wilson fold-over principle, so the design  $T$  is a Resolution  $IV^*$  design in a  $2^n$ -factorial.

Webb(1968) showed that for any Resolution  $IV$  design, we have  $n \geq 2m$ ; Margolin(1969) independently proved Webb's result, and also showed that if  $n=2m$ , and if every factor occurs at level 0 as often as at level 1, then a Resolution  $IV$  design is a necessarily a fold-over design. Hence minimal Resolution  $IV^*$  designs are fold-over types.

The coefficient matrix for the fold-over design (2.1) will be of the form

$$(2.2) \quad X = \begin{bmatrix} \mathbf{1}_b & X_1 & X_2 \\ \mathbf{1}_b & -X_1 & X_2 \end{bmatrix}$$

where  $\mathbf{1}_b$  is the  $b \times 1$  vector of 1's,  $[X_1' : -X_1']'$  and  $[X_2' : X_2']'$  are the coefficient matrices corresponding to the main effects and the two-factor interactions, respectively.

From(2.2)

$$(2.3) \quad X'X = \begin{bmatrix} 2b & 0 & 2\mathbf{1}'X_2 \\ 0 & 2X_1'X_1 & 0 \\ 2X_2'\mathbf{1} & 0 & 2X_2'X_2 \end{bmatrix}$$

This means that the main effects are estimable orthogonal to the general mean and two-factor interactions, if  $X_1'X_1$  is nonsingular.

We know that, from the incidence matrix  $N$ ,

$$(2.4) \quad X_1 = 2N' - J_{bm}'$$

where  $J_{bm}$  is the  $b \times m$  matrix each of whose element is unity, so

$$(2.5) \quad 2X_1'X_1 = 8NN' + (2b - 8r)J_{mm}.$$

Suppose that the matrix  $X_1'X_1$  is nonsingular, then the variance-covariance matrix of the estimators of the main effects under  $T = [N; \bar{N}]$  is

$$(2.6) \quad V_T = \frac{1}{2}(X_1'X_1)^{-1}\sigma^2 = \frac{1}{2}(4NN' + (b - 4r)J_{mm})^{-1}\sigma^2,$$

Now, let  $B_i$ ,  $i=1, 2, \dots, t$ , be the association matrices of the PBIB design in (2.1), then it is well known that

$$(2.7) \quad NN' = rB_0 + \lambda_1 B_1 + \dots + \lambda_t B_t,$$

where  $B_0 = I_m$  the  $m \times m$  identity matrix. So the matrix  $X_1'X_1$  in (2.6) can be expressed as follows: Since  $B_0 + B_1 + \dots + B_t = J_{mm}$ ,

$$(2.8) \quad X_1'X_1 = c_0 B_0 + c_1 B_1 + \dots + c_t B_t.$$

We will prove the following result in Appendix if the matrix  $X_1'X_1$  is nonsingular

$$(2.9) \quad (X_1'X_1)^{-1} = q_0 B_0 + q_1 B_1 + \dots + q_t B_t$$

for real numbers  $q_i$ 's.

This means that the variance-covariance matrix of the estimators of main effects under the design  $T = [N; \bar{N}]$  is

$$(2.10) \quad V_T = \frac{1}{2} \left( \sum_{i=0}^t q_i B_i \right) \sigma^2.$$

For a Partially Balanced Resolution  $\mathbb{N}^*$  design constructed from a PBIB design with above parameters, therefore, the variance of the estimate of every main effect is the same, and the covariances between two of such estimators have  $t$ -associate partially balanced scheme in the following sense:

Let  $\text{cov}(i, j)$  be a covariance between  $i$  th and  $j$  th main effects corresponding to the  $i$  th and  $j$  th treatments respectively which are the  $k$  th associates in the original PBIB design, then the value of  $\text{cov}(i, j)$  is  $\frac{1}{2} q_k \sigma^2$  and this uniquely determined only by the associates. In this sense the above design (2.1) can be called Partially Balanced Resolution

$\mathbb{N}^*$  design with parameters  $(m=v, n=2b; \mu_0, \mu_1, \dots, \mu_t)$ , where  $\mu_i = \frac{1}{2}q_i, i=0, 1, \dots, t$ .

### 3. Efficiency of Partially Balanced Resolution $\mathbb{N}^*$ design

In determining the efficiency of Balanced Resolution  $\mathbb{N}^*$  design, Srivastava and Anderson (1970) presented the following results.

Let  $B_n$  denotes the class of all balanced nonsingular fractional designs of Resolution  $\mathbb{N}^*$  with  $m$  factors and  $n$  runs. Let  $T \in B_n$ , then the lower bound of  $\text{tr}(V_T)$  is  $m/n$ , if  $n=4\alpha$ , and  $(m-1)/(n-2) + 1/(n-2+2m)$ , if  $n=4\alpha+2$ , where  $\alpha$  is an integer. Also they proved that (a)  $T$  is trace optimal in  $B_n$ , if  $T$  is determinant optimal and conversely, (b)  $T$  is maximum root optimal in  $B_n$ , if  $T$  is trace optimal and conversely.

Finally, they defined the following efficiency measure for Balanced Resolution  $\mathbb{N}^*$  designs:

$$(3.1) \quad \text{absolute (trace) efficiency} = \tau(m, n) / \text{tr}(V_T),$$

where

$$(3.2) \quad \tau(m, n) = \begin{cases} m/n, & \text{if } n=4\alpha, \\ (m-1)/(n-2) + 1/(n-2+2m), & \text{if } n=4\alpha+2 \end{cases}$$

where  $\alpha$  is an integer.

We could use the above efficiency measure (3.1) for Partially Balanced designs of Resolution  $\mathbb{N}^*$ , where, in this case,  $\tau(m, n)$  is lower bound defined above given  $m$  and  $n$  and  $V_T$  is the variance-covariance matrix of our concerning Partially Balanced design of Resolution  $\mathbb{N}^*$ . The ratio  $\tau(m, n) / \text{tr}(V_T)$  therefore measures the efficiency of our design relative to a (possibly nonexistent) optimum Balanced Resolution  $\mathbb{N}^*$  design given  $m$  and  $n$ .

### 4. Examples

We will present examples for 9 factors with 18 runs and 10 factors with 20 runs, because these designs are not appeared in Srivastava and Anderson (1970).

**Example 4.1** Consider a group divisible PBIB design with parameters  $(v=9, b=9, r=7, k=7; \lambda_1=6, \lambda_2=5)$ , whose blocks are (1, 2, 3, 4, 5, 6, 7), (2, 4, 5, 6, 7, 8, 9), (1, 2, 3, 5, 7, 8, 9), (3, 4, 5, 6, 7, 8, 9), (1, 2, 3, 4, 5, 6, 9), (1, 2, 3, 4, 5, 6, 8), (1, 2, 3, 6, 7, 8, 9), (1, 2, 3, 4, 7, 8, 9), and (1, 4, 5, 6, 7, 8, 9).

Then the incidence matrix  $N$  of the design is given by

$$N = \begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 & 1 & 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 & 1 & 1 & 1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 1 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 \end{pmatrix}$$

so

$$NN' = \begin{pmatrix} 7 & 6 & 6 & 5 & 5 & 5 & 5 & 5 & 5 & 5 \\ 6 & 7 & 6 & 5 & 5 & 5 & 5 & 5 & 5 & 5 \\ 6 & 6 & 7 & 5 & 5 & 5 & 5 & 5 & 5 & 5 \\ 5 & 5 & 5 & 7 & 6 & 6 & 5 & 5 & 5 & 5 \\ 5 & 5 & 5 & 6 & 7 & 6 & 5 & 5 & 5 & 5 \\ 5 & 5 & 5 & 6 & 6 & 7 & 5 & 5 & 5 & 5 \\ 5 & 5 & 5 & 5 & 5 & 5 & 7 & 6 & 6 & 6 \\ 5 & 5 & 5 & 5 & 5 & 5 & 6 & 7 & 6 & 6 \\ 5 & 5 & 5 & 5 & 5 & 5 & 6 & 6 & 7 & 7 \end{pmatrix} = I_3 \otimes I_3 + I_3 \otimes J_3 + 5J_3 \otimes J_3,$$

where symbol  $\otimes$  stands for the Kronecker product.

Using matrix  $N$ , we obtain the following fold-over design  $T$  of 9 factors in 18 runs:

$$T = [N : \bar{N}] = \begin{pmatrix} 1 & 0 & 1 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & 0 & 1 & 1 & 1 & 0 & 1 & 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 1 & 1 & 1 & 0 & 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \end{pmatrix}$$

Each of the 18 columns of the above array corresponds to one run or treatment combination, and each of the nine rows to one of the nine factors. Thus, in the first run, the factors 1, 2, 3, 4, 5, 6, and 7 are at level 1, and the rest at level 0. This design  $T$  is a Resolution  $\text{IV}^*$  design with 9 factors and 18 runs in a  $2^9$ -factorial since the design is a fold-over design.

In the design, from (2.5)

$$\begin{aligned} X_1'X_1 &= 4NN' - 19J_9 \\ &= 4I_3 \otimes I_3 + 4I_3 \otimes J_3 + J_3 \otimes J_3, \end{aligned}$$

so the matrix  $X_1'X_1$  has Property A defined by Kurkjian and Zelen (1963).

or

$$X_1'X_1 = 9B_0 + 5B_1 + B_2,$$

where  $B_0 = I_9$ ,  $B_1 = I_3 \otimes (J_3 - I_3)$ , and  $B_2 = (J_3 - I_3) \otimes J_3$ .

We obtain

$$\begin{aligned} (X_1'X_1)^{-1} &= \frac{1}{400} (100I_9 - 25I_3 \otimes J_3 - J_9) \\ &= \frac{1}{400} (74B_0 - 26B_1 - B_2) \end{aligned}$$

Therefore, the design  $T$  is a Partially Balanced Resolution  $\text{IV}^*$  design with parameters  $(m=9, n=18; \mu_0=37/400, \mu_1=-13/400, \mu_2=-1/800)$  in a  $2^9$ -factorial.

In the design

$$\text{tr}(V_T) = 74 \times 9/800 = 333/400$$

and therefore, since in this case  $\tau(m, n) = 9/17$ ,

$$\text{absolute (trace) efficiency} = 9 \times 400/17 \times 333 = 0.64.$$

Note: There exists BIB design with parameters  $(v=9, b=9, r=8, k=8, \lambda=7)$ , whose blocks are (1, 2, 3, 4, 5, 6, 7, 8, 9), (1, 2, 4, 5, 6, 7, 8, 9), (1, 2, 3, 5, 6, 7, 8, 9), (1, 2, 3, 4, 6, 7, 8, 9), (1, 2, 3, 4, 5, 7, 8, 9), (1, 2, 3, 4, 5, 6, 8, 9), (1, 2, 3, 4, 5, 6, 7, 9), (1, 2, 3, 4, 5, 6, 7, 8), and (2, 3, 4, 5, 6, 7, 8, 9). However, the efficiency of the Balanced Resolution  $\text{IV}^*$  design constructed from the BIB design is 0.52 which is less than the efficiency of the above Partially Balanced Resolution  $\text{IV}^*$  design.

**Example 4.2.** Consider a triangular association scheme PBIB design with parameters  $(v=10, b=10, r=4, k=4; \lambda_1=1, \lambda_2=2)$ , whose blocks are (2, 6, 7, 10), (1, 2, 5, 10), (2, 3, 7, 8), (2, 4, 6, 9), (1, 8, 9, 10), (3, 4, 5, 10), (1, 4, 7, 8), (3, 5, 7, 9), (1, 3, 6, 9), and (4, 5, 6, 8). In this design,  $n_1=6, n_2=3; p_{11}^1=3, p_{12}^1=p_{21}^1=2, p_{22}^1=1, p_{11}^2=4, p_{12}^2=p_{21}^2=2, p_{22}^2=0$ .

The association matrices are as follows

$$B_0 = I_{10}, \quad B_2 = J_{10} - B_1 - B_0,$$

where

$$B_1 = \begin{pmatrix} 0 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 1 & 0 & 0 & 1 & 1 & 0 \\ 1 & 1 & 0 & 1 & 0 & 0 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 & 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 0 & 0 & 1 & 1 & 1 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 & 1 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 & 0 & 1 & 1 & 1 & 1 & 0 \end{pmatrix}$$

In this case, we obtain the following Partially Balanced Resolution  $\mathbb{N}^*$  design of 10 factors in 20 runs in a  $2^{10}$ -factorial:

$$T = \begin{pmatrix} 0 & 1 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 1 & 1 & 1 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 1 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 & 0 & 1 & 1 & 0 & 0 & 1 & 0 & 1 & 1 & 0 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 1 & 0 & 1 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 1 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 0 & 1 & 1 & 1 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 1 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{pmatrix}$$

In this design

$$\begin{aligned} X_1'X_2 &= 4NN' - 6J_{10} \\ &= 10B_0 - 2B_1 + 2B_2. \end{aligned}$$

Therefore, from (a.14) in Appendix

$$(10\mathcal{P}_0' - 2\mathcal{P}_1' + 2\mathcal{P}_2') \begin{pmatrix} q_0 \\ q_1 \\ q_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

where

$$\mathcal{P}_0' = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \mathcal{P}_1' = \begin{pmatrix} 0 & 6 & 0 \\ 1 & 3 & 2 \\ 0 & 4 & 2 \end{pmatrix}, \quad \mathcal{P}_2' = \begin{pmatrix} 0 & 0 & 3 \\ 0 & 2 & 1 \\ 1 & 2 & 0 \end{pmatrix}.$$

We obtain the following solution:

$$q_0 = 5/32, \quad q_1 = 1/32, \quad q_2 = -1/32.$$

Hence

$$(X_1'X_1)^{-1} = \frac{1}{32}(5B_0 + B_1 - B_2).$$

Therefore, the design  $T$  is a Partially Balanced Resolution IV\* design with parameters ( $m=10$ ,  $n=20$ ;  $\mu_0=5/64$ ,  $\mu_1=1/64$ ,  $\mu_2=-1/64$ ).

In the design,

$$tr(V_T) = 10\left(\frac{1}{2}q_0\right) = 50/64,$$

while,  $\tau(m, n) = 10/20$ .

So,

$$\text{absolute (trace) efficiency} = 10 \times 54/20 \times 50 = 0.64.$$

### Appendix

Consider a PBIB design with parameters  $(v, b, r, k; \lambda_1, \lambda_2, \dots, \lambda_t)$ , and let  $n_i$ ,  $i=1, 2, \dots, t$ , and  $p_i^j$ ,  $i, j, k=1, 2, \dots, t$ , be the parameters of its association scheme.

Let

$$(a.1) \quad B_i = \begin{pmatrix} b_{11}^i & b_{12}^i & \dots & b_{1v}^i \\ b_{21}^i & b_{22}^i & \dots & b_{2v}^i \\ & & \dots & \\ b_{v1}^i & b_{v2}^i & \dots & b_{vv}^i \end{pmatrix}, \quad i=1, 2, \dots, t,$$

be the association matrices, then we observe that  $B_i$ ,  $i=1, 2, \dots, t$ , are symmetrical matrices, with row and column totals equal to  $n_i$ . In addition, let every treatment be the 0 th associate of itself and no other treatments. Then we see that

$$(a.2) \quad B_0 = I_v, \quad n_0 = 1, \quad p_{ij}^0 = \begin{cases} n_i, & \text{if } i=j, \\ 0, & \text{if } i \neq j, \end{cases} \quad p_{0k}^i = \begin{cases} 1, & \text{if } i=k, \\ 0, & \text{if } i \neq k, \end{cases} \quad \text{and } \lambda_0 = r.$$

Let  $N(v \times b)$  be the incidence matrix of the above PBIB design, then it is well known that

$$(a.3) \quad NN' = rB_0 + \lambda_1 B_1 + \dots + \lambda_t B_t.$$

Given two treatments  $\alpha$  and  $\beta$ , they are either 0 th, 1st, ..., or  $t$  th associates, and



hence only one of the elements  $b_{\alpha\beta}^0, b_{\alpha\beta}^1, \dots, b_{\alpha\beta}^t$  is unity. Hence

$$(a.4) \quad B_0 + B_1 + \dots + B_t = J_v.$$

From the same consideration, it follows that

$$c_0 B_0 + c_1 B_1 + \dots + c_t B_t = 0$$

holds if and only if

$$c_0 = c_1 = \dots = c_t = 0;$$

hence the linear function of  $B_0, B_1, \dots, B_t$  form a vector space of  $(t+1)$ - dimensionality with basis  $B_0, B_1, \dots, B_t$ .

Since the  $(\alpha, \beta)$  th element of  $B_j B_k$  can be interpreted as the number of treatments common to the  $j$  th associates of  $\alpha$  and the  $k$  th associates of  $\beta$ , we have

$$(a.5) \quad B_j B_k = \sum_{i=0}^t p_{jk}^i B_i, \quad j, k=0, 1, \dots, t,$$

and hence the multiplication is closed in the set of linear function of  $B_0, B_1, \dots, B_t$ .

Clearly this set of linear functions of  $B_0, B_1, \dots, B_t$  forms a commutative group.

Since matrix multiplication is associative

$$(a.6) \quad B_i (B_j B_k) = \sum_u \sum_j p_{jk}^u p_{iu}^i B_i \\ = (B_i B_j) B_k = \sum_u \sum_j p_{jk}^u p_{iu}^i B_u.$$

The independence of  $B_0, B_1, \dots, B_t$  implies that

$$(a.7) \quad \sum_u p_{jk}^u p_{iu}^r = \sum_u p_{ij}^u p_{uk}^r \text{ for } r=0, 1, \dots, t.$$

Now let us define  $\mathcal{P}_i$ -matrices by

$$(a.8) \quad \mathcal{P}_i = \begin{pmatrix} p_{0i}^0 & p_{0i}^1 & \dots & p_{0i}^t \\ p_{1i}^0 & p_{1i}^1 & \dots & p_{1i}^t \\ \dots & \dots & \dots & \dots \\ p_{ti}^0 & p_{ti}^1 & \dots & p_{ti}^t \end{pmatrix}, \quad i=0, 1, \dots, t.$$

In terms of  $\mathcal{P}_i$ -matrices, (a.7) implies that

$$(a.9) \quad \mathcal{P}_j \mathcal{P}_k = \sum_{i=0}^t p_{jk}^i \mathcal{P}_i$$

Thus the  $\mathcal{P}$ -matrices multiply in the same manner as the  $B$ -matrices. It can be easily verified that  $\mathcal{P}_0, \mathcal{P}_1, \dots, \mathcal{P}_t$  are linearly independent and they form the basis for a vector space and combine in the same way as the  $B$ -matrices as well as in multiplication.

Bose and Mesner (1959) proved the following Theorem a.1.

**Theorem a.1.** The distinct characteristic roots of

$$B = c_0 B_0 + c_1 B_1 + \dots + c_t B_t$$

and the distinct characteristic roots of

$$\mathcal{P} = c_0 \mathcal{P}_0 + c_1 \mathcal{P}_1 + \cdots + c_t \mathcal{P}_t$$

are the same.

Proof: Let  $f(\lambda)$  and  $g(\lambda)$  be the minimal polynomial of  $B$  and  $\mathcal{P}$ , respectively. They are monic polynomials. The term  $f(B)$  can be expressed as

$$f(B) = \sum_{i=0}^t d_i B_i,$$

whose representation in  $(t+1) \times (t+1)$  matrices will be

$$f(\mathcal{P}) = \sum_{i=0}^t d_i \mathcal{P}_i.$$

Since  $f(\lambda)$  is the minimal polynomial of  $B$ , we have  $f(B) = 0$ , which implies that  $d_0 = d_1 = \cdots = d_t = 0$  and hence  $f(\mathcal{P}) = 0$ . Since  $f(\mathcal{P}) = 0$  and since  $g(\lambda)$  is the minimal polynomial of  $\mathcal{P}$ , we have the result that  $g(\lambda)$  divides  $f(\lambda)$ . We can similarly show that  $f(\lambda)$  divides  $g(\lambda)$ . Since both  $f(\lambda)$  and  $g(\lambda)$  are monic polynomials,  $f(\lambda) = g(\lambda)$ . Thus  $B$  and  $\mathcal{P}$  have the same distinct characteristic roots.

Now we can prove the following Theorem a.2:

**Theorem a.2.** Let  $(\alpha NN' + \beta J_v)$  be a nonsingular matrix, then the inverse of the matrix has the form:

$$(a.10) \quad q_0 B_0 + q_1 B_1 + \cdots + q_t B_t$$

for real numbers  $q_i$ 's.

Proof: Since  $B_0 + B_1 + \cdots + B_t = J_v$ ,

$$(a.11) \quad \alpha NN' + \beta J_v = c_0 B_0 + c_1 B_1 + c_2 B_2 + \cdots + c_t B_t,$$

where  $c_0 = \alpha r + \beta$ , and  $c_i = \alpha \lambda_i + \beta$  for  $i = 1, 2, \dots, t$ .

Suppose that the inverse of  $(\alpha NN' + \beta J_v)$  has the following form

$$q_0 B_0 + q_1 B_1 + \cdots + q_t B_t,$$

then

$$\begin{aligned} I_v = B_0 &= \left( \sum_{i=0}^t q_i B_i \right) \left( \sum_{i=0}^t c_i B_i \right) \\ &= \sum_j \sum_k q_j c_k B_j B_k \\ &= \sum_j \sum_k q_j c_k \left( \sum_i p_{jk}^i B_i \right) \\ &= \sum_j \sum_k q_j c_k p_{jk}^0 B_0 + \sum_j \sum_k q_j c_k p_{jk}^1 B_1 + \cdots + \sum_j \sum_k q_j c_k p_{jk}^t B_t. \end{aligned}$$

Since  $B_i$ 's are independent,

$$\sum_j \sum_k q_j c_k p_{jk}^0 = 1$$

$$\begin{aligned}\sum_j \sum_k q_j c_k p_{j,k}^1 &= 0 \\ \dots \\ \sum_j \sum_k q_j c_k p_{j,k}^t &= 0,\end{aligned}$$

i.e.,

$$(a.13) \quad \left( c_0 \begin{bmatrix} p_{00}^0 & p_{10}^0 \cdots p_{i0}^0 \\ p_{00}^1 & p_{10}^1 \cdots p_{i0}^1 \\ \dots & \dots \\ p_{00}^t & p_{10}^t \cdots p_{i0}^t \end{bmatrix} + c_1 \begin{bmatrix} p_{01}^0 & p_{11}^0 \cdots p_{i1}^0 \\ p_{01}^1 & p_{11}^1 \cdots p_{i1}^1 \\ \dots & \dots \\ p_{01}^t & p_{11}^t \cdots p_{i1}^t \end{bmatrix} + \dots + c_t \begin{bmatrix} p_{0t}^0 & p_{1t}^0 \cdots p_{it}^0 \\ p_{0t}^1 & p_{1t}^1 \cdots p_{it}^1 \\ \dots & \dots \\ p_{0t}^t & p_{1t}^t \cdots p_{it}^t \end{bmatrix} \right) \begin{pmatrix} q_0 \\ q_1 \\ \vdots \\ q_t \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

Hence, the equation (a.13) can be written as follows:

$$(a.14) \quad [c_0 \mathcal{P}_0' + c_1 \mathcal{P}_1' + \dots + c_t \mathcal{P}_t'] \mathbf{q} = \mathbf{d},$$

where  $\mathbf{q}' = (q_1, q_1, \dots, q_t)$  and  $\mathbf{d}' = (1, 0, \dots, 0)$ .

From the Theorem a.1, since  $\sum_i c_i B_i$  is symmetric nonsingular matrix, the characteristic roots of the matrix  $\sum_i c_i \mathcal{P}_i'$  are all real and nonzero. Therefore, the equation (a.14) has the unique solution. This means that there exist the real numbers  $q_0, q_1, \dots, q_t$ , such that

$$(\alpha NN' + \beta J_v)^{-1} = q_0 B_0 + q_1 B_1 + \dots + q_t B_t.$$

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