

## Optimality and Efficiency of Circular Neighbor Balanced Design for Second Order Circular Auto Regressive Process

Jeevitha.M<sup>1</sup> and Dr.C.Santharam<sup>2</sup>

<sup>1</sup>Student, Department of Statistics, Loyola College,

<sup>2</sup>Faculty, Department of Statistics, Loyola College,

**Abstract:** This research paper deals with the optimality of Circular Neighbor Balanced Designs for total effects when the observation errors are correlated according to second order circular stationary autoregressive process. Few results pertaining to the optimality conditions under some specified conditions are provided and the efficiencies of circular neighbor balanced designs relative to the optimal continuous block designs are also investigated. The efficiency of the Circular Neighbor Balanced Designs is illustrated corresponding to the optimal continuous block designs.

**Key Words:** Auto regressive process, second order Block Design, Circular, Correlated observation, Total effect, Universal optimality.

### I. Introduction:

In many practical problems it is inevitable that a particular plot is being affected by neighboring effects. Even though it is harm in many cases, the plot is being gained by the neighboring effects in few cases. So it was necessary for the researchers to study the neighboring effects. Under the linear models with the neighbor effects, many optimality results of block designs are established for treatment and neighbor effects separately. Hedayat and Afsarinejad (1978), Cheng and Wu (1980), Kunert (1984b) and Kushner (1997) for cross-over designs, Kunert (1984a) and Aza's, Bailey and Monod(1993), Druilhet (1999). After studying the characters of neighboring effects, it was reasonable to make the assumptions on the dependency of the observations, because practically speaking, in many of the experiments, the observations are dependent on each other, if not overall within block at least. Hence the researches have invoked their thoughts to the models where the observations are dependent.

The optimal designs or highly efficient experiments (when the observations are dependent) have been studied by many authors. H.B.Kushner (1997) derived the necessary and sufficient condition for the universal optimality in the case repeated measurement designs. Kunert and Martin (2000b) have generalized the Kushner's condition by demonstrating the method of deriving the optimal designs in the case of two dimensional neighboring models. Filipiak and Markiewicz (2005) were dealt with circular neighbor- balanced designs.

It is very important to determine which treatment combination in a block will be optimal for the better result. Hence many authors have stepped into the next level of finding out the optimal continuous sequences. For example see Kunert and Martin (2000), Filipiak and Markiewicz (2005) and Ai, He and Yu (2009). Ai, Yu and He (2009) have discussed the optimality and efficiency of one dimensional and two dimensional neighboring designs when the errors are correlated according to first order circular auto regressive process.

In this paper we study the universal optimality of circular neighbor-balanced designs for total effects, but when the observation errors are correlated according to a second-order circular autoregressive process.

In this paper, Section 2 deals with some definitions and preliminaries. Section 3 presents the main results that circular neighbor- balanced designs are universally optimal under some conditions for the total effects in linear models when the observation errors are correlated according to a second-order circular autoregressive process. In order to discuss the efficiency of circular neighbor-balanced designs among all possible block designs with the same parameters, the optimal continuous block designs are characterized in Section 4. Section 5 presents the efficiency of circular neighbor-balanced designs with blocks of small size based on the previous structure of optimal equivalence classes of sequences.

### II. Model And Definition

Consider a set of circular block designs  $\Omega_{(t,b,k)}$ . For a design  $d \in \Omega_{(t,b,k)}$ , the linear effect additive model with the left and two sided neighbor effects can be written in the vector form as

$$(M_1) \Rightarrow Y = 1_{bk} \mu + T_d \tau + L_d \lambda + (I_b \otimes I_k) \beta + \varepsilon \quad \text{--- (1)}$$

$$(M_2) \Rightarrow Y = 1_{bk} \mu + T_d \tau + L_d \lambda + R_d \rho + (I_b \otimes I_k) \beta + \varepsilon \quad \text{--- (2)}$$

Where  $Y = (Y_{11}, \dots, Y_{1k}, \dots, Y_{b1}, \dots, Y_{bk})'$ ,  $Y_{ij}$  is the observation response on plot  $j$  of block  $i$ ,  $\mu$  is the general mean,  $\tau$ ,  $\lambda$  and  $\rho$  are, respectively, the  $t$ -dimensional vectors of the direct effects, left-Neighbor effects and right-Neighbor effects of the  $t$  treatments,  $T_d$ ,  $L_d$  and  $R_d$  are the corresponding incidence matrices,  $\beta$  is the  $b$ -dimensional vector of the block effects, and  $\varepsilon$  is the vector of random errors and  $1_n$  denote a  $n$ -dimensional vector of ones and the symbol  $\otimes$  denote the Kro- Necker product.

**Information Matrix, as an inverse of variance - co variance matrix:**

As like many cases of design of experiments, the amount of information obtained from the experiment is measured in terms of information matrix. Also we know that, the information matrix can also be viewed as the inverse of variance co- variance matrix. Hence in this research work, we consider the inverse of Variance Co-Variance matrix.

For any  $m \times n$  matrix  $A$ , we define  $Q_A = I_m - A(A'A)^{-1}A'$ , where  $(A'A)^{-1}$  denotes the generalized inverse of  $(A'A)$ . Then from Kunert and Martin (2000a) the information matrix of  $d$  for estimating  $\tau$  in the model (1) under normality is,

$$C_d = T_d' (I_b \otimes S)^{-1/2} Q_{(I_b \otimes S)^{-1/2} (L_d : (I_b \otimes 1_k))} (I_b \otimes S)^{-1/2} T_d$$

Where  $(I_b \otimes S)^{-1/2}$  is an  $b \times bk$  matrix with the property  $(I_b \otimes S)^{-1/2} (I_b \otimes S)^{-1/2} = (I_b \otimes S)^{-1}$

**Definition:**

A block design is said to be a circular block design neighbor-balanced at distance  $i \leq k-1$  if it is a circular binary block design in  $\Omega_{(t,b,k)}$ , and is a BIBD such that for each ordered pair of distinct treatments, there exist exactly  $m$  plots such that each of these plots receives the first chosen treatment and the right-neighbor of it at distance  $i$  receives the second treatment. A circular block design is said to be neighbor-balanced at distances up to  $\gamma$ , abbreviated by CNBD( $\gamma$ ), if it is neighborbalanced at distance  $i$  for all  $1 \leq i \leq \gamma$ .

Here assume that the errors in each block are correlated according to a second-order circular autoregressive process, denoted by AR(2,C), as in the case of Kunert and Martin (1987), Richard Cutler (1993) for first order and Martin.O.Grodona (1989) for second order. The AR (2, C) process can be represented in the recursive form  $\varepsilon_i = \rho_1 \varepsilon_{i-1} + \rho_2 \varepsilon_{i-2} + \eta_i$  with  $|\rho_i| < 1$ ,  $i=1,2$ . where the  $\eta_i$ 's are uncorrelated noises with  $E(\eta_i)=0$  and  $Var(\eta_i)=\sigma^2$ , and  $E(\varepsilon_0)=0$ . Then  $E(\varepsilon) = 0$   $Cov(\varepsilon) = \sigma^2 I_b \otimes S$  and The covariance function of a second order autoregressive process satisfies the difference equation (Fuller, 1976 p.53)

$$\sigma_\omega(h) - \rho_1 \sigma_\omega(h-1) - \rho_2 \sigma_\omega(h-2) = 0; h > 0$$

$$\sigma_\omega(h) - \rho_1 \sigma_\omega(h-1) - \rho_2 \sigma_\omega(h-2) = \sigma^2; h = 0$$

Where,  $\sigma_\omega(h) \equiv cov(\omega_{ij}, \omega_{i+h,j})$ , for all  $i=1,2,..$

Let  $S \equiv var(\omega_j)$  where  $\omega_j$  is the error vector from the  $j$ -th block. Then  $S^{-1}$ , the inverse of  $S$ , is given by (Wise, 1955; Siddiqui 1958, Martin D.Grona 1985)

$$\sigma^2 S^{-1} = (1 + \rho_1^2 + \rho_2^2)I_k + (\rho_1^2 + \rho_2^2)H_3 + (1 + \rho_1^2)H_4 - \rho_1(H + H') - \rho_2(H + H') + (\rho_1 + \rho_2)H_5 + \rho_2 H_6 + \frac{1}{1 - \rho_2} H_7 \quad \text{----- (3)}$$

$$\sigma^2 S^{-1} = \begin{bmatrix} 1 & a_4 & a_3 & 0 & 0 & \dots & 0 & 0 \\ a_4 & a_5 & a_2 & a_3 & 0 & \dots & 0 & 0 \\ a_3 & a_2 & a_1 & a_2 & a_3 & \dots & 0 & 0 \\ 0 & a_3 & a_2 & a_1 & a_2 & \dots & 0 & 0 \\ 0 & 0 & a_3 & a_2 & a_1 & \dots & 0 & 0 \\ 0 & 0 & 0 & a_3 & a_2 & \dots & 0 & 0 \\ \cdot & \cdot & \cdot & \cdot & \cdot & \dots & \cdot & \cdot \\ \cdot & \cdot & \cdot & \cdot & \cdot & \dots & \cdot & \cdot \\ 0 & 0 & 0 & 0 & 0 & \dots & a_5 & a_4 \\ 0 & 0 & 0 & 0 & 0 & \dots & a_4 & 1 \end{bmatrix}$$

Where,

$$\begin{aligned}
 a_1 &= 1 + \rho_1^2 + \rho_2^2 \\
 a_2 &= -\rho_1(1 - \rho_2) \\
 a_3 &= -\rho_2 \\
 a_4 &= -\rho_1 \\
 a_5 &= 1 + \rho_1^2
 \end{aligned}$$

Where  $H$  denotes the  $k \times k$  matrix with  $h_{1k} = 1$  and the  $(i, j)$ <sup>th</sup> element  $h_{ij} = 1$  if  $i - j = 1$  and 0 otherwise, and

$$H_3 = \begin{bmatrix} 1 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 1 \end{bmatrix}$$

$$H_4 = \begin{bmatrix} 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 1 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & 0 & \dots & 1 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \end{bmatrix}$$

$$H_5 = \begin{bmatrix} 0 & 0 & 0 & 0 & \dots & 0 & 1 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 1 & 0 & 0 & 0 & \dots & 0 & 0 \end{bmatrix}$$

$$H_6 = \begin{bmatrix} 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 1 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 1 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \end{bmatrix}$$

$$H_7 = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots & 0 & 0 \\ 1 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 1 & 0 & 0 & \dots & 0 & 1 \\ 0 & 0 & 0 & 0 & \dots & 1 & 0 \end{bmatrix}$$

Note that when  $\rho_1, \rho_2 = 0$ , the structure of errors is reduced to the popular i.i.d. case.

### III. Universal Optimality Of CNBD (2):

In this paper we follow the universal optimality criterion defined by Kiefer (1975).

**LEMMA 1:**

Let  $C_d[\alpha]$  be the information matrix for some effect  $\alpha$  based on a design  $d$ . Assume that a design  $d \in \Omega_{(t,b,k)}$  has its information matrix completely symmetric. Then,  $d$  is universally optimal for the effect  $\alpha$  over a class  $\Omega_{(t,b,k)}$  of designs if and only if  $tr(C_{d^*}[\alpha]) = \max_{d \in D} tr(C_d[\alpha])$

Let  $\varphi$  and  $\psi$  denote the total effects of the  $t$  treatments in the models  $(M_1)$  and  $(M_2)$ , respectively, that is  $\varphi = \tau + \lambda$  and  $\psi = \tau + \lambda + \rho$ . Thus, we can obtain the following universal optimality results of CNBD's for the total effects.

**THEOREM 1**

For  $3 \leq k \leq t$ , a CNBD (2) in  $\Omega_{(t,b,k)}$  is Universally Optimal for the total effects in the model  $(M_1)$  among all the designs with no treatment Neighbor of itself when  $0 \leq \rho < 1$ , and among all the designs with no treatment Neighbor of itself at distance 1 or 2 when  $-1 < \rho < 0$ .

**PROOF:**

We already have,

$$\begin{aligned}
 \tilde{S} &= S^{-1} - (1'_k S^{-1} 1_k)^{-1} S^{-1} 1_k 1'_k S^{-1} \\
 \tilde{S} &= S^{-1} - \frac{(1 - \rho_1 - \rho_2)^4 (1 + \rho_1^2 - 2\rho_1 - \rho_2 + \rho_1\rho_2)^4 (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2 - 2\rho_1\rho_2)}{2(k-4)(1 + \rho_1^2 + \rho_2^2) - 2(k-3)\rho_1(1 - \rho_2) - 2(k-2)\rho_2 - 4\rho_1 + 2(1 + \rho_1^2)} 1_k 1'_k
 \end{aligned}$$

But,

$$\begin{aligned}
 S^{-1} &= (1 + \rho_1^2 + \rho_2^2)I_k + (\rho_1^2 + \rho_2^2)H_3 + (1 + \rho_1^2)H_4 - \rho_1(H + H') - \rho_2(H + H') + (\rho_1 + \rho_2)H_5 \\
 &\quad + \rho_2 H_6 + \frac{1}{1 - \rho_2} H_7
 \end{aligned}$$

Hence,

$$\tilde{S} = (1 + \rho_1^2 + \rho_2^2)I_k + (\rho_1^2 + \rho_2^2)H_3 + (1 + \rho_1^2)H_4 - \rho_1(H + H') - \rho_2(H + H') + (\rho_1 + \rho_2)H_5 + \rho_2H_6 + \frac{1}{1 - \rho_2}H_7 - A1_k1_k'$$

Where,

$$A = \frac{(1 - \rho_1 - \rho_2)^4(1 + \rho_1^2 - 2\rho_1 - \rho_2 + \rho_1\rho_2)^4(1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2 - 2\rho_1\rho_2)}{2(k - 4)(1 + \rho_1^2 + \rho_2^2) - 2(k - 3)\rho_1(1 - \rho_2) - 2(k - 2)\rho_2 - 4\rho_1 + 2(1 + \rho_1^2)}$$

For a design,  $d \in \Omega_{(t,b,k)}$ , the information matrix  $C_d[\alpha]$  for the effect  $\alpha = [\tau', \lambda']'$  in the model (1) can be expressed as,

$$C_d[\alpha] = (T_d, L_d)' (I_b \otimes S^{-1/2})' pr_{(I_b \otimes S^{-1/2}I_k)}^\perp (I_b \otimes S^{-1/2})(T_d, L_d) = (C_{d_{ij}}) 1 \leq i, j \leq 2,$$

Where the submatrices  $(C_{d_{ij}}) 1 \leq i, j \leq 2$ , have the forms

$$C_{d_{11}} = T_d'(I_b \otimes \tilde{S})T_d$$

$$C_{d_{12}} = T_d'(I_b \otimes \tilde{S})L_d$$

$$C_{d_{22}} = L_d'(I_b \otimes \tilde{S})L_d$$

Since  $S$  is a cyclic matrix, so  $HSH' = H'SH = S$ . For a circular design  $d$ ,  $L_{du} = HT_{du}, 1 \leq u \leq b$ . It implies that  $C_{d_{11}} = C_{d_{22}}$ .

For a CNBD (2)  $d^*$ , we have

$$T_{d^*}'(I_b \otimes H)T_{d^*} = T_{d^*}'(I_b \otimes H')T_{d^*} = T_{d^*}'(I_b \otimes H'H')T_{d^*} = \frac{bk}{t(t-1)}(1_t1_t' - I_t)$$

Then,

$$C_{d_{11}}^* = T_{d^*}'(I_b \otimes \tilde{S})T_{d^*} = T_{d^*}'\{I_b \otimes (1 + \rho_1^2 + \rho_2^2)I_k + (\rho_1^2 + \rho_2^2)H_3 + (1 + \rho_1^2)H_4 - \rho_1(H + H') - \rho_2(H + H') + (\rho_1 + \rho_2)H_5 + \rho_2H_6 + \frac{1}{1 - \rho_2}H_7 - A1_k1_k'\}T_{d^*} = (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes I_k)T_{d^*} + (\rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H_3)T_{d^*} + (1 + \rho_1^2)T_{d^*}'(I_b \otimes H_4)T_{d^*} - \rho_1T_{d^*}'(I_b \otimes H)T_{d^*} - \rho_1T_{d^*}'(I_b \otimes H')T_{d^*} - \rho_2T_{d^*}'(I_b \otimes H)T_{d^*} - \rho_2T_{d^*}'(I_b \otimes H')T_{d^*} + (\rho_1 + \rho_2)T_{d^*}'(I_b \otimes H_5)T_{d^*} + \rho_2T_{d^*}'(I_b \otimes H_6)T_{d^*} + \frac{1}{1 - \rho_2}T_{d^*}'(I_b \otimes H_7)T_{d^*} - AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}$$

Similarly,

$$C_{d_{12}}^* = T_{d^*}'(I_b \otimes \tilde{S})L_{d^*} = T_{d^*}'(I_b \otimes \tilde{S})HT_{d^*} = (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes I_k)HT_{d^*} + (\rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H_3)HT_{d^*} + (1 + \rho_1^2)T_{d^*}'(I_b \otimes H_4)HT_{d^*} - \rho_1T_{d^*}'(I_b \otimes H)HT_{d^*} - \rho_1T_{d^*}'(I_b \otimes H')HT_{d^*} - \rho_2T_{d^*}'(I_b \otimes H)HT_{d^*} - \rho_2T_{d^*}'(I_b \otimes H')HT_{d^*} + (\rho_1 + \rho_2)T_{d^*}'(I_b \otimes H_5)HT_{d^*} + \rho_2T_{d^*}'(I_b \otimes H_6)HT_{d^*} + \frac{1}{1 - \rho_2}T_{d^*}'(I_b \otimes H_7)HT_{d^*} - AT_{d^*}'(I_b \otimes 1_k1_k')HT_{d^*}$$

$$\begin{aligned}
 &= (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H)T_{d^*}' + (\rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H_3H)T_{d^*}' + (1 + \rho_1^2)T_{d^*}'(I_b \otimes H_4H)T_{d^*}' \\
 &- \rho_1T_{d^*}'(I_b \otimes HH)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes H'H)T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes HH)T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes H'H)T_{d^*}' \\
 &+ (\rho_1 + \rho_2)T_{d^*}'(I_b \otimes H_5H)T_{d^*}' + \rho_2T_{d^*}'(I_b \otimes H_6H)T_{d^*}' + \frac{1}{1 - \rho_2}T_{d^*}'(I_b \otimes H_7H)T_{d^*}' - AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}' \\
 &= (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes HH)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes I_k)T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes HH)T_{d^*}' \\
 &- \rho_2T_{d^*}'(I_b \otimes I_k)T_{d^*}' - AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}'
 \end{aligned}$$

Hence all  $C_{d_{ij}^*}$  ( $1 \leq i, j \leq 2$ ) are completely symmetric.

Rewrite  $\phi = K'\alpha$  with  $K = 1_2 \otimes I_t$ . It is obvious that  $K'K = 2I_t$ .

Hence,

$$\begin{aligned}
 C_d[\phi] &\leq \frac{1}{4}K'C_d[\alpha]K \\
 &= \frac{1}{4}(C_{d_{11}} + C_{d_{12}} + C_{d_{21}} + C_{d_{22}}) \\
 &= \frac{1}{4}(T_d'(I_b \otimes \tilde{S})T_d + T_d'(I_b \otimes \tilde{S})L_d + L_d'(I_b \otimes \tilde{S})T_d + L_d'(I_b \otimes \tilde{S})L_d \\
 &= \frac{1}{4}\{(1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes I_k)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes H)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes H')T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes H)T_{d^*}' \\
 &- \rho_2T_{d^*}'(I_b \otimes H')T_{d^*}' - AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}' + (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes I_k)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes H)T_{d^*}' \\
 &- \rho_1T_{d^*}'(I_b \otimes H')T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes H)T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes H')T_{d^*}' - AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}' \\
 &+ (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes HH)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes I_k)T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes HH)T_{d^*}' \\
 &- \rho_2T_{d^*}'(I_b \otimes I_k)T_{d^*}' - AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}' + (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H)T_{d^*}' - \rho_1T_{d^*}'(I_b \otimes HH)T_{d^*}' \\
 &- \rho_1T_{d^*}'(I_b \otimes I_k)T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes HH)T_{d^*}' - \rho_2T_{d^*}'(I_b \otimes I_k)T_{d^*}' - AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}'\} \\
 &= \frac{1}{4}\{2(1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)T_{d^*}'(I_b \otimes I_k)T_{d^*}' - 4AT_{d^*}'(I_b \otimes 1_k1_k')T_{d^*}' \quad \text{---- (4)} \\
 &+ (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)T_{d^*}'(I_b \otimes H)T_{d^*}' - 2(\rho_1 + \rho_2)T_{d^*}'(I_b \otimes HH)T_{d^*}'\}
 \end{aligned}$$

Since  $C_{d_{ij}^*}$  ( $1 \leq i, j \leq 2$ ) are completely symmetric  $C_{d_{12}^*} = C_{d_{21}^*}$ . So  $C_{d^*}[\alpha]$  commutes with

$$pr_{(K)} = \frac{1}{2}(1_21_2' \otimes I_t).$$

Then

$$C_{d^*}[\phi] = \frac{1}{4}K'C_{d^*}[\alpha]K = \frac{1}{4}(C_{d_{11}^*} + C_{d_{12}^*} + C_{d_{21}^*} + C_{d_{22}^*})$$

and consequently  $C_{d^*}[\phi]$  is also completely symmetric. Consider now (4). When  $-1 < \rho < 0$ , for a design  $d$  in  $\Omega_{t,b,k}$  with no treatment neighbor of itself at distance 1 or 2, the traces of  $T_d'(I_b \otimes H)T_d, T_d'(I_b \otimes H')T_d, T_d'(I_b \otimes HH)T_d, T_d'(I_b \otimes H'H)T_d$  are all zero, and  $\text{tr}(T_d'(I_b \otimes I_k)T_d)$  is a constant. So  $\text{tr}\{C_d[\phi]\}$  depends only on  $\text{tr}T_d'(I_b \otimes 1_k1_k')T_d$ . Moreover, a CNBD(2) is a balanced block design, so it also minimizes  $T_d'(I_b \otimes 1_k1_k')T_d$  among all possible designs of the same size. Therefore  $\text{tr}\{C_d[\phi]\}$  attains the maximum. When  $0 \leq \rho < 1$ , the traces of both  $T_d'(I_b \otimes HH)T_d, T_d'(I_b \otimes H'H)T_d$  must be non-negative. However, for a CNBD(2)  $d^*$ , they are all zero. So for a design with no treatment neighbor of Therefore  $\text{tr}\{C_d[\phi]\}$  attains the maximum. When  $0 \leq \rho < 1$ , the traces of both  $T_d'(I_b \otimes HH)T_d, T_d'(I_b \otimes H'H)T_d$ , must be non-negative. However, for a CNBD(2)  $d^*$ , they are all zero. So for a design with no treatment neighbor of itself at

distance 1, it still holds that  $tr\{C_{d^*}[\phi]\} \geq tr\{C_d[\phi]\}$ .

Hence the theorem follows from Lemma 1.

**THEOREM**

For  $4 \leq k \leq t$ , a CNBD (3) in  $\Omega_{(t,b,k)}$  is Universally Optimal for the total effects in the model  $(M_2)$  among all the designs with no treatment Neighbor of itself when  $0 \leq \rho < 1$ , and among all the designs with no treatment Neighbor of itself at distance 1 or 2 when  $-1 < \rho < 0$ .

**PROOF:**

For a design,  $d \in \Omega_{(t,b,k)}$ , the information matrix  $C_d[\alpha]$  for the effect  $\alpha=[\tau', \lambda']'$  in the model (1) can be expressed as,

$$C_d[\alpha] = (T_d', L_d') (I_b \otimes S^{-1/2})' pr_{(I_b \otimes S^{-1/2} I_k)}^\perp (I_b \otimes S^{-1/2}) (T_d, L_d)$$

$$= (C_{d_{ij}}) 1 \leq i, j \leq 2,$$

Where the submatrices  $(C_{d_{ij}}) 1 \leq i, j \leq 2$ , have the forms

$$C_{d_{11}} = T_d' (I_b \otimes \tilde{S}) T_d$$

$$C_{d_{12}} = T_d' (I_b \otimes \tilde{S}) L_d$$

$$C_{d_{13}} = T_d' (I_b \otimes \tilde{S}) R_d$$

$$C_{d_{22}} = L_d' (I_b \otimes \tilde{S}) L_d$$

$$C_{d_{23}} = L_d' (I_b \otimes \tilde{S}) R_d$$

$$C_{d_{33}} = R_d' (I_b \otimes \tilde{S}) R_d$$

Since S is a cyclic matrix, so  $HSH' = H'SH = S$ . For a circular design d,  $L_{du} = HT_{du}, 1 \leq u \leq b$ . It implies that  $C_{d_{11}} = C_{d_{22}}$ .

For a CNBD (3)  $d^*$ , we have

$$T_{d^*}' (I_b \otimes HHH) T_{d^*} = T_{d^*}' (I_b \otimes H'H'H''') T_{d^*} = \frac{bk}{t(t-1)} (1_t 1_t' - I_t)$$

Then,

$$C_{d_{13}^*} = C_{d_{31}^*} = T_{d^*}' (I_b \otimes \tilde{S}) R_{d^*}$$

$$= T_{d^*}' \{ I_b \otimes (1 + \rho_1^2 + \rho_2^2) I_k + (\rho_1^2 + \rho_2^2) H_3 + (1 + \rho_1^2) H_4 - \rho_1 (H + H') - \rho_2 (H + H') + (\rho_1 + \rho_2) H_5$$

$$+ \rho_2 H_6 + \frac{1}{1 - \rho_2} H_7 - A 1_k 1_k' \} R_{d^*}$$

$$= (1 + \rho_1^2 + \rho_2^2) T_{d^*}' (I_b \otimes I_k) R_{d^*} + (\rho_1^2 + \rho_2^2) T_{d^*}' (I_b \otimes H_3) R_{d^*} + (1 + \rho_1^2) T_{d^*}' (I_b \otimes H_4) R_{d^*}$$

$$- \rho_1 T_{d^*}' (I_b \otimes H) R_{d^*} - \rho_1 T_{d^*}' (I_b \otimes H') R_{d^*} - \rho_2 T_{d^*}' (I_b \otimes H) R_{d^*} - \rho_2 T_{d^*}' (I_b \otimes H') R_{d^*}$$

$$+ (\rho_1 + \rho_2) T_{d^*}' (I_b \otimes H_5) R_{d^*} + \rho_2 T_{d^*}' (I_b \otimes H_6) R_{d^*} + \frac{1}{1 - \rho_2} T_{d^*}' (I_b \otimes H_7) R_{d^*} - A T_{d^*}' (I_b \otimes 1_k 1_k') R_{d^*}$$

$$= (1 + \rho_1^2 + \rho_2^2) T_{d^*}' (I_b \otimes H') T_{d^*} - \rho_1 T_{d^*}' (I_b \otimes HHH') T_{d^*} - \rho_1 T_{d^*}' (I_b \otimes H'H') T_{d^*}$$

$$- \rho_2 T_{d^*}' (I_b \otimes HH') T_{d^*} - \rho_2 T_{d^*}' (I_b \otimes H'H') T_{d^*} - A T_{d^*}' (I_b \otimes 1_k 1_k') T_{d^*}$$

$$\begin{aligned}
 &= (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes H')T_{d^*} - (\rho_1 + \rho_2)T_{d^*}'(I_b \otimes I_k)T_{d^*} - (\rho_1 + \rho_2)T_{d^*}'(I_b \otimes H'H')T_{d^*} \\
 &- AT_{d^*}'(I_b \otimes 1_k 1_k')T_{d^*} \\
 C_{d_{23}^*} &= C_{d_{32}^*} = L_{d^*}'(I_b \otimes \tilde{S})R_{d^*} \\
 &= L_{d^*}'\{I_b \otimes (1 + \rho_1^2 + \rho_2^2)I_k + (\rho_1^2 + \rho_2^2)H_3 + (1 + \rho_1^2)H_4 - \rho_1(H + H') - \rho_2(H + H') + (\rho_1 + \rho_2)H_5 \\
 &+ \rho_2H_6 + \frac{1}{1 - \rho_2}H_7 - A1_k 1_k'\}R_{d^*} \\
 &= (1 + \rho_1^2 + \rho_2^2)L_{d^*}'(I_b \otimes I_k)R_{d^*} + (\rho_1^2 + \rho_2^2)L_{d^*}'(I_b \otimes H_3)R_{d^*} + (1 + \rho_1^2)L_{d^*}'(I_b \otimes H_4)R_{d^*} \\
 &- \rho_1L_{d^*}'(I_b \otimes H)R_{d^*} - \rho_1L_{d^*}'(I_b \otimes H')R_{d^*} - \rho_2L_{d^*}'(I_b \otimes H)R_{d^*} - \rho_2L_{d^*}'(I_b \otimes H')R_{d^*} \\
 &+ (\rho_1 + \rho_2)L_{d^*}'(I_b \otimes H_5)R_{d^*} + \rho_2L_{d^*}'(I_b \otimes H_6)R_{d^*} + \frac{1}{1 - \rho_2}L_{d^*}'(I_b \otimes H_7)R_{d^*} - AL_{d^*}'(I_b \otimes 1_k 1_k')R_{d^*} \\
 &= (1 + \rho_1^2 + \rho_2^2)T_{d^*}'(I_b \otimes I_k)T_{d^*} - (\rho_1 + \rho_2)T_{d^*}'(I_b \otimes H)T_{d^*} - (\rho_1 + \rho_2)T_{d^*}'(I_b \otimes H')T_{d^*} \\
 &- AT_{d^*}'(I_b \otimes 1_k 1_k')T_{d^*}
 \end{aligned}$$

Hence all  $C_{d_{ij}^*}$  ( $1 \leq i, j \leq 2$ ) are completely symmetric.

Rewrite  $\psi = K' \alpha$  with  $K = 1_3 \otimes I_t$ . It is obvious that  $K'K = 2I_t$ . By Lemma and equation, for any design  $d \in \Omega_{(t,b,k)}$ ,

$$\begin{aligned}
 C_d[\psi] &\leq \frac{1}{9} K' C_d[\alpha] K \\
 &= \frac{1}{4} (C_{d_{11}} + C_{d_{12}} + C_{d_{13}} + C_{d_{21}} + C_{d_{22}} + C_{d_{23}} + C_{d_{31}} + C_{d_{32}} + C_{d_{33}}) \\
 &= \frac{1}{9} [3T_{d^*}'(I_b \otimes \tilde{S})T_{d^*} + 2T_{d^*}'(I_b \otimes \tilde{S})L_{d^*} + 2L_{d^*}'(I_b \otimes \tilde{S})T_{d^*} + L_{d^*}'(I_b \otimes \tilde{S})R_{d^*} + L_{d^*}'(I_b \otimes \tilde{S})L_{d^*} \\
 &= \frac{1}{3} \{ (3 + 3\rho_1^2 + 3\rho_2^2 - 4\rho_1 - 4\rho_2)T_{d^*}'(I_b \otimes I_k)T_{d^*} + (2 + 2\rho_1^2 + 2\rho_2^2 - 3\rho_1 - 4\rho_2)T_{d^*}'(I_b \otimes H)T_{d^*} + \\
 &(2 + 2\rho_1^2 + 2\rho_2^2 - 7\rho_1 - 7\rho_2)T_{d^*}'(I_b \otimes H')T_{d^*} - 2(\rho_1 + \rho_2)T_{d^*}'(I_b \otimes HH)T_{d^*} + \\
 &(2 + 2\rho_1^2 + 2\rho_2^2 - 2\rho_1 - 2\rho_2)T_{d^*}'(I_b \otimes H'H')T_{d^*} - 2(\rho_1 + \rho_2)T_{d^*}'(I_b \otimes H'H'H')T_{d^*} - 9AT_{d^*}'(I_b \otimes 1_k 1_k')T_{d^*}
 \end{aligned}$$

For any design in  $d \in \Omega_{(t,b,k)}$  with no treatment neighbor of itself at distance upto 3,

$tr(T_{d^*}'(I_b \otimes H)T_{d^*}), tr(T_{d^*}'(I_b \otimes H')T_{d^*}), tr(T_{d^*}'(I_b \otimes HH)T_{d^*}), tr(T_{d^*}'(I_b \otimes H'H')T_{d^*}),$   
 $tr(T_{d^*}'(I_b \otimes HHH)T_{d^*}), tr(T_{d^*}'(I_b \otimes H'H'H')T_{d^*})$  are all zero. Moreover, both  
 $tr(T_{d^*}'(I_b \otimes HHH)T_{d^*}), tr(T_{d^*}'(I_b \otimes H'H'H')T_{d^*})$  are non-negative. The remainder of the proof follows from the proof of the previous theorem.

#### IV. Optimal Equivalence Classes Of Sequences

In the present section we discuss the optimality of continuous block designs, by applying the method derived by Kunert and Martin (2000b). For  $u = 1, 2, \dots, b$ , let  $T_{du}$  be the incidence matrix of the direct effects of the treatment in block  $u$ ,  $1 \leq u \leq b$ . Then  $T_d = (T_{d_1}, T_{d_2}, \dots, T_{d_b})'$  is just the incidence matrix of the direct effects. For each  $u$ , define  $L_{du} = HT_{du}$ ,  $R_{du} = H'T_{du}$ . Thus, it is obvious that  $L_d = (I_b \otimes H)T_d$  and  $L_d = (I_b \otimes H')T_d$  are exactly the incidence matrices of the left-Neighbor effects and of the right-Neighbor effects. Now consider,

$$C_d[\phi] \leq C_d[K' \alpha K]$$

$$C_d[\phi] \leq \sum_{u=1}^b C_{du}$$

Where  $C_{du} = \frac{1}{4} \{ 2(1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)T'_{du}T_{du} - 4AT'_{du}1_k1'_kT_{du} + 2(1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)T'_{du}L_{du} - 2(\rho_1 + \rho_2)T'_{du}HHT_{du} \}$

Now deriving the trace of the matrix  $C_{du}$ , we get,

$$tr(C_{du}) = \frac{1}{2} tr \{ (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)T'_{du}T_{du} - 2AT'_{du}1_k1'_kT_{du} + (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)T'_{du}L_{du} - (\rho_1 + \rho_2)T'_{du}HHT_{du} \}$$

$$\Rightarrow tr(C_{du}) = \frac{1}{2} \{ (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)k - 2A \sum_{i=1}^t n_i^2 + (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2) \sum_{i=1}^t m_i - (\rho_1 + \rho_2) \sum_{i=1}^t p_i \}$$

Two sequences of treatments on a block are equivalent if one sequence can be obtained from the other by relabeling the treatments and denote by  $s$  the equivalence class of the sequence  $l$  on the block  $u$ . Because  $tr(C_{du})$  are invariant under permutations of treatment labels, so the value  $tr(C_{du})$  remains the same for any sequence in the same equivalence class. Thus, we can define,

$$c(s) = tr(C_{du}) = \frac{1}{2} \left[ (1 + \rho_1^2 + \rho_2^2 - \rho_1 - \rho_2)k + (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2) \sum_{i=1}^t m_i - (\rho_1 + \rho_2) \sum_{i=1}^t p_i - 2A \sum_{i=1}^t n_i^2 \right] \quad \text{----- (5)}$$

Where,

$$A = \frac{(1 - \rho_1 - \rho_2)^4 (1 + \rho_1^2 - 2\rho_1 - \rho_2 + \rho_1\rho_2)^4 (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2 - 2\rho_1\rho_2)^{2(k-4)}}{2(k-4)(1 + \rho_1^2 + \rho_2^2) - 2(k-3)\rho_1(1 - \rho_2) - 2(k-2)\rho_2 - 4\rho_1 + 2(1 + \rho_1^2)}$$

$n_i$  is the number of occurrences of treatment  $i$  in the sequence  $l$ ,

$m_i$  is the number of times treatment  $i$  is on the left-hand side of itself in the sequence  $l$

$p_i$  is the number of plots having treatment  $i$  both on the left-hand side and on the right-hand side.

In this section our ultimate aim would be in finding out the optimal equivalence classes of sequence. This optimal sequence is the sequence which maximizes the  $c(s)$  in (5) as explained by Kushner (1997).

**PROPOSITION:**

When  $\rho_1, \rho_2 \in (0.2199, 1)$  for any positive integer  $k \geq 5$ , if  $k$  is odd, then the optimal sequence has the form of  $'a_1 a_2 a_3 a_3 \dots a_{[k/2]} a_{[k/2]}'$ , while if  $k$  is even, then the optimal sequence has the form of  $'a_1 a_1 a_2 a_2 \dots a_{[k/2]} a_{[k/2]}'$ , where  $a_1, \dots, a_{[k/2]}$  are distinct treatments.

**PROOF:**

If  $\sum_{i=1}^t p_i$  decreases by one unit, then  $\sum_{i=1}^t m_i$  decreases definitely by one unit, and correspondingly  $c(s)$  (5) will increase by  $(\rho_1 + \rho_2) - 2(1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2)$ . Also for the value  $\rho_1$  and  $\rho_2$  between 0.2199 and 1, the above increment takes the positive value. Thus from the Proposition 3 of Ai, Yu and He (2009), we have the proof of this theorem.

Now consider the blocks of size  $k=6$ . It contains the possible treatment sequences for  $k=6$ . Out of which we are going to consider the optimal treatment sequence.

**Table 1. Optimal sequences for all possible pairs of  $\{v, v_1\}$  for  $k=6$**

S.No	OPTIMAL SEQUENCE	v	$v_1$	$tr(C_{du})$
1	aaabbb	2	0	$5\rho_1^2 + 5\rho_2^2 - 8\rho_1 - 8\rho_2 - 18A + 5$
2	aabbbb	2	1	$\frac{1}{2}(10\rho_1^2 + 10\rho_2^2 - 17\rho_1 - 17\rho_2 - 52A + 10)$
3	aabbcc	3	0	$\frac{3}{2}(3\rho_1^2 + 3\rho_2^2 - 4\rho_1 - 4\rho_2 - 8A + 3)$
4	abbccc	3	0	$\frac{1}{2}(9\rho_1^2 + 9\rho_2^2 - 12\rho_1 - 12\rho_2 - 28A + 9)$
5	abcdef	6	0	$3(\rho_1^2 + \rho_2^2 - \rho_1 - \rho_2 - 2A + 1)$

Here,  $A = \frac{(1 - \rho_1 - \rho_2)^4 (1 + \rho_1^2 - 2\rho_1 - \rho_2 + \rho_1\rho_2)^4 (1 + \rho_1^2 + \rho_2^2 - 2\rho_1 - 2\rho_2 - 2\rho_1\rho_2)^4}{6\rho_1^2 + 4\rho_2^2 - 10\rho_1 - 8\rho_2 + 6\rho_1\rho_2 + 6}$

From the above sequences, the sequence ‘‘aabbcc’’ is the optimal sequence by Proposition 3.

The below table represents all the optimal sequences for  $6 \leq k \leq 11$ . Also Note that the below table shows the optimal sequence and the last column lists the values  $tr(C_{du})$  of a CNBD (2) d.

Block Size	Optimal sequence	c (s*)	$tr(C_{du})$
6	aabbcc	$1/3(\rho^2 - \rho + 1)$	$3/2 (\rho^2 - 2\rho + 1)$
7	aabbccc abbccdd	$1/7(27\rho^2 - 33\rho + 27)$ $1/7(26\rho^2 - 31\rho + 26)$	$2(\rho^2 - 2\rho + 1)$
8	aabbccdd	$5\rho^2 - 6\rho + 5$	$5/2(\rho^2 - 2\rho + 1)$
9	aaabbbccc aabbccddd abbccdde	$6\rho^2 - 9\rho + 6$ $1/9(15\rho^2 - 11\rho + 15)$ $1/9(51\rho^2 - 66\rho + 51)$	$3(\rho^2 - 2\rho + 1)$
10	aabbccddd aabbccdee	$1/10(71\rho^2 - 102\rho + 71)$ $7\rho^2 - 9\rho + 7$	$7/2(\rho^2 - 2\rho + 1)$
11	aaabbbccddd aabbccdeee abbccdeeff	$1/11(99\rho^2 - 144\rho + 99)$ $1/11(89\rho^2 - 123\rho + 89)$ $1/11(84\rho^2 - 113\rho + 84)$	$4(\rho^2 - 2\rho + 1)$

**5. Efficiency of CNBD (2) with blocks of size  $6 \leq k \leq 11$**

In this section we are going to discuss the Efficiency of CNBD (2) for various block size. In previous section we showed that for different block size k, the CNBD(2) is universally optimal over the class of all designs from  $\Omega_{(t,b,k)}$  for  $|\rho| < 1$ . Here the efficiency of CNBD (2) is demonstrated by having the optimal continuous block design as the base. Since the values  $tr(C_{du})$  are invariant to any block u for aCNBD (2), so we can define the efficiency of aCNBD (2) d relative to the optimal continuous block design  $d^*$  as

$$Eff(d) = \frac{tr(C_d)}{tr(C_{d^*})} = \frac{tr(C_{du})}{c(s^*)}$$

We will demonstrate the calculation of  $tr(C_{du})$  by making use of the expression derived in the previous sections just by substituting different values for  $\rho_1, \rho_2$  and we can find out the efficiency for various block size. We are going to assume the values for both  $\rho_1, \rho_2$  to be -1 to +1 with 0.2 increments, avoiding the other combinations of  $\rho_1, \rho_2$  since these combinations giving the negative values for the efficiency. The below tables show the calculations of  $tr(C_{du})$  and  $c(s^*)$  for  $k= 6, 7, \dots, 11$

**Table 2 Efficiency of CNBD (2) when the block size k=6**

$\rho_1$	$\rho_2$	tr(Ca)	tr(Ca')	Eff(d)
-1	-1	-62,934.60	-31,462.80	0.4999
-0.8	-0.8	-17,131.43	-8,562.29	0.4998
-0.6	-0.6	-3,796.02	-1,895.43	0.4993
-0.4	-0.4	-623.20	-309.62	0.4968
-0.2	-0.2	-57.20	-26.98	0.4716
0	0	5.00	4.00	0.8
0.2	0.2	4.87	4.05	0.8329
0.4	0.4	2.28	3.12	1.3684
0.6	0.6	1.08	3.12	2.8888
0.8	0.8	1.32	4.08	3.0915
1	1	3.00	6.00	2

Now we present the efficiency of CNBD(2) corresponding to the optimal continuous block design for different block size.

**Table 3 Efficiency of CNBD (2)**

$\rho_1$	$\rho_2$	Block Size						
		5	6	7	8	9	10	11
-1	-1	0.5555	0.4999	0.5384	0.4999	0.4283	0.4998	0.4397
-0.8	-0.8	0.5554	0.4998	0.5382	0.4997	0.4279	0.4996	0.4392
-0.6	-0.6	0.5551	0.4993	0.5376	0.4989	0.4265	0.4985	0.4374
-0.4	-0.4	0.5533	0.4968	0.5343	0.4949	0.4189	0.4929	0.4277
-0.2	-0.2	0.5372	0.4717	0.4987	0.4469	0.3352	0.4113	0.2990
0	0	1.0000	0.8000	0.7778	0.7273	0.7895	0.7059	0.7476
0.2	0.2	0.8646	0.8329	0.8522	0.8311	0.8974	0.8305	0.8838
0.4	0.4	0.5556	1.3684	1.3000	1.3684	1.6957	1.3684	1.6250
0.6	0.6	2.0968	2.8889	2.2750	2.8889	6.1579	2.8889	5.1071
0.8	0.8	2.1789	3.0916	2.3801	3.0911	7.2867	3.0910	5.8441
1	1	1.6667	2.0000	1.7500	2.0000	3.0000	2.0000	2.7500

**V. Conclusion**

In this research paper, the Optimality and efficiency of the circular neighbor balanced design have been investigated by having the assumption that the errors in each block are correlated according to second order circular auto regressive process. Few results pertaining to the universal optimal designs have been proved for second order model. The traces of optimal sequence of treatments for different block size have been derived. Also the efficiency factor for CNBD (2) corresponding to the optimal continuous block design was calculated.

**References**

- [1]. AZA IS, J.M., BAILEY, R.A. and MONOD, H., (1993). A catalogue of efficient Neighbor- designs with border plots. *Biometrics* 49.
- [2]. AI, M.-y., GE, G.-n. and CHAN, L.-y. (2007). Circular neighbor-balanced designs universally optimal for total effects. *Sci. China Ser. A* 50 821–828. MR2353065
- [3]. AI, M., YU, Y. and HE, S. (2009). Optimality of circular neighbor-balanced designs for total effects with autoregressive correlated observations. *J. Statist. Plann. Inference* 139 2293–2304. MR2507991
- [4]. BAILEY, R.A. and DRUILHET, P., (2004). Optimality of Neighbor-balanced designs for total effects. *Ann. Statist.* 32(4), 1650-1661.
- [5]. BOX, G.E.P., HUNTER, J.S. and HUNTER, W.G., (2005). *Statistics for Experimenters*. John Wiley & Sons, New Jersey.
- [6]. CHENG, C.S. and WU, C.F.J., (1980). Balanced repeated measurements designs. *Ann. Statist.* 8, 1272-1283.
- [7]. D. RICHARD CUTLER. (1990) Efficient block designs for comparing test treatments to a control when the errors are correlated. *Journal of Statistical Planning and Inference* 37 (1993) 393-412 North-Holland
- [8]. DRUILHET, P., (1999). Optimality of Neighbor balanced designs. *J. Statist. Plann. Inference* 81, 141-152.
- [9]. FILIPIAK, K. and MARKIEWICZ, A., (2005). Optimality and efficiency of circular Neighbor balanced designs for correlated observations. *Metrika* 61, 17-27.
- [10]. FILIPIAK, K. (2012). Universally optimal designs under an interference model with equal left- and right-neighbor effects. *Statist. Probab. Lett.* 82 592–598. MR2887476
- [11]. HEDAYAT, A. and AFSARINEJAD, K., (1978). Repeated measurements design II. *Ann. Statist.* 6, 619 - 628.
- [12]. KUNERT, J. and MARTIN, R.J., (2000). On the determination of optimal designs for an interference model. *Ann. Statist.* 28, 1728-1742.
- [13]. KUNERT, J., (1984a). Design balanced for circular residual effects. *Comm. Statist. A-Theory Methods* 13, 2665-2671.
- [14]. KUNERT, J., (1984b). Optimality of balanced uniform repeated measurements designs. *Ann. Statist.* 12, 1006-1017.

- [15]. KUSHNER, H.B., (1997). Optimal repeated measurements designs: The linear optimality equations. *Ann. Statist.* 25, 2328-2344.
- [16]. MAGDA, C.D., (1980). Circular balanced repeated measurements design. *Comm. Statist. A- Theory Methods* 9, 1901-1918.
- [17]. PIERRE DRUILHET and TINSSON WALTER (2013). Efficient Circular neighbor designs for spatial interference model. *Journal Of Statistical Planning and Inference*. 1161 – 1169.
- [18]. SANTHARAM, C, PONNUSAMY, K.N., AND CHANDRASEKAR B., 1996: Universal Optimality of nearest Neighbor balanced block designs using ARMA models. *Biometrical J.* 38, 725 – 730.
- [19]. SANTHARAM, C, and PONNUSAMY, K.N., 1997: On the efficiencies of nearest Neighbor balanced block design with correlated error structure. *Biometrical J.* 39, 85 – 98.