# Block designs for symmetric parallel line assays with block size odd 

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# BLOCK DESIGNS FOR SYMMETRIC PARALLEL LINE ASSAYS WITH BLOCK SIZE ODD 

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$S U M M A R Y$. In a parallel line assay, there are three treatment contrasts of major importance. Block designs allowing the estimability of all the three contrasts free from block effects, called $L$-designs, necessarily have the block sizes even. For odd block sizes, we provide here a class of highly efficient designs, called nearly $L$-designs. These nearly $L$-designs have been constructed by establishing a link with linear and nearly linear trend-free designs.

## 1. Introduction

Biological assays or bioassays involve two stimuli applied to subjects. One preparation of the stimulus, called the standard preparation, has a known effect on subjects, while the other preparation of the stimulus, called the test preparation, has an unknown strength. A major purpose of a bioassay is to estimate the potency of the test relative to the standard preparation. The relative potency is defined as the ratio of two equivalent doses of the standard to the test preparation. In a bioassay, we thus have two groups of treatments, one for standard preparation and the other for test preparation. Often, within each group, the treatment effect is represented by a polynomial in the logarithm of the dose. In particular, when the polynomial has degree one and both the groups share the same slope, then the assay is called a parallel line assay. If the number of doses of both the preparations are same, then the parallel line assay is called symmetric, otherwise, it is called asymmetric. In the context of parallel line assays, three treatment contrasts (contrasts among dose effects) are of major importance. The first two, the preparation contrast and the combined regression contrast, provide an estimate of the relative potency and the third one, the parallelism contrast, is used to test the parallelism of the two regression lines. For an excellent description of the theory and application of bioassays, the reader is referred to Finney (1978).

If a block design is used for the assay, it is desirable that the design allows
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the estimability of these three contrasts with full efficiency. For symmetric parallel line assays, an equireplicate block design is called an $L$-design if the three contrasts of importance are estimated with full efficiency. L-designs have been studied quite extensively; see the review by Gupta and Mukerjee (1996) where more references can be found. Most of these $L$-designs are for even number of doses of each of the preparations. Gupta and Mukerjee (1990) suggested a somewhat unified method of construction of $L$-designs. They provided a (i) complete solution of $L$-designs for even number of doses, and (ii) table of $L$-designs for all odd number $(\leq 15)$ of doses. However, there are situations where it is impossible to construct an $L$-design. For such situations, Chai and Das (2001) introduced a class of designs, called nearly $L$-designs for symmetric parallel line assays. Recall that a necessary condition for the existence of an $L$-design is that the block size be even. The designs of of Chai and Das (2001) also require the block size to be even.

Thus, it appears that a systematic study for obtaining efficient block designs for parallel line assays with odd block sizes has not been attempted. In this paper we propose a class of designs, called nearly $L$-designs, with odd block sizes. In Section 2, some preliminaries on linear trend-free designs are given. Nearly $L$-designs are introduced in Section 3 and a link between linear trend-free (nearly linear trend-free) designs and nearly $L$-designs is established. With the help of this connection, a necessary and sufficient condition for the existence of nearly $L$-designs as well as a construction method is provided. The proposed designs are shown to be highly efficient.

## 2. Linear trend-free designs

Throughout, $D(v, b, k, r)$ will denote the class of all connected block designs with $v$ treatments each replicated $r$ times and arranged in $b$ blocks each of size $k \geq 2$. Similarly, $D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$ will denote the class of all connected block designs with $v$ treatments, $b$ blocks each of size $k \geq 2$ and the $i$ th treatment replicated $r_{i}$ times, $1 \leq i \leq v$.

Trend-free block designs were introduced by Bradley and Yeh (1980). The setup they considered involves $v$ treatments and $b$ blocks each of size $k(\geq 2)$, where, the $k$ experimental units within each block are linearly ordered over time and space. Thus each block has $k$ periods, numbered $1,2, \ldots, k$. Suppose that, in addition to treatment and block effects, there is a common polynomial trend effect within each block. The postulated model for an observation in period $l$ of block $j$ is

$$
\begin{equation*}
y_{j l}=\mu+\sum_{i=1}^{v} \delta_{j l}^{i} \tau_{i}+\beta_{j}+\sum_{\alpha=1}^{p} \phi_{\alpha}(l) \theta_{\alpha}+\epsilon_{j l} \tag{2.1}
\end{equation*}
$$

where $\mu$ is a general mean, $\tau_{1}, \ldots, \tau_{v}$, the treatment effects, $\beta_{1}, \ldots, \beta_{b}$, the block effects and $\theta_{1}, \ldots, \theta_{p}$, the trend effects. Moreover, for $1 \leq \alpha \leq p, \phi_{\alpha}(l)$, is an orthogonal polynomial of degree $\alpha$, based on $1,2, \ldots, k$, with $\sum_{l=1}^{k} \phi_{\alpha}(l)=0$ and $\sum_{l=1}^{k} \phi_{\alpha}(l) \phi_{\alpha^{\prime}}(l)=\delta_{\alpha \alpha^{\prime}}, \quad \delta_{\alpha \alpha^{\prime}}$ being the

Kronecker delta, $\alpha, \alpha^{\prime}=1, \ldots, p$. Also,

$$
\delta_{j l}^{i}= \begin{cases}1, & \text { if treatment } i \text { is applied in period } l \text { of block } j, \\ 0, & \text { otherwise },\end{cases}
$$

with $\sum_{i=1}^{v} \delta_{j l}^{i}=1$.
Let $\boldsymbol{\tau}=\left(\tau_{1}, \ldots, \tau_{v}\right)^{\prime}, \boldsymbol{\beta}=\left(\beta_{1}, \ldots, \beta_{b}\right)^{\prime}$ and $\boldsymbol{\theta}=\left(\theta_{1}, \ldots, \theta_{p}\right)^{\prime}$. A trend-free block design has the property that the presence of trend effect in a treatment-block model does not affect the analysis of the treatment effects. A design $d$ is said to be $p$-trend-free if

$$
\begin{equation*}
R_{d}(\boldsymbol{\tau} \mid \mu, \boldsymbol{\beta}, \boldsymbol{\theta})=R_{d}(\boldsymbol{\tau} \mid \mu, \boldsymbol{\beta}), \tag{2.2}
\end{equation*}
$$

where $R_{d}(\boldsymbol{\tau} \mid \mu, \boldsymbol{\beta}, \boldsymbol{\theta})$ denotes the adjusted treatment sum of squares under (2.1) and $R_{d}(\boldsymbol{\tau} \mid \mu, \boldsymbol{\beta})$ denotes $R_{d}(\boldsymbol{\tau} \mid \mu, \boldsymbol{\beta}, \boldsymbol{\theta})$ when $\boldsymbol{\theta}=\mathbf{0}$ in (2.1).

If $p=1$, then a design $d \in D(v, b, k, r)$ satisfying (2.2) is called a linear trend-free block design. Equivalently, $d$ is called a linear trend-free block design if

$$
\sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l=\frac{r(k+1)}{2}, \quad 1 \leq i \leq v
$$

Clearly, a necessary condition for a design $d \in D(v, b, k, r)$ to be linear trend-free is

$$
\begin{equation*}
r(k+1) \equiv 0(\bmod 2) . \tag{2.3}
\end{equation*}
$$

Stufken (1988) showed that (2.3) is both necessary and sufficient for the existence of a linear trend-free block design. The result of Stufken (1988) has recently been generalized by Chai (2002), who shows that a linear trend-free design exists in $D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$ if and only if $r_{i}(k+1) \equiv 0(\bmod 2)$ for each $i, 1 \leq i \leq v$. In $D(v, b, k, r)$, when $k$ is even and $r$ is odd, Yeh, Bradley and Notz (1985) defined a class of designs, called nearly linear trend-free design. We give a more general definition of a nearly linear trend-free design belonging to $D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$.

Definition 1. For a design belonging to $D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$, suppose $k$ is even and at least one of the $r_{i}$ 's is odd. Then $d \in D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$ is called a nearly linear trend-free block design if for $1 \leq i \leq v, \sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l$ equals (a) either $\frac{r_{i}(k+1)-1}{2}$ or $\frac{r_{i}(k+1)+1}{2}$, if $r_{i}$ is odd and, (b) $\frac{r_{i}(k+1)}{2}$, if $r_{i}$ is even.

We have the following result which shows that a nearly linear trend-free design, as per Definition 1, always exists. A proof of Theorem 1 appears in the Appendix.

Theorem 1. Suppose $k$ is even and at least one of $r_{i}$ 's is odd. Then a nearly linear trend-free block design exists in $D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$.

For the purpose of obtaining designs for parallel line assays, we need to consider a class of nearly linear trend-free designs with the following parametric structure : there are $v \equiv$ $0(\bmod 4)$ treatments which can be split into two sets, say $S_{1}$ and $S_{2}$, both with cardinality $\frac{1}{2} v$. Furthermore, each treatment in $S_{1}$ has replication $r_{1} \equiv 1(\bmod 2)$ and each treatment belonging to $S_{2}$ has replication $r_{2} \equiv 0(\bmod 2)$. Then, since $\sum_{i \in S_{1}} \sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l=v r_{1}(k+1) / 4$, it is easy to see that for $\frac{1}{4} v$ treatments in $S_{1}, \sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l=\frac{r_{1}(k+1)-1}{2}$ and for the remaining $\frac{1}{4} v$ treatments in $S_{1}, \sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l=\frac{r_{1}(k+1)+1}{2}$. Also, for each treatment $i \in S_{2}, \sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l=$ $\frac{r_{2}(k+1)}{2}$.

## 3. Nearly $L$-designs

The three contrasts of major importance in the context of parallel line assays are preparation $\left(L_{p}\right)$, combined regression $\left(L_{1}\right)$ and parallelism $\left(L_{1}^{\prime}\right)$. The three contrasts, $L_{p}, L_{1}, L_{1}{ }^{\prime}$, in the context of symmetric parallel line assays, can be explicitly written as

$$
\begin{equation*}
L_{p}=m^{-1}\left(\mathbf{1}_{m}^{\prime},-\mathbf{1}_{m}^{\prime}\right) \boldsymbol{\tau}, \quad L_{1}=\delta_{0}\left(\boldsymbol{w}^{\prime}, \boldsymbol{w}^{\prime}\right) \boldsymbol{\tau}, \quad L_{1}^{\prime}=2 \delta_{0}\left(\boldsymbol{w}^{\prime},-\boldsymbol{w}^{\prime}\right) \boldsymbol{\tau} \tag{3.1}
\end{equation*}
$$

where $v=2 m, \mathbf{1}_{s}$ is a $s \times 1$ vector of all ones, $\delta_{0}=6 /\left\{\theta_{0} \log h\right\}, \theta_{0}=m\left(m^{2}-1\right)$ and $\boldsymbol{w}=(1,2, \ldots, m)^{\prime}-\frac{1}{2}(m+1) \mathbf{1}_{m}$.

Suppose a symmetric parallel line assay involving $m$ doses of each of the preparations is conducted in $b$ blocks each of size $k$. As mentioned earlier, there are $v=2 m$ treatments, in which the first $m$ treatments represent the doses of standard preparation and the last $m$ treatments represent the doses of the test preparation. Each treatment is replicated $r=b k / v$ times. Let $N_{d}$ be the incidence matrix of $d \in D(v=2 m, b, k, r)$. We postulate a fixed effects additive model for the data collected through $d$, making the usual assumption that errors are independent with mean zero and variance $\sigma^{2}$. Under such a model the information matrix of the reduced normal equations for estimating contrasts among dose effects, using a design $d$, is $C_{d}=r I-k^{-1} N_{d} N_{d}^{\prime}$ where $I$ is the identity matrix. Every contrast among dose effects is estimable via $d$ if and only if $\operatorname{Rank}\left(C_{d}\right)=v-1$ and in such a case the design $d$ is called connected. Note that $N_{d}$ may be partitioned as $N_{d}=\left(N_{1 d}^{\prime}, N_{2 d}^{\prime}\right)^{\prime}$, where $N_{1 d}\left(N_{2 d}\right)$ is the $m \times b$ incidence matrix for the $m$ doses of the standard (test) preparation. Hence we have

$$
\begin{align*}
& \mathbf{1}_{m}^{\prime} N_{1 d}+\mathbf{1}_{m}^{\prime} N_{2 d}=k \mathbf{1}_{b}^{\prime}  \tag{3.2}\\
& N_{i d} \mathbf{1}_{b}=r \mathbf{1}_{m}, \quad i=1,2 \tag{3.3}
\end{align*}
$$

From Lemma 3.1 of Gupta and Mukerjee (1996), a design $d \in D(v=2 m, b, k, r)$ retains full information on $L_{p}, L_{1}$ and $L_{1}{ }^{\prime}$ if and only if

$$
\left[\begin{array}{rr}
\mathbf{1}_{m}^{\prime} & -\mathbf{1}_{m}^{\prime}  \tag{3.4}\\
\boldsymbol{w}^{\prime} & \boldsymbol{w}^{\prime} \\
\boldsymbol{w}^{\prime} & -\boldsymbol{w}^{\prime}
\end{array}\right]\left[\begin{array}{c}
N_{1 d} \\
N_{2 d}
\end{array}\right]=\mathbf{0}
$$

where $\mathbf{0}$ is a null matrix (or, vector) of appropriate order.
A block design $d \in D(v=2 m, b, k, r)$ satisfying (3.4) is called an $L$-design. It follows from (3.2) - (3.4), that $d \in D(v=2 m, b, k, r)$ is an $L$-design if and only if

$$
\begin{equation*}
\mathbf{1}_{m}^{\prime} N_{1 d}=\mathbf{1}_{m}^{\prime} N_{2 d}=\frac{1}{2} k \mathbf{1}_{b}^{\prime} ; \quad \boldsymbol{w}^{\prime} N_{1 d}=\boldsymbol{w}^{\prime} N_{2 d}=\mathbf{0} \tag{3.5}
\end{equation*}
$$

Clearly, from (3.5) it follows that a necessary condition for an $L$-design to exist is that $k \equiv 0(\bmod 2)$. Furthermore, Chai (2002) has shown that a necessary and sufficient condition for an $L$-design in $D(v=2 m, b, k, r)$ to exist is that $\frac{1}{2} k(m+1) \equiv 0(\bmod 2)$. Thus, one cannot construct an $L$-design if either of the following conditions hold:
(i) $k \equiv 1(\bmod 2)$;
(ii) $k \equiv 2(\bmod 4)$ and $m \equiv 0(\bmod 2)$.

When $k \equiv 2(\bmod 4)$ and $m \equiv 0(\bmod 2)$, Chai and Das $(2001)$ defined a class of designs, called nearly $L$-designs. These designs allow the estimability of $L_{p}$ and $L_{1}$ free from block effects. In this paper, we attempt to construct highly efficient block designs for parallel line assays when $k$ is odd. In the rest of the paper, we take $k>2$ to be an odd integer. We continue to call such designs nearly $L$-designs. Clearly, in such a case, $\mathbf{1}_{m}^{\prime} N_{1 d} \neq \mathbf{1}_{m}^{\prime} N_{2 d}$ and thus such designs do not allow the estimability $L_{p}$ free from block effects. We formally define nearly $L$-designs considered in this paper.

Definition 2. A block design $d \in D(v=2 m, b, k, r)$ with $k(>2)$ odd is called a nearly $L$-design if the following are true :
(a) $\mathbf{1}_{m}^{\prime} N_{1 d}=\left(\frac{k+1}{2} \mathbf{1}_{\frac{b}{2}}^{\prime}, \frac{k-1}{2} \mathbf{1}_{\frac{b}{2}}^{\prime}\right)$;
(b) $\mathbf{1}_{m}^{\prime} N_{2 d}=\left(\frac{k-1}{2} \mathbf{1}_{\frac{b}{2}}^{\prime}, \frac{k-1}{2} \mathbf{1}_{\frac{b}{2}}^{\prime}\right)$.

Furthermore if $m$ is odd,
(c) $\boldsymbol{w}^{\prime} N_{1 d}=\boldsymbol{w}^{\prime} N_{2 d}=\mathbf{0}^{\prime}$,
and, if $m$ is even,
(c') $\boldsymbol{w}^{\prime} N_{1 d}=\frac{1}{2}\left(\mathbf{1}_{\frac{b}{4}}^{\prime},-\mathbf{1}_{\frac{b}{4}}^{\prime}, \mathbf{0}^{\prime}\right) ; \boldsymbol{w}^{\prime} N_{2 d}=\frac{1}{2}\left(\mathbf{0}^{\prime}, \mathbf{1}_{\frac{b}{4}}^{\prime},-\mathbf{1}_{\frac{b}{4}}^{\prime}\right)$.
The normalized contrasts corresponding to $L_{p}, L_{1}$ and $L_{1}{ }^{\prime}$ are given respectively by $\boldsymbol{g}_{1}^{\prime} \boldsymbol{\tau}, \boldsymbol{g}_{2}^{\prime} \boldsymbol{\tau}, \boldsymbol{g}_{3}^{\prime} \boldsymbol{\tau}$, where $\boldsymbol{g}_{1}=(2 m)^{-1 / 2}\left(\mathbf{1}_{m}^{\prime}, \quad-\mathbf{1}_{m}^{\prime}\right)^{\prime}, \boldsymbol{g}_{2}=\left[m\left(m^{2}-1\right) / 6\right]^{-1 / 2}\left(\boldsymbol{w}^{\prime}, \quad \boldsymbol{w}^{\prime}\right)^{\prime}$ and $\boldsymbol{g}_{3}=\left[m\left(m^{2}-\right.\right.$ $1) / 6]^{-1 / 2}\left(\boldsymbol{w}^{\prime},-\boldsymbol{w}^{\prime}\right)^{\prime}$. Let $G=\left(\left(g_{i j}\right)\right)$ be a $3 \times v$ matrix with rows $\boldsymbol{g}_{1}^{\prime}, \boldsymbol{g}_{2}^{\prime}$ and $\boldsymbol{g}_{3}^{\prime}$.

Let $N_{d}=\left(N_{1 d}^{\prime}, N_{2 d}^{\prime}\right)^{\prime}$ be the incidence matrix of a nearly $L$-design $d \in D(v=2 m, b, k, r)$. Further, we restrict attention to a convenient family of nearly $L$-designs $\{d\}$ for which $N_{1 d}=$ $\left[M_{1 d}, M_{2 d}\right], \quad N_{2 d}=\left[M_{2 d}, M_{1 d}\right]$ for some $m \times \frac{1}{2} b$ matrices $M_{d 1}, M_{d 2}$. Then, from Definition 2, we have

$$
\mathbf{1}_{m}^{\prime} M_{1 d}=\frac{k+1}{2} \mathbf{1}_{\frac{b}{2}}^{\prime} ; \mathbf{1}_{m}^{\prime} M_{2 d}=\frac{k-1}{2} \mathbf{1}_{\frac{b}{2}}^{\prime} .
$$

Let $G=V C_{d}$ for some $3 \times 2 m$ matrix $V=\left[\begin{array}{cc}\boldsymbol{v}_{11}^{\prime} & \boldsymbol{v}_{12}^{\prime} \\ \boldsymbol{v}_{21}^{\prime} & \boldsymbol{v}_{22}^{\prime} \\ \boldsymbol{v}_{31}^{\prime} & \boldsymbol{v}_{32}^{\prime}\end{array}\right]$ where each $\boldsymbol{v}_{i j}$ is an $m \times 1$ vector (see pages $883-884$ of Gupta and Mukherjee (1996)).

For an arbitrary $d \in D(v=2 m, b, k, r)$, the covariance matrix of $G \hat{\boldsymbol{\tau}}$, the best linear unbiased estimator of $G \boldsymbol{\tau}$, under $d$, is

$$
\begin{gathered}
\operatorname{Cov}(G \hat{\boldsymbol{\tau}})_{d}=\sigma^{2} V C_{d} V^{\prime} \\
\operatorname{Cov}(G \hat{\boldsymbol{\tau}})-\frac{\sigma^{2}}{r} G G^{\prime}=\sigma^{2} V\left(C_{d}-r^{-1} C_{d} C_{d}^{\prime}\right) V^{\prime} .
\end{gathered}
$$

Hence,

$$
\begin{equation*}
\operatorname{Cov}(G \hat{\boldsymbol{\tau}})_{d}=\sigma^{2} r^{-1} G G^{\prime}+\sigma^{2} V\left(C_{d}-r^{-1} C_{d}^{2}\right) V^{\prime}=\sigma^{2}\left(r^{-1} G G^{\prime}+(r k)^{-1} G N_{d} N_{d}^{\prime} V^{\prime}\right) \tag{3.6}
\end{equation*}
$$

We consider two cases, according as $m$ is odd or even.
Case (i). $m$ is odd. We seek a nearly $L$-design $d_{0}$ for parallel line assays with parameters $k=2 k_{1}+1, m=2 m_{1}+1\left(\Rightarrow v=2 m=2\left(2 m_{1}+1\right)\right), b=2 b_{1}, r=\frac{b k}{v}=\frac{b_{1}\left(2 k_{1}+1\right)}{2 m_{1}+1}$. Here, $k_{1}$ and $m_{1}$ are positive integers.

As a first step, we construct a linear trend-free design $d^{*}$ with parameters $v^{*}=b, b^{*}=$ $r, k^{*}=m, r_{1}^{*}=\cdots=r_{\frac{v^{*}}{2}}^{*}=k_{1}+1 ; r_{\frac{v^{*}}{2}+1}^{*}=\cdots=r_{v^{*}}^{*}=k_{1}$. Such a design can be a constructed, since $r_{i}^{*}\left(k^{*}+1\right) \equiv 0(\bmod 2)$ for $1 \leq i \leq v^{*}$.

From $d^{*}$, we construct a design $d_{0}$ as follows : Suppose, without loss of generality that $k_{1}$ is even, so that $k_{1}+1$ is odd. Write the blocks of $d^{*}$ as columns of a $k^{*} \times b^{*}$ matrix, say $\Delta$. Now construct a $k^{*} \times v^{*}$ matrix, $N_{1 d_{0}}$, whose columns are indexed by the $v^{*}$ treatments of $d^{*}$ and the rows by the positions of the treatments in each column (block). If a treatment symbol $j$ appears in the $i$ th row of $\Delta n_{i j}^{1}$ times, then the $(i, j)$ th element of $N_{1 d_{0}}$ is $n_{i j}^{1}$ and, zero, otherwise. Let the $m \times \frac{1}{2} b$ matrix consisting of the first $\frac{1}{2} b$ columns of $N_{1 d_{0}}$ be denoted by $M_{1 d_{0}}$ and the matrix consisting of the last $\frac{1}{2} b$ columns of $N_{1 d_{0}}$ be $M_{2 d_{0}}$. Then, $N_{1 d_{0}}=\left[M_{1 d_{0}}, M_{2 d_{0}}\right]$. Define $N_{2 d_{0}}=\left[M_{2 d_{0}}, M_{1 d_{0}}\right]$. The required nearly $L$-design $d_{0} \in D(v=2 m, b, k, r)$ has incidence matrix $N_{d_{0}}=\left[\begin{array}{c}N_{1 d_{0}} \\ N_{2 d_{0}}\end{array}\right]$. Then $d_{0}$ has the following properties :
(i) $N_{1 d_{0}}=\left[M_{1 d_{0}}, M_{2 d_{0}}\right]$ and $N_{2 d_{0}}=\left[M_{2 d_{0}}, M_{1 d_{0}}\right]$.
(ii) $\mathbf{1}_{m}^{\prime} M_{1 d_{0}}=\left(k_{1}+1\right) \mathbf{1}_{b_{1}}^{\prime}, \quad \mathbf{1}_{m}^{\prime} M_{2 d_{0}}=k_{1} \mathbf{1}_{b_{1}}^{\prime}, \quad \boldsymbol{w}^{\prime} M_{1 d_{0}}=\mathbf{0}^{\prime}, \quad \boldsymbol{w}^{\prime} M_{2 d_{0}}=\mathbf{0}^{\prime}$.

From (i) and (ii) above, it follows that

$$
\begin{aligned}
& \frac{1}{r k} G N_{d_{0}} N_{d_{0}}^{\prime} V^{\prime} \\
= & \frac{1}{r k} \frac{1}{\sqrt{2 m}}\left(\begin{array}{ccc}
\mathbf{1}_{b_{1}}^{\prime}\left(M_{1 d_{0}}-M_{2 d_{0}}\right)^{\prime}\left(\boldsymbol{v}_{11}-\boldsymbol{v}_{12}\right) & 0 & 0 \\
0 & 0 & 0 \\
0 & 0 & 0
\end{array}\right) .
\end{aligned}
$$

Hence, under $d_{0}$,

$$
\operatorname{Cov}(G \hat{\boldsymbol{\tau}})_{d_{0}}=\sigma^{2}\left(\begin{array}{ccc}
\frac{1}{r}+\frac{1}{r k} \frac{1}{\sqrt{2 m}} \mathbf{1}_{b_{1}}^{\prime}\left(M_{1 d_{0}}-M_{2 d_{0}}\right)^{\prime}\left(\boldsymbol{v}_{11}-\boldsymbol{v}_{12}\right) & 0 & 0  \tag{3.7}\\
0 & \frac{1}{r} & 0 \\
0 & 0 & \frac{1}{r}
\end{array}\right)
$$

From (3.7), we have $\operatorname{Var}\left(\boldsymbol{g}_{1}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d_{0}}=\sigma^{2}\left(\frac{1}{r}+\frac{1}{r k} \frac{1}{\sqrt{2 m}} 1_{b_{1}}^{\prime}\left(M_{1 d_{0}}-M_{2 d_{0}}\right)^{\prime}\left(\boldsymbol{v}_{11}-\boldsymbol{v}_{12}\right)\right.$ and $\operatorname{Var}\left(g_{2}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d_{0}}=$ $\operatorname{var}\left(g_{3}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d_{0}}=\sigma^{2} r^{-1}$, where for $i=1,2,3, \operatorname{Var}\left(\boldsymbol{g}_{i}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d_{0}}$ is the variance of the best linear unbiased estimator of $\boldsymbol{g}_{i}^{\prime} \boldsymbol{\tau}$ under $d_{0}$. Since for an arbitrary design $d \in D(v=2 m, b, k, r)$, $\operatorname{Var}\left(\boldsymbol{g}_{i}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} \geq \sigma^{2} / r, i=1,2,3$, it follows that the design $d_{0}$ estimates the contrasts $L_{1}$ and $L_{1}^{\prime}$ with full information.

Now let us concentrate on the contrast $\boldsymbol{g}_{1}^{\prime} \boldsymbol{\tau}$. Let $d \in D(v=2 m, b, k, r)$ be arbitrary and as before, let $\operatorname{Var}\left(\boldsymbol{g}_{1}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d}$ denote the variance of the best linear unbiased estimator of $\boldsymbol{g}_{1}^{\prime} \boldsymbol{\tau}$ under $d$. Then,

$$
\sigma^{-2} \operatorname{Var}\left(\boldsymbol{g}_{1}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d}=\boldsymbol{g}_{1}^{\prime} C_{d}^{-} \boldsymbol{g}_{1} \geq\left(\boldsymbol{g}_{1}^{\prime} C_{d} \boldsymbol{g}_{1}\right)^{-1} \geq \frac{1}{\max _{d \in D} \boldsymbol{g}_{1}^{\prime} C_{d} \boldsymbol{g}_{1}}
$$

Now,

$$
\begin{aligned}
\max _{d \in D} \boldsymbol{g}_{1}^{\prime} C_{d} \boldsymbol{g}_{1} & =r-\min _{d \in D} k^{-1} \boldsymbol{g}_{1}^{\prime} N_{d} N_{d}^{\prime} \boldsymbol{g}_{1} \\
& =r-(2 m k)^{-1} \min _{d \in D} \sum_{j=1}^{b}\left(a_{d 1 j}-a_{d 2 j}\right)^{2}
\end{aligned}
$$

where $\left(a_{d 11}, \ldots, a_{d 1 b}\right)=\mathbf{1}_{m}^{\prime} N_{1 d},\left(a_{d 21}, \ldots, a_{d 2 b}\right)=\mathbf{1}_{m}^{\prime} N_{2 d}$. The minimum of $\sum_{j=1}^{b}\left(a_{d 1 j}-a_{d 2 j}\right)^{2}$ is attained when

$$
\begin{equation*}
\left|a_{d 1 j}-a_{d 2 j}\right|=1, \text { for all } j=1, \ldots, b \tag{3.8}
\end{equation*}
$$

as $k>2$ is odd. Hence, when (3.8) holds, we have

$$
\max _{d \in D}\left(\boldsymbol{g}_{1}^{\prime} C_{d} \boldsymbol{g}_{1}\right)=r-b(2 m k)^{-1}=r\left(1-k^{-2}\right)
$$

Therefore,

$$
\sigma^{-2} \operatorname{Var}\left(\boldsymbol{g}_{1}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} \geq \frac{1}{r\left(1-k^{-2}\right)}
$$

On the basis of the above analysis, one can obtain a lower bound to the efficiency factor of the contrast $\boldsymbol{g}_{1}^{\prime} \boldsymbol{\tau}$ under a design $d$ as

$$
e_{d}=\sigma^{2} /\left\{\operatorname{Var}\left(\boldsymbol{g}_{1}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} r\left(1-k^{-2}\right)\right\}
$$

Also, a lower bound to the measure of an overall efficiency factor of a design $d$, based on all the three contrasts, is given by $\bar{e}_{d}=\frac{\sigma^{2}\left(3 k^{2}-2\right)}{r\left(k^{2}-1\right)} / \sum_{1 \leq i \leq 3} \operatorname{Var}\left(\boldsymbol{g}_{i}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d}$.

Note that the proposed design $d_{0}$ satisfies (3.8) and also estimates $\boldsymbol{g}_{i}^{\prime} \boldsymbol{\tau}, \quad i=2,3$ with efficiency one. Thus, it is expected that $d_{0}$ will have a high overall efficiency factor - in fact, in several examples, it is found that $\bar{e}_{d_{0}}>0.95$. Thus when $m$ is odd, the design $d_{0}$ allows the estimability of the contrasts $\boldsymbol{g}_{2}^{\prime} \boldsymbol{\tau}$ and $\boldsymbol{g}_{3}^{\prime} \boldsymbol{\tau}$ with efficiency one, while the efficiency factor of $\boldsymbol{g}_{1}^{\prime} \boldsymbol{\tau}\left(L_{p}\right)$ is expected to be close to unity for appropriately chosen $d_{0}$.

Example 1. Let $m=5, k=5, b=12, r=6$. We first find a linear trend-free block design $d^{*} \in D\left(v^{*}=12, b^{*}=6, k^{*}=5, r_{1}^{*}=\cdots=r_{6}^{*}=3 ; r_{7}^{*}=\cdots=r_{12}^{*}=2\right)$. Such a design, with
columns as blocks, is shown below.

$$
d^{*} \equiv \begin{array}{rrrrrr}
5 & 4 & 9 & 7 & 8 & 6 \\
11 & 12 & 1 & 3 & 2 & 10 \\
1 & 2 & 6 & 4 & 5 & 3 \\
12 & 10 & 2 & 1 & 3 & 11 \\
8 & 7 & 5 & 9 & 6 & 4
\end{array}
$$

Following the method of construction of $d_{0}$ described in this section, we have

$$
\begin{aligned}
N_{1 d_{0}} & =\left[\begin{array}{llllllllllll}
0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 \\
1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 \\
1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 \\
0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0
\end{array}\right] \\
N_{2 d_{0}} & =\left[\begin{array}{llllllllllll}
1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 \\
0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 \\
0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 \\
1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1
\end{array}\right]
\end{aligned}
$$

and $N_{d_{0}}=\left[\begin{array}{c}N_{1 d_{0}} \\ N_{2 d_{0}}\end{array}\right]$ is the incidence matrix of $d_{0} \in D(10,12,5,6)$. The blocks of the design $d_{0}$ are

$$
\begin{aligned}
& (2,3,4,6,10),(2,3,4,6,10),(2,3,4,6,10),(1,3,5,7,9),(1,3,5,7,9),(1,3,5,7,9) \\
& (1,5,7,8,9),(1,5,7,8,9),(1,5,7,8,9),(2,4,6,8,10),(2,4,6,8,10),(2,4,6,8,10)
\end{aligned}
$$

where $1, \ldots, 5$ are the standard doses and $6, \ldots, 10$ are the test doses. The efficiency factor for the contrast $\boldsymbol{g}_{1}^{\prime} \boldsymbol{\tau}$ under this design is at least 0.9921 and the overall efficiency factor of the design is at least 0.9973 .

Case (ii). $m$ is even. Here the parameters of the design that we seek are $k=2 k_{1}+1, m=$ $2 m_{1}\left(\Rightarrow v=2 m=4 m_{1}\right), b=4 b_{1}, r=\frac{b k}{v}=\frac{4 b_{1}\left(2 k_{1}+1\right)}{4 m_{1}}=\frac{b_{1}\left(2 k_{1}+1\right)}{m_{1}}$.

As in Case $(i)$, we assume that $k_{1}$ is even. Now, let $d^{*}$ be a nearly linear trend-free design in $D\left(v^{*}=b, b^{*}=r, k^{*}=m, r_{1}^{*}=\cdots=r_{\frac{v^{*}}{2}}^{*}=k_{1}+1 ; r_{\frac{v^{*}+1}{2}}^{*}=\cdots=r_{v^{*}}^{*}=k_{1}\right)$. Note that a linear trend-free design cannot exist in this case, as $r_{i}^{*}\left(k^{*}+1\right) \neq 0(\bmod 2)$ for all $i, 1 \leq i \leq v^{*}$. Following the discussion in the last paragraph of Section 2, without loss of generality, we take a $d^{*}$, such that for $1 \leq i \leq \frac{1}{4} v^{*}, \sum_{j=1}^{b^{*}} \sum_{l=1}^{k^{*}} \delta_{j l}^{i} l=\frac{r_{1}^{*}\left(k^{*}+1\right)-1}{2}$, for $\frac{1}{4} v^{*}+1 \leq i \leq \frac{1}{2} v^{*}, \sum_{j=1}^{b^{*}} \sum_{l=1}^{k^{*}} \delta_{j l}^{i} l=\frac{r_{1}^{*}\left(k^{*}+1\right)+1}{2}$ and for the remaining $\frac{1}{2} v^{*}$ treatments, $\sum_{j=1}^{b^{*}} \sum_{l=1}^{k^{*}} \delta_{j l}^{i} l=\frac{r_{v^{*}}^{*}\left(k^{*}+1\right)}{2}=\frac{\left(r_{1}^{*}-1\right)\left(k^{*}+1\right)}{2}$.

From $d^{*}$, we obtain the matrices $N_{1 d_{0}}$ and $N_{2 d_{0}}$ leading to the proposed design $d_{0} \in$ $D\left(v=2 m, b=4 b_{1}, k=2 k_{1}+1, r=\frac{b_{1}\left(2 k_{1}+1\right)}{m_{1}}\right)$, with incidence matrix $N_{d_{0}}=\left[\begin{array}{l}N_{1 d_{0}} \\ N_{2 d_{0}}\end{array}\right]$,
where $N_{1 d_{0}}, N_{2 d_{0}}$ are obtained from $d^{*}$ exactly in the same manner as in Case (i). Then

$$
N_{d_{0}}=\left[\begin{array}{ll}
M_{1 d_{0}} & M_{2 d_{0}} \\
M_{2 d_{0}} & M_{1 d_{0}}
\end{array}\right]
$$

with

$$
\begin{array}{r}
\mathbf{1}_{m}^{\prime} M_{1 d_{0}}=\left(k_{1}+1\right) \mathbf{1}_{2 b_{1}}^{\prime}, \mathbf{1}_{m}^{\prime} M_{2 d_{0}}=k_{1} \mathbf{1}_{2 b_{1}}^{\prime} \\
\boldsymbol{w}^{\prime} M_{1 d_{0}}=\frac{1}{2}\left[\mathbf{1}_{b_{1}}^{\prime},-\mathbf{1}_{b_{1}}^{\prime}\right] \text { and } \boldsymbol{w}^{\prime} M_{2 d_{0}}=\mathbf{0}^{\prime} .
\end{array}
$$

For an arbitray design $d \in D(v=2 m, b, k, r)$, consider the matrix $G C_{d} G^{\prime}=r G G^{\prime}-$ $k^{-1} G N_{d} N_{d}{ }^{\prime} G^{\prime}$. Now,

$$
\begin{aligned}
G N_{d} & =\left[\begin{array}{rr}
(2 m)^{-1 / 2} \mathbf{1}_{m}^{\prime} & -(2 m)^{-1 / 2} \mathbf{1}_{m}^{\prime} \\
\left(m\left(m^{2}-1\right) / 6\right)^{-1 / 2} \boldsymbol{w}^{\prime} & \left(m\left(m^{2}-1\right) / 6\right)^{-1 / 2} \boldsymbol{w}^{\prime} \\
\left(m\left(m^{2}-1\right) / 6\right)^{-1 / 2} \boldsymbol{w}^{\prime} & -\left(m\left(m^{2}-1\right) / 6\right)^{-1 / 2} \boldsymbol{w}^{\prime}
\end{array}\right]\left[\begin{array}{l}
N_{1 d} \\
N_{2 d}
\end{array}\right] \\
& =\left[\begin{array}{r}
(2 m)^{-1 / 2}\left(\mathbf{1}_{m}^{\prime} N_{1 d}-\mathbf{1}_{m}^{\prime} N_{2 d}\right) \\
\left(2 m\left(m^{2}-1\right) / 3\right)^{-1 / 2}\left(\boldsymbol{f}^{\prime} N_{1 d}+\boldsymbol{f}^{\prime} N_{2 d}\right) \\
\left(2 m\left(m^{2}-1\right) / 3\right)^{-1 / 2}\left(\boldsymbol{f}^{\prime} N_{1 d}-\boldsymbol{f}^{\prime} N_{2 d}\right)
\end{array}\right]
\end{aligned}
$$

where $\boldsymbol{f}=2 \boldsymbol{w}$.
As before, $\mathbf{1}_{m}^{\prime} N_{1 d}=\left(a_{d 11}, \ldots, a_{d 1 b}\right), \mathbf{1}_{m}^{\prime} N_{d_{2}}=\left(a_{d 21}, \ldots, a_{d 2 b}\right)$, and let $\boldsymbol{f}^{\prime} N_{1 d}=\left(c_{d 11}, \ldots, c_{d 1 b}\right)$ and $\boldsymbol{f}^{\prime} N_{2 d}=\left(c_{d 21}, \ldots, c_{d 2 b}\right)$. Then, since the design is equireplicate, $\sum_{j=1}^{b} c_{d 1 j}=\sum_{j=1}^{b} c_{d 2 j}=0$. In order to maximize $\boldsymbol{g}_{i}^{\prime} C_{d} \boldsymbol{g}_{i}, i=1,2,3$ over $D(v=2 m, b, k, r)$, we need to minimize $\boldsymbol{g}_{i}^{\prime} N_{d} N_{d}^{\prime} \boldsymbol{g}_{i}$, since $\boldsymbol{g}_{i}^{\prime} \boldsymbol{g}_{i}=1$ is fixed. Now,

$$
\begin{aligned}
& \boldsymbol{g}_{1}^{\prime} N_{d} N_{d}^{\prime} \boldsymbol{g}_{1}=\sum_{j=1}^{b}(2 m)^{-1}\left(a_{d 1 j}-a_{d 2 j}\right)^{2} \\
& \boldsymbol{g}_{2}^{\prime} N_{d} N_{d}^{\prime} \boldsymbol{g}_{2}=\sum_{j=1}^{b}\left(2 m\left(m^{2}-1\right) / 3\right)^{-1}\left(c_{d 1 j}+c_{d 2 j}\right)^{2} \\
& \boldsymbol{g}_{3}^{\prime} N_{d} N_{d}^{\prime} \boldsymbol{g}_{3}=\sum_{j=1}^{b}\left(2 m\left(m^{2}-1\right) / 3\right)^{-1}\left(c_{d 1 j}-c_{d 2 j}\right)^{2} .
\end{aligned}
$$

To begin with, recall that as in Case (i), $\boldsymbol{g}_{1}^{\prime} N_{d} N_{d}^{\prime} \boldsymbol{g}_{1}$ is minimized when $\left|a_{d 1 j}-a_{d 2 j}\right|=1$ for all $j=1, \ldots, b$. Since $m$ is even, $\boldsymbol{f}^{\prime}=2 \boldsymbol{w}^{\prime}=(-(m-1),-(m-3), \ldots,-3,-1,1,3, \ldots, m-3, m-1)$ and $\boldsymbol{\alpha}^{\prime}=\boldsymbol{f}^{\prime}+\mathbf{1}^{\prime}=(-(m-2),-(m-4), \ldots,-4,-2,0,2,4, \ldots, m-4, m-2, m)$. We now show that for $1 \leq j \leq b, c_{d 1 j} \pm c_{d 2 j} \neq 0$. To see this, if possible, let $c_{d 1 j}+c_{d 2 j}=0$ for some $j$. Then, for this $j, c_{d 1 j}+c_{d 2 j}=\boldsymbol{f}^{\prime}\left(\boldsymbol{n}_{1 d j}+\boldsymbol{n}_{2 d j}\right)=0$, where $\boldsymbol{n}_{1 d j}\left(\boldsymbol{n}_{2 d j}\right)$ is the $j$ th column of $N_{1 d}\left(N_{2 d}\right)$. Also, $\mathbf{1}^{\prime}\left(\boldsymbol{n}_{1 d j}+\boldsymbol{n}_{2 d j}\right)=k=2 k_{1}+1$. Thus, $\boldsymbol{\alpha}^{\prime}\left(\boldsymbol{n}_{1 d j}+\boldsymbol{n}_{2 d j}\right)=\left(\boldsymbol{f}^{\prime}+\mathbf{1}^{\prime}\right)\left(\boldsymbol{n}_{1 d j}+\boldsymbol{n}_{2 d j}\right)=k$. But each element of $\boldsymbol{\alpha}$ is even and thus we have a contradiction. Again, if possible, let $c_{d 1 j}-c_{d 2 j}=0$ for some $j$. Then, $c_{d 1 j}-c_{d 2 j}=\boldsymbol{f}^{\prime}\left(\boldsymbol{n}_{1 d j}-\boldsymbol{n}_{2 d j}\right)=0$. Also, $\mathbf{1}^{\prime}\left(\boldsymbol{n}_{1 d j}-\boldsymbol{n}_{2 d j}\right)$ is an odd integer. Thus, $\boldsymbol{\alpha}^{\prime}\left(\boldsymbol{n}_{1 d j}-\boldsymbol{n}_{2 d j}\right)=\left(\boldsymbol{f}^{\prime}+\mathbf{1}^{\prime}\right)\left(\boldsymbol{n}_{1 d j}-\boldsymbol{n}_{2 d j}\right)$ is odd and the proof is complete.

The minimum of each of $\sum_{j=1}^{b}\left(c_{d 1 j}+c_{d 2 j}\right)^{2}$ and $\sum_{j=1}^{b}\left(c_{d 1 j}-c_{d 2 j}\right)^{2}$ is therefore attained when $c_{d 1 j}+c_{d 2 j}= \pm 1,1 \leq j \leq b$ and $c_{d 1 j}-c_{d 2 j}= \pm 1,1 \leq j \leq b$ respectively. With these values of $\left\{a_{d i j}\right\}, i=1,2$ and $\left\{c_{d 1 j} \pm c_{d 2 j}\right\}, j=1, \ldots, b$, the minimum of $\boldsymbol{g}_{i}^{\prime} N_{d} N_{d}^{\prime} \boldsymbol{g}_{i}, i=1,2,3$ are given respectively by

$$
\frac{b}{2 m}, \frac{3 b}{2 m\left(m^{2}-1\right)}, \frac{3 b}{2 m\left(m^{2}-1\right)} .
$$

Hence,

$$
\begin{gathered}
\max _{d \in D}\left(\boldsymbol{g}_{1}^{\prime} C_{d} \boldsymbol{g}_{1}\right)=r\left(1-k^{-2}\right), \\
\max _{d \in D}\left(\boldsymbol{g}_{2}^{\prime} C_{d} \boldsymbol{g}_{2}\right)=r\left(1-3 k^{-2}\left(m^{2}-1\right)^{-1}\right)=\max _{d \in D}\left(\boldsymbol{g}_{3}^{\prime} C_{d} \boldsymbol{g}_{3}\right),
\end{gathered}
$$

and these maximum values are attained by the proposed design $d_{0}$.
Therefore, $\sigma^{-2} \operatorname{Var}\left(\boldsymbol{g}_{1}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} \geq\left(r\left(1-k^{-2}\right)\right)^{-1}$ and for $i=2,3, \sigma^{-2} \operatorname{Var}\left(\boldsymbol{g}_{i}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} \geq\{r(1-$ $\left.\left.3 k^{-2}\left(m^{2}-1\right)^{-1}\right)\right\}^{-1}$. One can now obtain a lower bound to the efficiency factor of the contrasts $\boldsymbol{g}_{i}^{\prime} \boldsymbol{\tau}$ under a design $d$ as

$$
e_{1 d}=\sigma^{2} /\left\{\operatorname{Var}\left(\boldsymbol{g}_{1}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} r\left(1-k^{-2}\right)\right\}
$$

and

$$
e_{i d}=\sigma^{2} /\left\{\operatorname{Var}\left(\boldsymbol{g}_{i}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} r\left(1-3 k^{-2}\left(m^{2}-1\right)^{-1}\right)\right\}, i=2,3 .
$$

Also, a lower bound to a measure of the overall efficiency factor of a design $d$ is given by

$$
\bar{e}_{d}=\frac{\sigma^{2} k^{2}\left\{\left(m^{2}-1\right)\left(3 k^{2}-2\right)-3\right\}}{r\left(k^{2}-1\right)\left\{k^{2}\left(m^{2}-1\right)-3\right\}} / \sum_{1 \leq i \leq 3} \operatorname{Var}\left(\boldsymbol{g}_{i}^{\prime} \hat{\boldsymbol{\tau}}\right)_{d} .
$$

For the proposed class of nearly $L$-designs, we find in several appropriately chosen examples that $\bar{e}_{d_{0}}>0.95$.

Example 2. Let $m=6, k=9, b=8, r=6$. We find a nearly linear trend-free block design $d^{*} \in D\left(v^{*}=8, b^{*}=6=k^{*}, r_{1}^{*}=\cdots=r_{4}^{*}=5 ; r_{5}^{*}=\cdots=r_{8}^{*}=4\right)$. The design $d^{*}$, with columns as blocks is shown below.

$$
d^{*} \equiv \begin{array}{cccccc}
2 & 3 & 6 & 5 & 4 & 1 \\
3 & 1 & 5 & 7 & 2 & 4 \\
7 & 4 & 8 & 3 & 6 & 8 \\
1 & 6 & 2 & 8 & 8 & 7 \\
4 & 2 & 7 & 1 & 3 & 5 \\
6 & 5 & 4 & 2 & 1 & 3
\end{array}
$$

Following the method of construction of $d_{0}$ described above, we have

$$
\begin{aligned}
& N_{1 d_{0}}=\left[\begin{array}{llll|llll}
1 & 1 & 1 & 1 & 1 & 1 & 0 & 0 \\
1 & 1 & 1 & 1 & 1 & 0 & 1 & 0 \\
0 & 0 & 1 & 1 & 0 & 1 & 1 & 2 \\
1 & 1 & 0 & 0 & 0 & 1 & 1 & 2 \\
1 & 1 & 1 & 1 & 1 & 0 & 1 & 0 \\
1 & 1 & 1 & 1 & 1 & 1 & 0 & 0
\end{array}\right]=\left[M_{1 d_{0}} \mid M_{2 d_{0}}\right] \\
& N_{2 d_{0}}=\left[M_{2 d_{0}} \mid M_{1 d_{0}}\right] .
\end{aligned}
$$

Then, $N_{d_{0}}=\left[\begin{array}{l}N_{1 d_{0}} \\ N_{2 d_{0}}\end{array}\right]$ is the incidence matrix of the design $d_{0} \in D(12,8,9,6)$. For this design $\bar{e}_{d_{0}}=0.9994$.

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## Appendix

The following definition and lemmas are needed in the proof of Theorem 1.
Definition 3. Given a block size $k$, two plots $l$ and $l^{\prime}$ are said to be mirror-symmetric if $l+l^{\prime}=k+1$.

Let $S_{k}=\{1,2, \ldots, k\}, V_{1}=\{1,2, \ldots, t\}$ and $V_{2}=\{t+1, \ldots, v\}$.

Lemma 1. Suppose $k=2 h$ and $p=2 l+1,3 p<k$. Then there exists $p$ sets of size 3, say $X_{1}, X_{2}, \ldots, X_{p}$, such that
(i) $\bigcup_{i=1}^{p} X_{i}=S_{k} \backslash(\{1,2, \ldots,(k-3 p-1) / 2\} \cup\{k / 2+1\} \cup\{(k+3 p+1) / 2+1,(k+3 p+1) / 2+2, \ldots, k\})$;
(ii) $X_{i} \cap X_{j}=\emptyset$, for all $i \neq j$, where $\emptyset$ is the null set ;
(iii) For $1 \leq i \leq l, \sum_{x \in X_{i}} x=(3 k+4) / 2$ and for $l+1 \leq i \leq p, \sum_{x \in X_{i}} x=(3 k+2) / 2$.

Proof. Let $z=(k-3 p-1) / 2$. Define

$$
X_{i}=\{i+z, 3 l+3+i+z, 6 l+5-2 i+z\}, 1 \leq i \leq l,
$$

and

$$
X_{l+j}=\{l+j+z, 2 l+1+j+z, 6 l+6-2 j+z\}, 1 \leq j \leq l+1 .
$$

The results (i), (ii) and (iii) follow easily from the construction of $X_{i}$ 's.

Lemma $1^{\prime}$. Let $\tilde{X}_{i}=\left\{k+1-x \mid x \in X_{i}\right\}$, where $X_{i} ' s, 1 \leq i \leq p$, are constructed as in Lemma 1. Then
(i) $\bigcup_{i=1}^{p} \tilde{X}_{i}=S_{k} \backslash(\{1,2, \ldots,(k-3 p-1) / 2\} \cup\{k / 2\} \cup\{(k+3 p+1) / 2+1,(k+3 p+1) / 2+2, \ldots, k\})$;
(ii) $\tilde{X}_{i} \cap \tilde{X}_{j}=\emptyset$, for all $i \neq j$;
(iii) For $1 \leq i \leq l, \sum_{x \in \tilde{X}_{i}} x=(3 k+2) / 2$ and for $l+1 \leq i \leq p, \sum_{x \in \tilde{X}_{i}} x=(3 k+4) / 2$.

Lemma 2. Suppose $k=2 h$ and $p=2 l, 3 p \leq k$. Then there exists $p$ sets of size 3, say $Y_{1}, Y_{2}, \ldots, Y_{p}$, such that
(i) $\bigcup_{i=1}^{p} Y_{i}=S_{k} \backslash(\{1,2, \ldots,(k-3 p) / 2\} \cup\{(k+3 p) / 2+1,(k+3 p) / 2+2, \ldots, k\})$;
(ii) $Y_{i} \cap Y_{j}=\emptyset$, for all $i \neq j$;
(iii) For $1 \leq i \leq l, \sum_{y \in Y_{i}} y=(3 k+4) / 2$ and for $l+1 \leq i \leq p, \sum_{y \in Y_{i}} y=(3 k+2) / 2$.

Proof. Let $z=(k-6 l) / 2$. Define

$$
Y_{i}=\{i+z, 3 l+i+z, 6 l+2-2 i+z\}, 1 \leq i \leq l .
$$

and

$$
Y_{l+j}=\{l+j+z, 2 l+j+z, 6 l+1-2 j+z\}, 1 \leq j \leq l .
$$

The results (i), (ii) and (iii) follow easily from the construction of $Y_{i}$ 's.
Lemma 3. (Yeh et al. (1985)). Suppose $k$ is even. Then there exists a nearly linear trend-free block design $d \in D(k, 3, k, 3)$.

Proof. The three blocks of $d$ can be constructed as follows:
block 1: $(1,2, \ldots, k / 2, k / 2+1, k / 2+2, \ldots, k-1, k)$;
block 2: $(k, k-2, \ldots, 4,2, k-1, k-3, \ldots, 3,1)$;
block 3: $(k-1, k-3, \ldots, 3,1, k, k-2, \ldots, 4,2)$.
For $1 \leq i \leq k / 2, \sum_{j=1 l=1}^{3} \sum_{j l}^{k} \delta_{j l}^{2 i-1} \cdot l=(3 k+2) / 2$ and $\sum_{j=1 l=1}^{3} \sum_{j l}^{k} \delta_{j l}^{2 i} \cdot l=(3 k+4) / 2$. Hence $d$ is a nearly linear trend-free block design.

Next, we will prove that a nearly linear trend-free block design $d \in$ $D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$ exists when $k$ is even and at least one of $r_{i}$ 's is odd. Suppose $k$ is even and at least one of $r_{i}$ 's is odd. First, in Lemma 4, we handle the basic case of $r_{i}$ equal to 2 or 3 . Then in Theorem 1, we consider the general case $r_{i} \geq 2$.

The key idea for the proof of Lemma 4 is the following. Suppose $r_{i}=3, i \in V_{1}$, and $r_{j}=2, j \in V_{2}$.

Step I : With the help of Lemmas 1, 2 and 3, we can identify and fill those three proper plots with three replications of treatment $i, i \in V_{1}$, in an un-filled $k \times b$ array $d$ such that (i) $\sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l=(3 k+2) / 2$ or $(3 k+4) / 2,1 \leq i \leq t$; (ii) All remaining un-filled plots in $d$ are mirror-symmetric in pairs, i.e., if a plot $l$ in block $j$ is unfilled, then there always exists another unfilled plot $k+1-l$ in some block $j^{\prime}$.

Step II : Fill each pair of mirror-symmetric plots with two replications of treatment $j, j \in V_{2}$. From the property of the mirror-symmetric plots, we get $\sum_{j=1 l=1}^{b} \sum_{j l}^{k} \delta_{j l}^{i} l=r_{i}(k+1) / 2=$ $2(k+1) / 2, t+1 \leq i \leq v$. Therefore, the filled array $d$ is a nearly linear trend-free block design.

Lemma 4. Suppose $k$ is even, $r_{i}=3, i \in V_{1}$, and $r_{j}=2, j \in V_{2}$. Then a nearly linear trend-free block design $d \in D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$ exists.

Proof. Note that $t$ is even since $t=b k-2 v$. Let $t=p k+q, 0 \leq q \leq k-1$ and $b=3 p+b_{1}$. Consequently $q$ is even. Our desired nearly linear trend-free block design will be constructed as

$$
d=\left[\begin{array}{ccccc:c} 
& \vdots & \vdots & & \vdots & \vdots \\
d_{1} & d_{2} & \ldots & d_{p} & d_{p+1} & \\
& \vdots & \vdots & & \vdots & \\
& & & & \\
d_{p+2}
\end{array}\right],
$$

where (i) for $1 \leq i \leq p, d_{i} \in D(k, 3, k, 3)$ is a nearly linear trend-free block design consisting of treatments $(i-1) k+1,(i-1) k+2, \ldots, i k$, constructed from Lemma 3; (ii) $d_{p+1}$ is a $k \times h$ array
and can be written as $\left[\begin{array}{c}d_{q_{1}} \\ d_{q} \\ d_{q_{2}}\end{array}\right]$, where $h$ could be 3, 2, 1 depending on cases. Furthermore in $d_{p+1}, d_{q}$ occupies the middle $\rho$ rows ( $\rho$ could be $q, 3 q / 2$ or $3 q$ depending on cases), contains treatments $p k+1, p k+2, \ldots, p k+q$ and maybe some treatments from $V_{2}, d_{q_{1}}$ and $d_{q_{2}}$ occupies the first $(k-\rho) / 2$ rows and the last $(k-\rho) / 2$ rows respectively, containing treatments from $V_{2}$ only; (iii) $d_{p+2}$ has $b_{1}-h$ blocks containing treatments from $V_{2}$ only. Our goal is to construct proper $d_{p+1}$ and $d_{p+2}$ to make $d$ a nearly linear trend-free block design. We divide the proof into three cases.

Case 1. $\quad b_{1} \geq 3$. Here $h=3$.
From Lemma 3, we can construct a nearly linear trend-free block design $d_{q} \in D(q, 3, q, 3)$ consisting of treatments $p k+1, p k+2, \ldots, p k+q$. Observe that all un-filled plots in $d_{q_{1}}, d_{q_{2}}$ and $d_{p+2}$ are mirror-symmetric in pairs. Hence, fill each pair of mirror-symmetric plots with two replications of treatment $j, j \in V_{2}$. From the property of mirror-symmetric plots, we get $\sum_{j=1 l=1}^{b} \sum_{j l}^{k} \delta_{j l}^{i} l=r_{i}(k+1) / 2=2(k+1) / 2, t+1 \leq i \leq v$. The resulting design $d$ is our desired nearly linear trend-free block design.

Case 2. $b_{1}=2$. Here $h=2$ and $d_{p+2}$ vanishes. Let $q=2 m$. Write $d_{q}$ as

$$
\left[\begin{array}{cc} 
& \vdots \\
d_{q}^{m_{1}} & d_{q}^{m_{2}} \\
& \vdots
\end{array}\right]
$$

where $d_{q}^{m_{i}}, 1 \leq i \leq 2$, is a $3 m \times 1$ vector consisting of treatments $p k+(i-1) m+1, p k+(i-$ 1) $m+2, \ldots, p k+i m$.
(a) $m$ is odd. $d_{q}^{m_{1}}$ is constructed by inserting three replications of treatment $p k+j$ into the $(x)_{t h}$ plot of an un-filled $3 m \times 1$ vector, where $x \in X_{j}, 1 \leq j \leq m$ and $X_{j}$ 's are obtained from Lemma 1 by letting $p=m$ and $k=3 m+1$. $d_{q}^{m_{2}}$ is constructed by inserting three replications of treatment $p k+m+j$ into the $(x)_{t h}$ plot of an un-filled $3 m \times 1$ vector, where $x \in \tilde{X}_{j}, 1 \leq j \leq m$ and $\tilde{X}_{j}$ 's are obtained from Lemma $1^{\prime}$ by letting $p=m$ and $k=3 m+1$.
(b) $m$ is even. $d_{q}^{m_{i}}$ is constructed by placing three replications of treatment $p k+(i-1) m+j$ into the $(y)_{t h}$ plot of an un-filled $3 m \times 1$ vector, where $y \in Y_{j}, 1 \leq i \leq 2,1 \leq j \leq m$ and $Y_{j}$ 's are obtained from Lemma 2 by letting $p=m$ and $k=3 m$. Check the remaining unfilled plots in $d_{p+1}$, they are mirror-symmetric in pairs. Hence the desired design is constructed.

Case 3. $b_{1}=1$. Here $h=1$ and $d_{p+2}$ vanishes. $d_{q}$ is a $3 q \times 1$ vector consisting of treatments $p k+1, p k+2, \ldots, p k+q$. Now, let $d_{q}$ act as the $d_{q}^{m_{1}}$ of Case 2(b), then the proof follows along the lines of Case 2(b).

The resulting designs in the above cases are nearly linear trend-free but not necessarily
connected. Hence, we have to horizontally shift the positions of the treatments among $d_{i}$ 's to make $d$ a connected design. That completes the proof.

Proof of Theorem 1. Without loss of generality, we assume $r_{i}$ is odd, $1 \leq i \leq t$ and $r_{i}$ is even, $t+1 \leq i \leq v$. For $1 \leq i \leq t$, we can write $r_{i}=3+2 r_{1 i}$. Then $r_{i}$ replications of treatment $i$ can be renamed as 3 replications of one new treatment plus 2 replications of $r_{1 i}$ new treatments. For $t+1 \leq i \leq v$, we can write $r_{i}=2 r_{1 i}$. Then $r_{i}$ replications of treatment $i$ can be renamed as 2 replications of $r_{1 i}$ new treatments. Let $v^{*}=t+\sum_{i=1}^{v} r_{1 i}$. In other words, a design belonging to $D\left(v, b, k, r_{1}, r_{2}, \ldots, r_{v}\right)$ can be renamed as another design belonging to $D\left(v^{*}, b, k, r_{1}^{*}, \ldots, r_{t}^{*}, r_{t+1}^{*}, \ldots, r_{v^{*}}^{*}\right)$ with $r_{i}^{*}=3,1 \leq i \leq t$ and $r_{i}^{*}=2, t+1 \leq i \leq v^{*}$. From Lemma 4, we know a nearly linear trend-free block design $d^{*} \in D\left(v^{*}, b, k, r_{1}^{*}, \ldots, r_{v^{*}}^{*}\right)$ exists. In $d^{*}$, revert new treatments back to the original treatments resulting into our required d. Obviously, $d \in D\left(v, b, k, r_{1}, \ldots, r_{v}\right)$ and for $1 \leq i \leq t, \quad \sum_{j=1 l=1}^{b} \sum_{j l}^{k} \delta_{l}^{i} l=((3(k+1)-1) / 2$ or $(3(k+1)+1) / 2)+r_{1 i} \cdot 2(k+1) / 2=\left(r_{i}(k+1)-1\right) / 2$ or $\left(r_{i}(k+1)+1\right) / 2$ and for $t+1 \leq i \leq v, \quad \sum_{j=1}^{b} \sum_{l=1}^{k} \delta_{j l}^{i} l=r_{1 i} \cdot 2(k+1) / 2=r_{i}(k+1) / 2$. Hence, $d$ is a nearly linear trend-free block design.

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