

Optimal Crossover Designs
for Comparing Test Treatments to a Control Treatment
When Subject Effects are Random

A.S. Hedayat and Wei Zheng

Department of Mathematics, Statistics, and Computer Science
University of Illinois at Chicago

- A Taste of Optimal Designs
- Motivation from Statistics
 - Background
 - Results
- Construction of Optimal Designs
 - Characteristics of Optimal Designs
 - Guidelines for Construction
 - Methods of Constructions
- Further Problems

A design as a mapping

1	2	3	0	0	2	0	3	2	1
0	3	0	1	4	0	1	2	3	4
2	0	3	3	0	4	4	2	4	0

$d : (k, u) \mapsto i$ where $1 \leq k \leq p$, $1 \leq u \leq n$ and $0 \leq i \leq t$

In this example, $n = 10$, $p = 3$, $t = 4$.

What's special about these two designs?

```
0 0 0 2 3 1 2 3 1
1 2 3 0 0 0 1 2 3
2 3 1 1 2 3 0 0 0
```

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1 3 2 3 2 0 1 3 0 2 1 0
2 1 3 2 0 3 3 0 1 1 0 2
0 0 0 1 1 1 2 2 2 3 3 3
```

Notations

- $n_{diu} = \sum_{k=1}^p I_{[d(k,u)=i]}$.
- $\tilde{n}_{diu} = \sum_{k=1}^{p-1} I_{[d(k,u)=i]}$.
- $l_{dik} = \sum_{u=1}^n I_{[d(k,u)=i]}$.
- $m_{dij} = \sum_{u=1}^n \sum_{k=1}^{p-1} I_{[d(k,u)=i, d(k+1,u)=j]}$.
- $r_{di} = \sum_{u=1}^n \sum_{k=1}^p I_{[d(k,u)=i]}$.
- $\tilde{r}_{di} = \sum_{u=1}^n \sum_{k=1}^{p-1} I_{[d(k,u)=i]}$.

In general, we need...

A design d is said to be a totally balanced test-control incomplete crossover design (TBTCI) if:

- 1 Each element from $\{1, 2, \dots, t\}$ show up in each column at most once.
- 2 Each element from $\{0, 1, \dots, t\}$ is equally replicated in each row.
- 3 $|n_{d0u} - n_{d0v}| \leq 1$ and $|\tilde{n}_{d0u} - \tilde{n}_{d0v}| \leq 1$ for all $1 \leq u, v \leq n$.
- 4 m_{d0i} , m_{di0} and m_{dij} are constants across all $1 \leq i \neq j \leq t$ and $m_{dij} = 0$ for all $0 \leq i \leq t$.
- 5 r_{di} and \tilde{r}_{di} are constants across all $1 \leq i \leq t$.
- 6 $\sum_{u=1}^n n_{d0u} n_{diu}$, $\sum_{u=1}^n n_{diu} n_{dju}$, $\sum_{u=1}^n \tilde{n}_{d0u} \tilde{n}_{diu}$, $\sum_{u=1}^n \tilde{n}_{diu} \tilde{n}_{dju}$, $\sum_{u=1}^n n_{d0u} \tilde{n}_{diu}$, $\sum_{u=1}^n \tilde{n}_{d0u} n_{diu}$, and $\sum_{u=1}^n n_{diu} \tilde{n}_{dju}$, are constants across all $1 \leq i \neq j \leq t$.

Let $N_d = (n_{diu})$ and $\tilde{N}_d = (\tilde{n}_{diu})$ when $0 \leq i \leq t$ and $1 \leq u \leq n$.
 Conditions 5 and 6 are equivalent to

$$N_d N'_d = \begin{pmatrix} a_1 & b_1 1'_t \\ b_1 1_t & (e_1 - f_1) I_t + f_1 J_t \end{pmatrix} \quad (1)$$

$$N_d \tilde{N}'_d = \begin{pmatrix} a_2 & b_2 1'_t \\ c_2 1_t & (e_2 - f_2) I_t + f_2 J_t \end{pmatrix} \quad (2)$$

$$\tilde{N}_d \tilde{N}'_d = \begin{pmatrix} a_3 & b_3 1'_t \\ b_3 1_t & (e_3 - f_3) I_t + f_3 J_t \end{pmatrix} \quad (3)$$

$$\begin{aligned} a_1 &= \sum_{u=1}^n n_{d0u}^2 \\ b_1 &= \sum_{u=1}^n n_{d0u} n_{d1u} \\ e_1 &= \sum_{u=1}^n n_{d1u}^2 \\ f_1 &= \sum_{u=1}^n n_{d1u} n_{d2u} \end{aligned}$$

$$\begin{aligned} a_2 &= \sum_{u=1}^n n_{d0u} \tilde{n}_{d0u} \\ b_2 &= \sum_{u=1}^n n_{d0u} \tilde{n}_{d1u} \\ c_2 &= \sum_{u=1}^n \tilde{n}_{d0u} n_{d1u} \\ e_2 &= \sum_{u=1}^n n_{d1u} \tilde{n}_{d1u} \\ f_2 &= \sum_{u=1}^n n_{d1u} \tilde{n}_{d2u} \end{aligned}$$

$$\begin{aligned} a_3 &= \sum_{u=1}^n \tilde{n}_{d0u}^2 \\ b_3 &= \sum_{u=1}^n \tilde{n}_{d0u} \tilde{n}_{d1u} \\ e_3 &= \sum_{u=1}^n \tilde{n}_{d1u}^2 \\ f_3 &= \sum_{u=1}^n \tilde{n}_{d1u} \tilde{n}_{d2u} \end{aligned}$$

Definition

A $p \times n$ array with symbols from $\{0, 1, 2, \dots, t\}$ is said to be a crossover design if columns represent subjects, rows represent periods and symbols represent treatments.

- Our goal
Compare the test treatments, $\{1, 2, \dots, t\}$, with the control treatment $\{0\}$.
- Important notations
 - n : number of subjects/units/patients
 - p : number of periods
 - t : number of test treatments
 - r_{d0} : replications of the control treatment in design d .

An Example

	1	2	3	0	0	2	0	3	2	1
d:	0	3	0	1	4	0	1	2	3	4
	2	0	3	3	0	4	4	2	4	0

An Example

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d:	0	3	0	1	4	0	1	2	3	4
	2	0	3	3	0	4	4	2	4	0

$$n = 10, \quad p = 3, \quad t = 4 \quad r_{d0} = 9$$

An Example

$$\begin{array}{cccccccccc} & 1 & 2 & 3 & 0 & 0 & 2 & 0 & 3 & 2 & 1 \\ \text{d:} & 0 & 3 & 0 & 1 & 4 & 0 & 1 & 2 & 3 & 4 \\ & 2 & 0 & 3 & 3 & 0 & 4 & 4 & 2 & 4 & 0 \end{array}$$

$$n = 10, \quad p = 3, \quad t = 4 \quad r_{d0} = 9$$

- n could be hundreds or thousands depending on the study.
- p is usually not large due to ethic or other issues.
- t is not large either; we will investigate $t + 1 \geq p \geq 3$.

An Example

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d:	0	3	0	1	4	0	1	2	3	4
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- n could be hundreds or thousands depending on the study.
- p is usually not large due to ethic or other issues.
- t is not large either; we will investigate $t + 1 \geq p \geq 3$.
- # of designs (identical up to an isomorphism):

$$\frac{(N+n-1)!}{t!n!(N-1)!} \geq \frac{1}{t!(N-1)!} n^{N-1}, \quad N = (t+1)^p.$$
- Isomorphism: in the sense of relabelling the subjects and test treatments.

Model

$$Y_{dku} = \mu + \alpha_k + \beta_u + \tau_{d(k,u)} + \gamma_{d(k-1,u)} + \epsilon_{ku} \quad (4)$$
$$\beta_u \text{ iid } N(0, \sigma_\beta^2), \quad \epsilon_{ku} \text{ iid } N(0, \sigma^2), \quad \beta_u \perp \epsilon_{ku}$$

- Y_{dku} : Response from unit (subject) u in period k in design d .
- α_k : Effect of period k .
- β_u : Effect of subject u .
- $d(k, u)$: Treatment specified by the design d for unit u in period k . (Control $\{0\}$; Test $\{1, 2, \dots, t\}$)
- τ_i : Direct effect of treatment i
- γ_i : Carryover effect of treatment i (by convention $\gamma_{d(0,u)} = 0$)

Model (In Matrix Form)

$$\begin{aligned} E(\mathbf{Y}_d) &= \mathbf{1}_{np}\mu + P\boldsymbol{\alpha} + T_d\boldsymbol{\tau} + F_d\boldsymbol{\gamma} \\ \text{var}(\mathbf{Y}_d) &= \sigma^2(I_n \otimes (I_p + \theta J_p)) \end{aligned} \quad (5)$$

Where

- $\theta = \sigma_{\beta}^2 / \sigma^2$.
- $\boldsymbol{\alpha} = (\alpha_1, \dots, \alpha_p)'$, $\boldsymbol{\tau} = (\tau_0, \dots, \tau_t)'$, $\boldsymbol{\gamma} = (\gamma_0, \dots, \gamma_t)'$.
- $P = \mathbf{1}_n \otimes I_p$.
- T_d and F_d denote the treatment and carryover incidence matrices.
- \otimes denote the Kronecker product.

The information matrix C_d for τ is

$$C_d = T_d' V^{-1/2} pr^\perp (V^{-1/2} [1_{np} | P | F_d]) V^{-1/2} T_d \quad (6)$$

where $V = I_n \otimes (I_p + \theta J_p)$ which depends on θ only,
and $pr^\perp A = I - A(A'A)^- A'$ is a projection.

- If $\theta = \infty$ (Hedayat and Yang (2005))
 C_d becomes the information matrix for the model with fixed subject effects (β_u is nonrandom.)
- If $\theta = 0$
 C_d becomes the information matrix for the model without subject effects ($\beta_u \equiv 0$)

The information matrix for $(\tau_1 - \tau_0, \tau_2 - \tau_0, \dots, \tau_t - \tau_0)'$ is

$$M_d = T' C_d T \quad \text{where} \quad T = \begin{pmatrix} 0_{1 \times t} \\ I_{t \times t} \end{pmatrix} \quad (7)$$

Thus, M_d can be simply obtained from C_d by deleting the first row and the first column of C_d .

- A-Optimal: $\min_d \sum_{i=1}^t \text{Var}(\widehat{\tau_i - \tau_0})$ (i.e. $\min_d \text{Tr}(M_d^{-1})$)

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- A-Optimal: $\min_d \sum_{i=1}^t \text{Var}(\widehat{\tau_i - \tau_0})$ (i.e. $\min_d \text{Tr}(M_d^{-1})$)
- MV-Optimal: $\min_d \max_{1 \leq i \leq t} \text{Var}(\widehat{\tau_i - \tau_0})$

Lemma

An A-optimal design is also an MV-optimal design if its information matrix, M_d , is a completely symmetric matrix.

How to find $d^* = \operatorname{argmin}_d \operatorname{Tr}(M_d^{-1})$

$$\operatorname{Tr}(M_d^{-1}) \geq B_1(d) \geq B_2(d) \geq B_3(d) \dots \geq B_m(d) \geq B_0$$

- ① $B_i, i \geq 1$ are functions of d ; B_0 is a constant depending on n, p, t, θ .
- ② Each inequality should hold for every competing design d .
- ③ There should exist a design d^* with all the equalities hold, i.e.

$$\operatorname{Tr}(M_{d^*}^{-1}) = B_0$$

Then we have $\operatorname{Tr}(M_d^{-1}) \geq \operatorname{Tr}(M_{d^*}^{-1})$.

Note that $\operatorname{Tr}(M_d^{-1})$ is essentially a complicated function of the variables $n_{diu}, \tilde{n}_{diu}, l_{dik}, m_{dij}, r_{di}$ and \tilde{r}_{di} for $0 \leq i \neq j \leq t$ and $1 \leq k \leq p$.

The first step:

$$C_d = T_d' V^{-1/2} pr^\perp(V^{-1/2} [1_{np} | P | F_d]) V^{-1/2} T_d$$

$$\leq T_d' V^{-1/2} pr^\perp(1_{np} | V^{-1/2} F_d) V^{-1/2} T_d \quad (8)$$

$$= T_d' V^{-1/2} pr^\perp(1_{np}) V^{-1/2} T_d \quad (9)$$

$$- T_d' V^{-1/2} pr^\perp(1_{np}) V^{-1/2} F_d$$

$$\times (F_d' V^{-1/2} pr^\perp(1_{np}) V^{-1/2} F_d)^{-}$$

$$\times F_d' V^{-1/2} pr^\perp(1_{np}) V^{-1/2} T_d$$

In (8), $A \leq B$ means $B - A$ is n.n.d. and the equality holds when $l_{dik} = r_{di}/p, i = 0, 1, \dots, t$

Note that the matrix under the operator pr^\perp becomes easy to evaluate.

The equality (9) uses the following fact:

$$pr^\perp([A|B]) = pr^\perp(A) - pr^\perp(A)B(B'pr^\perp(A)B)B'pr^\perp(A)$$

A middle step: $Tr(M_d^{-1}) \geq \frac{t(t-1)^2}{x_0} + \frac{t}{y_0}$ where $x_0 = \alpha - \frac{\beta^2}{\gamma}$ with

$$\alpha = \frac{1 + \theta p - \theta}{1 + \theta p} (tnp - tr_{d0}) - \frac{t \sum_{i=1}^t r_{di}^2 - r_{d0}^2}{(1 + \theta p)pn} - r_{d0} + \frac{\theta \sum_{u=1}^n n_{d0u}^2}{1 + \theta p}$$

$$\beta = t \sum_{i=1}^t m_{dii} + \frac{r_{d0}}{p} - l_{d01} - m_{d00} - \frac{\theta t}{1 + \theta p} \sum_{i=1}^t \sum_{u=1}^n n_{diu} \tilde{n}_{diu}$$

$$- \frac{t}{(1 + \theta p)pn} \sum_{i=1}^t r_{di} \tilde{r}_{di} + \frac{\theta}{1 + \theta p} \sum_{u=1}^n n_{d0u} \tilde{n}_{d0u} + \frac{r_{d0} \tilde{r}_{d0}}{(1 + \theta p)pn}$$

$$\gamma = (t + 1 - \frac{2}{p} - \frac{\theta t}{1 + \theta p})(n(p - 1) - \tilde{r}_{d0}) - \frac{n}{p}(p - 1)^2 - \frac{t}{(1 + \theta p)pn} \sum_{i=1}^t \tilde{r}_{di}^2$$

$$+ \frac{\tilde{r}_{d0}^2}{(1 + \theta p)pn} + \frac{\theta}{1 + \theta p} \sum_{u=1}^n \tilde{n}_{d0u}^2$$

and

$$\begin{aligned}
 y_0 = & \left(r_{d0} - \frac{\theta}{1 + \theta p} \sum_{u=1}^n n_{d0u}^2 - \frac{r_{d0}^2}{(1 + \theta p)pn} \right) \\
 & - \left\{ (n(p-1) - \tilde{r}_{d0}) \left(m_{d00} - \frac{\theta}{1 + \theta p} \sum_{u=1}^n n_{d0u} \tilde{n}_{d0u} - \frac{1}{(1 + \theta p)pn} r_{d0} \tilde{r}_{d0} \right) \right. \\
 & \left. + \tilde{r}_{d0} \left(\frac{r_{d0}}{p} - l_{d01} - m_{d00} + \frac{\theta}{1 + \theta p} \sum_{u=1}^n n_{d0u} \tilde{n}_{d0u} + \frac{1}{(1 + \theta p)pn} r_{d0} \tilde{r}_{d0} \right) \right\}^2 \\
 & \times \left\{ n(p-1) \left(\tilde{r}_{d0} - \frac{\theta}{1 + \theta p} \sum_{u=1}^n \tilde{n}_{d0u}^2 - \frac{\tilde{r}_{d0}^2}{(1 + \theta p)pn} \right) - \frac{\tilde{r}_{d0}^2}{p} \right\}^{-1}.
 \end{aligned}$$

- $\Omega_{n,p,t}$: The collection of all the designs with the number of subjects n , number of periods p , number of test treatments t .
- $\Lambda_{n,p,t}$: A subclass of $\Omega_{n,p,t}$ with the restrictions that the control treatment is equally replicated in each period and no treatment is immediately preceded by itself. ($l_{d0k} = r_{d0}/p$ and $m_{dii} = 0$ for all $1 \leq k \leq p$ and $0 \leq i \leq t$)

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Lemma

When $t \geq 3$ and $t + 1 \geq p \geq 3$, $\text{Tr}(M_d^{-1}) \geq B_m(d) = f(n, p, t, r_{d0}, \theta)$ for all designs in $\Lambda_{n,p,t}$. The equality is obtained by a design in a form of TBTCI. When $p = 3, t = 2$, the conclusion still holds but only within a subclass of $\Lambda_{n,p,t}$ in which $r_{d0}/n \geq 0.6306$.

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There is no closed form for $\operatorname{argmin}_r f(n, p, t, r, \theta)$.

$f(n, p, t, r_{d0}, \theta) = t(t-1)^2/\tilde{x}_0 + t/\tilde{y}_0$ where \tilde{x}_0 and \tilde{y}_0 are derived from x_0 and y_0 by replacing all of the variables therein related to d with functions of r_{d0} .

Theorem

When $t \geq 3$ and $t + 1 \geq p \geq 3$, a design d^ is optimal among designs in $\Lambda_{n,p,t}$ if it is a TBTCL and r_{d^*0} minimizes $f(n, p, t, r_{d0}, \theta)$ given n, p, t, θ . When $p = 3, t = 2$, the design d^* is optimal in the same sense as in the lemma.*

Remark: Similar results can be found in Hedayat and Yang (2005) when $\theta = \infty$. We extend the result for any value of $\theta \geq 0$.

Examples of TBTCI designs

- $TBTCI(9, 3, 3, 9)$

```

0 0 0 2 3 1 2 3 1
1 2 3 0 0 0 1 2 3
2 3 1 1 2 3 0 0 0
    
```

- $TBTCI(12, 3, 3, 9)$

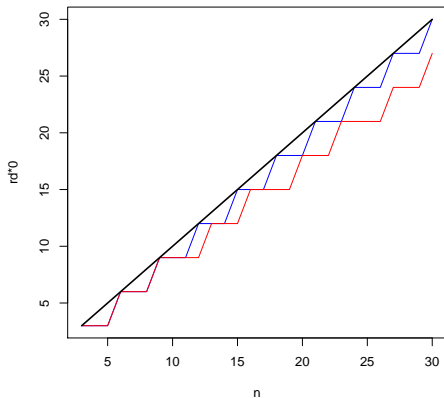
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1 3 2 3 2 0 1 3 0 2 1 0
2 1 3 2 0 3 3 0 1 1 0 2
0 0 0 1 1 1 2 2 2 3 3 3
    
```

Remark: $TBTCI(n, p, t, r_{d0})$ denotes a TBTCI with n units, p periods, t test treatments and r_{d0} replications of the control treatment.

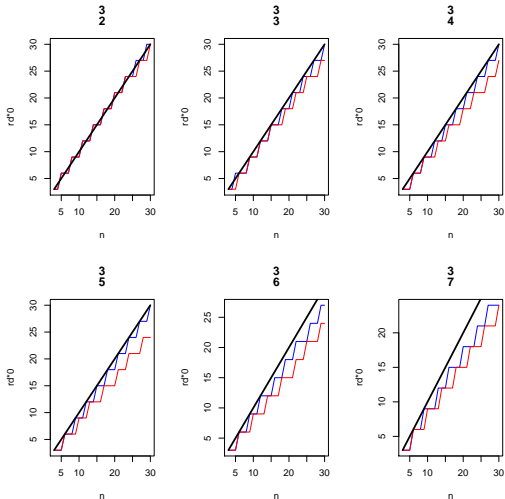
Graphical Description: $p = 3$ and $t = 4$

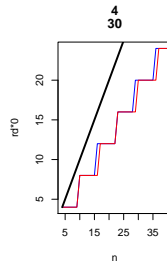
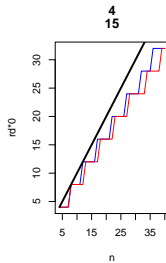
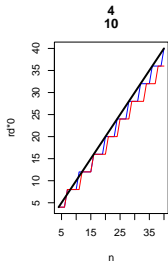
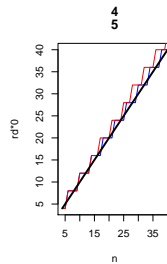
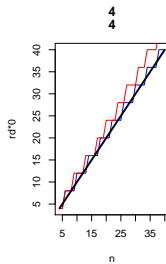
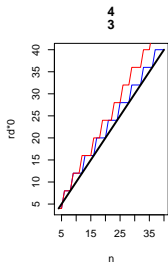
The optimal r_{d^*0} as a function of n in the sense of minimizing $f(n, p, t, r_{d0}, \theta)$.

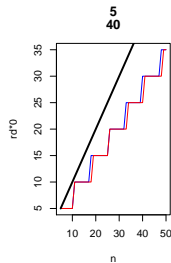
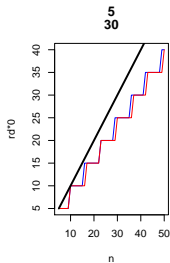
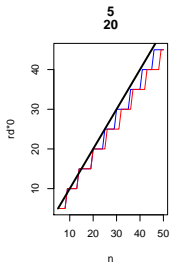
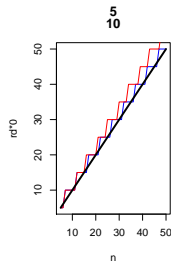
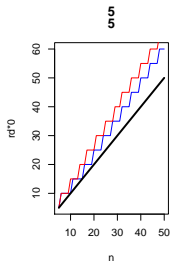
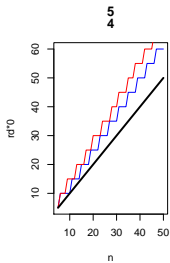


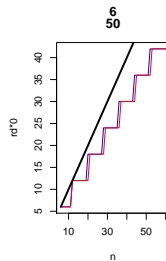
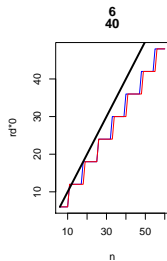
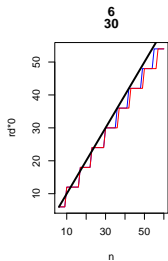
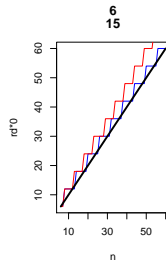
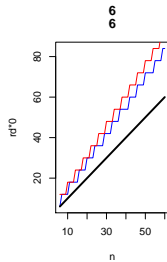
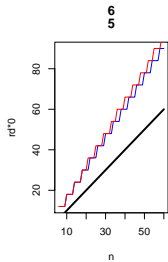
- $\theta \geq 0$ is unknown but predetermined.
- The curves correspond to $\theta = \infty$ and $\theta = 0$.
- The bold line represents the equation $r_{d^*0} = n$.
- Whenever the curve for θ crosses the bold line, we have $r_{d^*0} = n$.
- r_{d^*0} is slightly smaller than n in general.
- r_{d^*0} jumps by $p = 3$ each time for any value of θ .

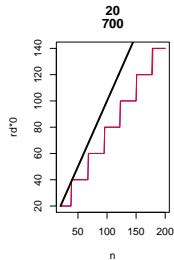
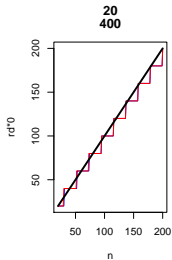
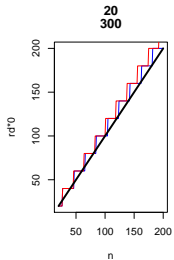
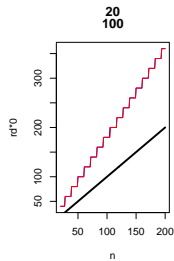
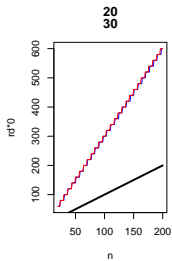
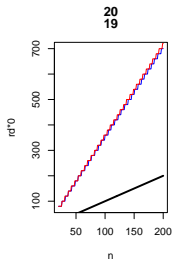
$p=3$ and $t=2,3,\dots,7$











Revisit the Notations

- $n_{diu} = \sum_{k=1}^p I_{[d(k,u)=i]}$.
- $\tilde{n}_{diu} = \sum_{k=1}^{p-1} I_{[d(k,u)=i]}$.
- $l_{dik} = \sum_{u=1}^n I_{[d(k,u)=i]}$.
- $m_{dij} = \sum_{u=1}^n \sum_{k=1}^{p-1} I_{[d(k,u)=i, d(k+1,u)=j]}$.
- $r_{di} = \sum_{u=1}^n \sum_{k=1}^p I_{[d(k,u)=i]}$.
- $\tilde{r}_{di} = \sum_{u=1}^n \sum_{k=1}^{p-1} I_{[d(k,u)=i]}$.

Revisit the Definition

A design d is said to be a totally balanced test-control incomplete crossover design (TBTCI) if:

- 1 Each element from $\{1, 2, \dots, t\}$ show up in each column at most once.
- 2 Each element from $\{0, 1, \dots, t\}$ is equally replicated in each row.
- 3 $|n_{d0u} - n_{d0v}| \leq 1$ and $|\tilde{n}_{d0u} - \tilde{n}_{d0v}| \leq 1$ for all $1 \leq u, v \leq n$.
- 4 m_{d0i} , m_{di0} and m_{dij} are constants across all $1 \leq i \neq j \leq t$ and $m_{dij} = 0$ for all $0 \leq i \leq t$.
- 5 r_{di} and \tilde{r}_{di} are constants across all $1 \leq i \leq t$.
- 6 $\sum_{u=1}^n n_{d0u} n_{diu}$, $\sum_{u=1}^n n_{diu} n_{dju}$, $\sum_{u=1}^n \tilde{n}_{d0u} \tilde{n}_{diu}$, $\sum_{u=1}^n \tilde{n}_{diu} \tilde{n}_{dju}$,
 $\sum_{u=1}^n n_{d0u} \tilde{n}_{diu}$, $\sum_{u=1}^n \tilde{n}_{d0u} n_{diu}$, and $\sum_{u=1}^n n_{diu} \tilde{n}_{dju}$, are constants across all $1 \leq i \neq j \leq t$.

Revisit (Continued)

Let $N_d = (n_{diu})$ and $\tilde{N}_d = (\tilde{n}_{diu})$ with the dimension of $0 \leq i \leq t$ and $1 \leq u \leq n$. Conditions 5 and 6 are equivalent to

$$N_d N_d' = \begin{pmatrix} a_1 & b_1 1_t' \\ b_1 1_t & (d_1 - e_1)I_t + e_1 J_t \end{pmatrix} \quad (10)$$

$$N_d \tilde{N}_d' = \begin{pmatrix} a_2 & b_2 1_t' \\ c_2 1_t & (d_2 - e_2)I_t + e_2 J_t \end{pmatrix} \quad (11)$$

$$\tilde{N}_d \tilde{N}_d' = \begin{pmatrix} a_3 & b_3 1_t' \\ b_3 1_t & (d_3 - e_3)I_t + e_3 J_t \end{pmatrix} \quad (12)$$

For $r_{d0} < n, p = 3$

Definition

A type I orthogonal array $OA_I(n, k, s, t)$ is a $k \times n$ array based on s symbols, where the columns of any $t \times n$ subarray contains all $s!/(s-t)!$ permutations of t distinct symbols.

Theorem

A type I orthogonal array $OA_I(t(t+1), 3, t+1, 2)$ and a $TBTCI(t(t+1), 3, t, 3t)$ coexists.

Given an $OA_I(t(t+1), 3, t+1, 2)$ with symbols from $\{0, 1, \dots, t\}$, label the rows as periods, columns as units and symbols as treatments, then by definition, this OA_I is a $TBTCI(t(t+1), 3, t, 3t)$.

A Latin square of order $t + 1$ with entries from $\{0, 1, 2, \dots, t\}$, could be transformed into a $TBTCI(t(t + 1), 3, t, 3t)$ as long as it has at least one transversal. For example:

0	4	3	2	1	→	0	4	3	2	1
3	1	4	0	2		3	1	4	0	2
4	2	1	3	0		1	0	2	4	3
1	0	2	4	3		4	2	1	3	0
2	3	0	1	4		2	3	0	1	4

→ $TBTCI(20, 3, 4, 12)$:

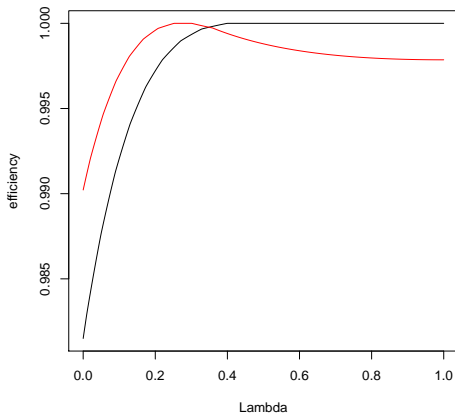
Element: 4 3 2 1 3 4 0 2 1 0 4 3 4 2 1 0 2 3 0 1
 Column: 1 2 3 4 0 2 3 4 0 1 3 4 0 1 2 4 0 1 2 3
 Row: 0 0 0 0 1 1 1 1 2 2 2 2 3 3 3 3 4 4 4 4

Theorem

The juxtaposition of any finite collection of TBTCI's with the common number of periods and treatments would still be a TBTCI as long as we still have $|n_{d0u} - n_{d0v}| \leq 1$ and $|\tilde{n}_{d0u} - \tilde{n}_{d0v}| \leq 1$ where u and v are two different subjects in the resulting design.

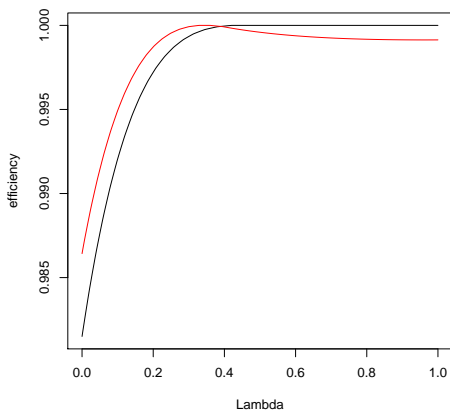
$$\begin{array}{c}
 \text{TBTCI}(36, 3, 4, 36) \\
 \downarrow \\
 \text{TBTCI}(180, 3, 4, 180) \\
 + \\
 \text{TBTCI}(20, 3, 4, 12) \\
 \parallel \\
 \text{TBTCI}(200, 3, 4, 192)
 \end{array}$$

$$\begin{array}{c}
 \text{TBTCI}(36, 3, 4, 36) \\
 \downarrow \\
 \text{TBTCI}(360, 3, 4, 360) \\
 + \\
 \text{TBTCI}(20, 3, 4, 12) \\
 \parallel \\
 \text{TBTCI}(380, 3, 4, 372)
 \end{array}$$



$$\lambda = \theta / (1 + \theta).$$

$TBTCI(180, 3, 4, 180)$ vs $TBTCI(200, 3, 4, 192)$



$$\lambda = \theta / (1 + \theta).$$

$TBTCI(360, 3, 4, 360)$ vs $TBTCI(380, 3, 4, 272)$

For $r_{d0} < n, p \geq 4$

Starting from the special case of $p = 5, t = 4$, we have the following 4 mutually orthogonal Latin Squares:

$$L_1 : \begin{array}{ccccc} 0 & 1 & 2 & 3 & 4 \\ 1 & 2 & 3 & 4 & 0 \\ 2 & 3 & 4 & 0 & 1 \\ 3 & 4 & 0 & 1 & 2 \\ 4 & 0 & 1 & 2 & 3 \end{array}$$

$$L_2 : \begin{array}{ccccc} 0 & 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 0 & 1 \\ 4 & 0 & 1 & 2 & 3 \\ 1 & 2 & 3 & 4 & 0 \\ 3 & 4 & 0 & 1 & 2 \end{array}$$

$$L_3 : \begin{array}{ccccc} 0 & 1 & 2 & 3 & 4 \\ 3 & 4 & 0 & 1 & 2 \\ 1 & 2 & 3 & 4 & 0 \\ 4 & 0 & 1 & 2 & 3 \\ 2 & 3 & 4 & 0 & 1 \end{array}$$

$$L_4 : \begin{array}{ccccc} 0 & 1 & 2 & 3 & 4 \\ 4 & 0 & 1 & 2 & 3 \\ 3 & 4 & 0 & 1 & 2 \\ 2 & 3 & 4 & 0 & 1 \\ 1 & 2 & 3 & 4 & 0 \end{array}$$

Since L_1, L_2 and L_3 has the main diagonal as a common transversal, we rename the symbols to get the following:

$$\begin{array}{ccccc}
 0 & 3 & 1 & 4 & 2 \\
 3 & 1 & 4 & 2 & 0 \\
 L'_1: & 1 & 4 & 2 & 0 & 3 \\
 4 & 2 & 0 & 3 & 1 \\
 2 & 0 & 3 & 1 & 4
 \end{array}
 \quad
 \begin{array}{ccccc}
 0 & 2 & 4 & 1 & 3 \\
 4 & 1 & 3 & 0 & 2 \\
 L'_2: & 3 & 0 & 2 & 4 & 1 \\
 2 & 4 & 1 & 3 & 0 \\
 1 & 3 & 0 & 2 & 4
 \end{array}
 \quad
 \begin{array}{ccccc}
 0 & 4 & 3 & 2 & 1 \\
 2 & 1 & 0 & 4 & 3 \\
 L'_3: & 4 & 3 & 2 & 1 & 0 \\
 1 & 0 & 4 & 3 & 2 \\
 3 & 2 & 1 & 0 & 4
 \end{array}$$

When we go through each entry except the main diagonal, we have $TBTCI(20, 5, 4, 20)$

$$\begin{array}{l}
 L'_1: 3\ 1\ 4\ 2\ 3\ 4\ 2\ 0\ 1\ 4\ 0\ 3\ 4\ 2\ 0\ 1\ 2\ 0\ 3\ 1 \\
 L'_2: 2\ 4\ 1\ 3\ 4\ 3\ 0\ 2\ 3\ 0\ 4\ 1\ 2\ 4\ 1\ 0\ 1\ 3\ 0\ 2 \\
 L'_3: 4\ 3\ 2\ 1\ 2\ 0\ 4\ 3\ 4\ 3\ 1\ 0\ 1\ 0\ 4\ 2\ 3\ 2\ 1\ 0 \\
 \text{Column: } 1\ 2\ 3\ 4\ 0\ 2\ 3\ 4\ 0\ 1\ 3\ 4\ 0\ 1\ 2\ 4\ 0\ 1\ 2\ 3 \\
 \text{Row: } 0\ 0\ 0\ 0\ 1\ 1\ 1\ 1\ 2\ 2\ 2\ 2\ 3\ 3\ 3\ 3\ 4\ 4\ 4\ 4
 \end{array}$$

By selecting any 4 or 3 of the 5 rows of the $TBTCI(20, 5, 4, 20)$, we get $TBTCI(20, 4, 4, 16)$ or $TBTCI(20, 3, 4, 12)$ respectively.

Theorem

A type I orthogonal array $OA_I(t(t+1), p, t+1, 2)$ and a $TBTCI(t(t+1), p, t, pt)$ coexists.

Corollary

When there exists m mutually orthogonal Latin Squares of order $t+1$, we can construct $TBTCI(t(t+1), p, t, pt)$ for all $p \leq m+1$.

Remark: Note that $r_{d0}/n = p/(t+1)$ for the constructed designs. However these Latin Square based TBTCI designs are not optimal due to having small values of r_{d0}/n whenever $p/(t+1)$ is small. One way to rectify this problem is to juxtapose these designs with TBTCI designs with $r_{d0}/n = 1$ — This is an open problem when $p < t$ [for $p = t, t+1$ see Hedayat and Yang (2005)]

For $r_{d0} > n, p = 4$

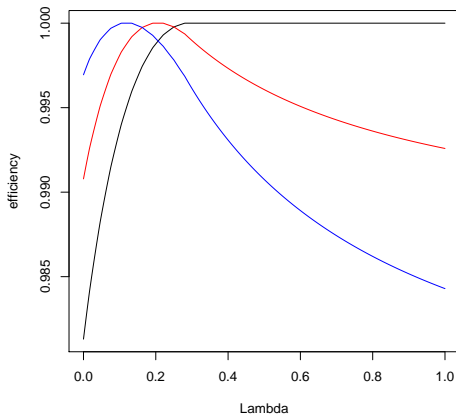
We can construct a $TBTCI(2t(t-1), 4, t, 4t(t-1))$ with $r_{d0}/n = 2$ as the following:

Order the units from 1 to $2t(t-1)$. For each of the first $t(t-1)$ units, assign the control treatment in periods 1 and 3, and for periods 2 and 4, use the $t(t-1)$ ordered pair of different test treatments. For each of the remaining $t(t-1)$ units, assign the control treatment in periods 2 and 4, and for periods 1 and 3, use the $t(t-1)$ ordered pair of different test treatments.

$TBTCI(12, 4, 3, 24)$:

0	0	0	0	0	0	2	3	1	3	1	2
2	3	1	3	1	2	0	0	0	0	0	0
0	0	0	0	0	0	1	1	2	2	3	3
1	1	2	2	3	3	0	0	0	0	0	0

$TBTCI(4, 4, 3, 4)$	$TBTCI(4, 4, 3, 4)$	$TBTCI(4, 4, 3, 4)$
↓	↓	↓
$TBTCI(224, 4, 3, 224)$	$TBTCI(212, 4, 3, 212)$	$TBTCI(200, 4, 3, 200)$
+	+	+
$0 \times TBTCI(12, 4, 3, 24)$	$1 \times TBTCI(12, 4, 3, 24)$	$2 \times TBTCI(12, 4, 3, 24)$
$TBTCI(224, 4, 3, 224)$	$TBTCI(224, 4, 3, 236)$	$TBTCI(224, 4, 3, 248)$



$$\lambda = \theta / (1 + \theta).$$

$TBTCI(224, 4, 3, 248)$, $TBTCI(224, 4, 3, 236)$ and $TBTCI(224, 4, 3, 224)$.

- Construction of TBTCI designs with $r_{d0} > n$ for $p \geq 5$.
- Alternative methods of constructing TBTCI designs for cases with solutions.
- Search for optimal designs within larger class of competing designs and the related construction problems.
- Trade-off Problems

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Min Yang, Ph.D.
Professor of Statistics
University of Missouri

Thank You!