

OPTIMAL DESIGNS FOR 2^k FACTORIAL EXPERIMENTS WITH BINARY RESPONSE

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Supplementary Materials

Connection between General Equivalence Theorem and Theorem 3.1.1:

Extending the notations of this paper, we consider the problem when a design $\xi = \{(\mathbf{x}_i, p_i), i = 1, \dots, 2^k\}$ maximizes the D-criterion $|M(\xi)| = |X'WX|$, where \mathbf{x}_i is the i th row of X and X is the $2^k \times (d+1)$ model matrix.

General Equivalence Theorem (see, for example, Atkinson et al. (2007)): ξ maximizes $|M(\xi)|$ (or equivalently minimizes $\Psi\{M(\xi)\} = -\log|M(\xi)|$) if and only if

$$w_i \mathbf{x}_i' (X'WX)^{-1} \mathbf{x}_i \leq d+1$$

for each $i = 1, \dots, 2^k$ and equality holds if $p_i > 0$.

Here's the outline of the proof of the General Equivalence Theorem described in Atkinson et al. (2007, §9.2, page 122): For each $i = 1, \dots, 2^k$, let $\bar{\xi}_i$ be the design supported only on \mathbf{x}_i , or in other words, it puts unit mass at the point \mathbf{x}_i and let $\xi'_i = (1-\alpha)\xi + \alpha\bar{\xi}_i$. The derivative of Ψ in the direction $\bar{\xi}_i$ or \mathbf{x}_i is

$$\phi(\mathbf{x}_i, \xi) = \lim_{\alpha \rightarrow 0^+} \frac{1}{\alpha} [\Psi(M(\xi'_i)) - \Psi(M(\xi))] = (d+1) - w_i \mathbf{x}_i' (X'WX)^{-1} \mathbf{x}_i.$$

Then ξ is D-optimal if and only if $\min_i \phi(\mathbf{x}_i, \xi) = 0$ and $\phi(\mathbf{x}_i, \xi) = 0$ if $p_i > 0$. Comparing with our proof of Theorem 3.1.1, $\xi'_i = (1-\alpha)\xi + \alpha\bar{\xi}_i = \xi + \alpha(\bar{\xi}_i - \xi)$ corresponds to our $\mathbf{p}_r + u\boldsymbol{\delta}_i^{(r)}$ with u replaced by α and $\boldsymbol{\delta}_i^{(r)}$ replaced by $\bar{\xi}_i - \xi$. Therefore, $\phi(\mathbf{x}_i, \xi)$ is equal to $\left. \frac{\partial f^{(r)}(\mathbf{p}_r + u\boldsymbol{\delta}_i^{(r)})}{\partial u} \right|_{u=0}$ and the if and only if condition comparing Atkinson et al. (2007) becomes

$$\left. \frac{\partial f^{(r)}(\mathbf{p}_r + u\boldsymbol{\delta}_i^{(r)})}{\partial u} \right|_{u=0} \begin{matrix} = 0 & \text{if } p_i > 0; \\ \leq 0 & \text{otherwise.} \end{matrix}$$

The major difference between the general equivalence theorem and Theorem 3.11 is that the general equivalence theorem ends up with the inverse of $X'WX$, while we expressed the same set of conditions in terms of determinants with the aid of Lemma 3.1.1, as well as Lemma S1.2 and Lemma S1.3. \square

Additional Results for Example 4.1: Consider a 2^3 main-effects model with logit link. Suppose $\beta_1 = 0$. As a corollary of Theorem 4.1.4, the regular fractions $\{1, 4, 6, 7\}$, $\{2, 3, 5, 8\}$ are D-optimal half-fractions if and only

$$4 \nu(|\beta_0| + |\beta_2| + |\beta_3|) \geq \nu \left(|\beta_0| + |\beta_2| + |\beta_3| - 2 \max_{0 \leq i \leq 3} |\beta_i| \right).$$

Note that $\nu(\eta) = \frac{1}{2+e^\eta+e^{-\eta}}$ for logit link, which is symmetric about 0. To simplify the notations, let $\beta_{2\vee 3} = \max\{|\beta_2|, |\beta_3|\}$ and $\beta_{2\wedge 3} = \min\{|\beta_2|, |\beta_3|\}$. The regular fractions $\{1, 4, 6, 7\}$, $\{2, 3, 5, 8\}$ are D-optimal half-fractions if and only if one of three conditions below is satisfied:

$$\begin{aligned} \text{(i)} \quad & |\beta_2| + |\beta_3| \leq \log 2; \tag{S.1} \\ \text{(ii)} \quad & |\beta_2| + |\beta_3| > \log 2, \beta_{2\vee 3} \leq \log \left(1 + e^{-\beta_{2\wedge 3}} + \left[1 + e^{-\beta_{2\wedge 3}} + e^{-2\beta_{2\wedge 3}} \right]^{1/2} \right), \\ & \text{and } |\beta_0| \leq \log \left(\frac{2 \exp\{|\beta_2| + |\beta_3|\} - 1}{\exp\{|\beta_2| + |\beta_3|\} - 2} \right); \\ \text{(iii)} \quad & \beta_{2\vee 3} > \log \left(1 + e^{-\beta_{2\wedge 3}} + \left[1 + e^{-\beta_{2\wedge 3}} + e^{-2\beta_{2\wedge 3}} \right]^{1/2} \right), \\ & |\beta_{2\vee 3}| \leq \log \left(\frac{2e^{|\beta_{2\wedge 3}|} - 1}{e^{|\beta_{2\wedge 3}|} - 2} \right) \text{ and } |\beta_0| \leq \log \left(\frac{2e^{\beta_{2\vee 3}} - 1}{e^{\beta_{2\vee 3}} - 2} \right) - \beta_{2\wedge 3}. \end{aligned}$$

The above result is displayed in the right panel of Figure 4.2. In the x - and y -axis, we have plotted β_2 and β_3 respectively. The rhomboidal region at the center (marked as ∞) represents the region where the regular fractions will always be D-optimal, irrespective of the values of β_0 . The contours outside this region are for the upper bound of $|\beta_0|$. Regular fractions will be D-optimal if the values of $|\beta_0|$ will be smaller than the upper bound with β_2 and β_3 falling inside the region outlined by the contour.

Proofs

We need two lemmas before the proof of Theorem 3.1.1.

Lemma S1.2 Suppose $\mathbf{p} = (p_1, \dots, p_{2^k})'$ satisfies $f(\mathbf{p}) > 0$. Given $i = 1, \dots, 2^k$,

$$f_i(z) = a_i z(1-z)^d + b_i(1-z)^{d+1}, \quad (\text{S.2})$$

for some constants a_i and b_i . If $p_i > 0$, $b_i = f_i(0)$, $a_i = \frac{f(\mathbf{p}) - b_i(1-p_i)^{d+1}}{p_i(1-p_i)^d}$; otherwise, $b_i = f(\mathbf{p})$, $a_i = f_i(\frac{1}{2}) \cdot 2^{d+1} - b_i$. Note that $a_i \geq 0$, $b_i \geq 0$, and $a_i + b_i > 0$. \square

Lemma S1.3 Let $h(z) = az(1-z)^d + b(1-z)^{d+1}$ with $0 \leq z \leq 1$ and $a \geq 0, b \geq 0, a + b > 0$. If $a > b(d+1)$, then $\max_z h(z) = \left(\frac{d}{a-b}\right)^d \left(\frac{a}{d+1}\right)^{d+1}$ at $z = \frac{a-b(d+1)}{(a-b)(d+1)} < 1$. Otherwise, $\max_z h(z) = b$ at $z = 0$. \square

Proof of Theorem 3.1.1: Note that $f(\mathbf{p}) > 0$ implies $0 \leq p_i < 1$ for each $i = 1, \dots, 2^k$. Since $\sum_i p_i = 1$, without any loss of generality, we assume $p_{2^k} > 0$. Define $\mathbf{p}_r = (p_1, \dots, p_{2^k-1})'$, and $f^{(r)}(\mathbf{p}_r) = f(p_1, \dots, p_{2^k-1}, 1 - \sum_{i=1}^{2^k-1} p_i)$.

For $i = 1, \dots, 2^k - 1$, let $\boldsymbol{\delta}_i^{(r)} = (-p_1, \dots, -p_{i-1}, 1 - p_i, -p_{i+1}, \dots, -p_{2^k-1})'$. Then $f_i(z) = f^{(r)}(\mathbf{p}_r + u\boldsymbol{\delta}_i^{(r)})$ with $u = \frac{z-p_i}{1-p_i}$. Since the determinant $|\boldsymbol{\delta}_1^{(r)}, \dots, \boldsymbol{\delta}_{2^k-1}^{(r)}| = p_{2^k} \neq 0$, $\boldsymbol{\delta}_1^{(r)}, \dots, \boldsymbol{\delta}_{2^k-1}^{(r)}$ are linearly independent and thus may serve as a new basis of

$$S_r = \{(p_1, \dots, p_{2^k-1})' \mid \sum_{i=1}^{2^k-1} p_i \leq 1, \text{ and } p_i \geq 0, i = 1, \dots, 2^k - 1\}. \quad (\text{S.3})$$

Since $\log f^{(r)}(\mathbf{p}_r)$ is concave, \mathbf{p}_r maximizes $f^{(r)}$ if and only if along each direction $\boldsymbol{\delta}_i^{(r)}$,

$$\left. \frac{\partial f^{(r)}(\mathbf{p}_r + u\boldsymbol{\delta}_i^{(r)})}{\partial u} \right|_{u=0} = 0 \text{ if } p_i > 0; \leq 0 \text{ otherwise.}$$

That is, $f_i(z)$ attains its maximum at $z = p_i$, for each $i = 1, \dots, 2^k - 1$ (and thus for $i = 2^k$). Based on Lemma S1.2 and Lemma S1.3, it implies one of the two cases:

- (i) $p_i = 0$ and $f_i(\frac{1}{2}) \cdot 2^{d+1} - f(\mathbf{p}) \leq f(\mathbf{p})(d+1)$;

- (ii) $p_i > 0$, $a > b(d+1)$, and $a - b(d+1) = p_i(a-b)(d+1)$, where $b = f_i(0)$,
and $a = \frac{f(\mathbf{p}) - b(1-p_i)^{d+1}}{p_i(1-p_i)^d}$.

The conclusion needed can be obtained by simplifying those two cases above. \square

Proof of Theorem 3.1.2: Let \mathbf{p}_I be the minimally supported design satisfying $p_{i_1} = p_{i_2} = \dots = p_{i_{d+1}} = \frac{1}{d+1}$. Note that if $|X[i_1, i_2, \dots, i_{d+1}]| = 0$, \mathbf{p}_I can not be D-optimal. Suppose $|X[i_1, i_2, \dots, i_{d+1}]| \neq 0$, \mathbf{p}_I is D-optimal if and only if \mathbf{p}_I satisfies the conditions of Theorem 3.1.1. By Lemma 3.1.1, $f(\mathbf{p}_I) = (d+1)^{-(d+1)} |X[i_1, i_2, \dots, i_{d+1}]|^2 w_{i_1} w_{i_2} \dots w_{i_{d+1}}$.

For $i \in \mathbf{I}$, $p_i = \frac{1}{d+1}$, $f_i(0) = 0$. By case (ii) of Theorem 3.1.1, $p_i = \frac{1}{d+1}$ maximizes $f_i(x)$. For $i \notin \mathbf{I}$, $p_i = 0$,

$$\begin{aligned} f_i\left(\frac{1}{2}\right) &= [2(d+1)]^{-(d+1)} |X[i_1, \dots, i_{d+1}]|^2 w_{i_1} \dots w_{i_{d+1}} \\ &+ 2^{-(d+1)} (d+1)^{-d} w_i \cdot w_{i_1} \dots w_{i_{d+1}} \sum_{j \in \mathbf{I}} \frac{|X[\{i\} \cup \mathbf{I} \setminus \{j\}]|^2}{w_j}. \end{aligned}$$

Then $p_i = 0$ maximizes $f_i(x)$ if and only if $f_i\left(\frac{1}{2}\right) \leq f(\mathbf{p})^{\frac{d+2}{2d+1}}$, which is equivalent to

$$\sum_{j \in \mathbf{I}} \frac{|X[\{i\} \cup \mathbf{I} \setminus \{j\}]|^2}{w_j} \leq \frac{|X[i_1, i_2, \dots, i_{d+1}]|^2}{w_i}.$$

\square

Proof of Theorem 3.3.3: Suppose the lift-one algorithm or its modified version converges at $\mathbf{p}^* = (p_1^*, \dots, p_{2^k}^*)'$. According to the algorithm, $|X'WX| > 0$ at \mathbf{p}^* and $p_i^* < 1$ for $i = 1, \dots, 2^k$. The proof of Theorem 3.1.1 guarantees that \mathbf{p}^* maximizes $f(\mathbf{p}) = |X'WX|$.

Now we show that the modified lift-one algorithm must converge to the maximum value $\max_{\mathbf{p}} |X'WX|$. Based on the algorithm, we obtain a sequence of designs $\{\mathbf{p}_n\}_{n \geq 0} \subset S_r$ defined in (S.3) such that $|X'WX| > 0$. We only need to check the case when the sequence is infinite. To simplify the notation, here we still denote $f(\mathbf{p}) = f(p_1, \dots, p_{2^k-1}, 1 - \sum_{i=1}^{2^k-1} p_i)$ for $\mathbf{p} = (p_1, \dots, p_{2^k-1})' \in S_r$. Since that $f(\mathbf{p})$ is bounded from above on S_r and $f(\mathbf{p}_n)$ strictly increases with n , then $\lim_{n \rightarrow \infty} f(\mathbf{p}_n)$ exists.

Suppose $\lim_{n \rightarrow \infty} f(\mathbf{p}_n) < \max_{\mathbf{p}} |X'WX|$. Since S_r is compact, there exists a $\mathbf{p}_* = (p_1^*, \dots, p_{2^k-1}^*)' \in S_r$ and a subsequence $\{\mathbf{p}_{n_s}\}_{s \geq 1} \subset \{\mathbf{p}_{10m}\}_{m \geq 0} \subset \{\mathbf{p}_n\}_{n \geq 0}$ such that

$$0 < f(\mathbf{p}_*) = \lim_{n \rightarrow \infty} f(\mathbf{p}_n) = \lim_{s \rightarrow \infty} f(\mathbf{p}_{n_s}) \text{ and } \|\mathbf{p}_{n_s} - \mathbf{p}_*\| \rightarrow 0 \text{ as } s \rightarrow \infty,$$

where “ $\|\cdot\|$ ” represents the Euclidean distance. Since \mathbf{p}_* is not a solution maximizing $|X'WX|$, by the proof of Theorem 3.1.1 and the modified algorithm, there exists a $\delta_i^{(r)}$ at \mathbf{p}_* and an optimal $u_* \neq 0$ such that $\mathbf{p}_* + u_* \delta_i^{(r)}(\mathbf{p}_*) \in S_r$ and $\Delta := f(\mathbf{p}_* + u_* \delta_i^{(r)}(\mathbf{p}_*)) - f(\mathbf{p}_*) > 0$.

As $s \rightarrow \infty$, $\mathbf{p}_{n_s} \rightarrow \mathbf{p}_*$, its i th direction $\delta_i^{(r)}(\mathbf{p}_{n_s})$ determined by the algorithm $\rightarrow \delta_i^{(r)}(\mathbf{p}_*)$, and the optimal $u(\mathbf{p}_{n_s}) \rightarrow u_*$. Thus $\mathbf{p}_{n_s} + u(\mathbf{p}_{n_s}) \delta_i^{(r)}(\mathbf{p}_{n_s}) \rightarrow \mathbf{p}_* + u_* \delta_i^{(r)}(\mathbf{p}_*)$ and

$$f(\mathbf{p}_{n_s} + u(\mathbf{p}_{n_s}) \delta_i^{(r)}(\mathbf{p}_{n_s})) - f(\mathbf{p}_{n_s}) \rightarrow f(\mathbf{p}_* + u_* \delta_i^{(r)}(\mathbf{p}_*)) - f(\mathbf{p}_*) = \Delta.$$

For all large enough s , $f(\mathbf{p}_{n_s} + u(\mathbf{p}_{n_s}) \delta_i^{(r)}(\mathbf{p}_{n_s})) - f(\mathbf{p}_{n_s}) > \Delta/2 > 0$. However,

$$f(\mathbf{p}_{n_s} + u(\mathbf{p}_{n_s}) \delta_i^{(r)}(\mathbf{p}_{n_s})) - f(\mathbf{p}_{n_s}) \leq f(\mathbf{p}_{n_s+1}) - f(\mathbf{p}_{n_s}) \leq f(\mathbf{p}_*) - f(\mathbf{p}_{n_s}) \rightarrow 0$$

The contradiction implies that $\lim_{n \rightarrow \infty} f(\mathbf{p}_n) = \max_{\mathbf{p}} |X'WX|$. \square

Proof of Theorem 4.1.4: Given $\beta_1 = 0$, we have $w_1 = w_5 = \nu(\beta_0 + \beta_2 + \beta_3)$, $w_2 = w_6 = \nu(\beta_0 + \beta_2 - \beta_3)$, $w_3 = w_7 = \nu(\beta_0 - \beta_2 + \beta_3)$, $w_4 = w_8 = \nu(\beta_0 - \beta_2 - \beta_3)$. The goal is to find a half-fraction $\mathbf{I} = \{i_1, i_2, i_3, i_4\}$ which maximizes $s(\mathbf{I}) := |X[i_1, i_2, i_3, i_4]|^2 w_{i_1} w_{i_2} w_{i_3} w_{i_4}$. For regular half-fractions $\mathbf{I} = \{1, 4, 6, 7\}$ or $\{2, 3, 5, 8\}$, $s(\mathbf{I}) = 256 w_1 w_2 w_3 w_4$. Note that $|X[i_1, i_2, i_3, i_4]|^2 = 0$ for 12 half-fractions identified by $1 = \pm A$, $1 = \pm B$, $1 = \pm C$, $1 = \pm AB$, $1 = \pm AC$, or $1 = \pm BC$; and $|X[i_1, i_2, i_3, i_4]|^2 = 64$ for all other 56 cases.

Without any loss of generality, suppose $w_1 \geq w_2 \geq w_3 \geq w_4$. Note that the half-fraction $\{1, 5, 2, 6\}$ identified by $1 = B$ leads to $s(\mathbf{I}) = 0$. Then the competitive half-fractions consist of both 1 and 5, one element from the second block $\{2, 6\}$, and one element from the third block $\{3, 7\}$. The corresponding $s(\mathbf{I}) = 64 w_1^2 w_2 w_3$. In this case, the regular fractions are optimal ones if and only

if $4w_4 \geq w_1$. □

We need the lemma below for Theorem 5.1.6:

Lemma S1.4 *Suppose $k \geq 3$ and $d(d+1) \leq 2^{k+1} - 4$. For any index set $\mathbf{I} = \{i_1, \dots, i_{d+1}\} \subset \{1, \dots, 2^k\}$, there exists another index set $\mathbf{I}' = \{i'_1, \dots, i'_{d+1}\}$ such that*

$$|X[i_1, \dots, i_{d+1}]|^2 = |X[i'_1, \dots, i'_{d+1}]|^2 \text{ and } \mathbf{I} \cap \mathbf{I}' = \emptyset. \quad (\text{S.4})$$

Proof of Lemma S1.4: Note that $k \geq 3$ and $d(d+1) \leq 2^{k+1} - 4$ imply $d+1 \leq 2^{k-1}$ and $\frac{d(d+1)}{2} < 2^k - 1$. Let $\mathbf{I} = \{i_1, \dots, i_{d+1}\} \subset \{1, \dots, 2^k\}$ be the given index set. It can be verified that there exists a nonempty subset $\mathbf{J} \subset \{1, 2, \dots, k\}$, such that (i) the i_1 th, \dots , i_{d+1} th rows of the matrix $[C_1, C_2, \dots, C_k]$ are same as the i'_1 th, \dots , i'_{d+1} th rows of the matrix $[A_1, A_2, \dots, A_k]$, where A_1, \dots, A_k are the columns of X corresponding to the main effects, $C_i = -A_i$ if $i \in \mathbf{J}$ and $C_i = A_i$ otherwise; (ii) $\mathbf{I}' = \{i'_1, \dots, i'_{d+1}\}$ satisfies conditions (S.4). Actually, the index set \mathbf{I}' satisfying (i) always exists once \mathbf{J} is given, since the 2^k rows of matrix $[A_1, \dots, A_k]$ contain all possible vectors in $\{-1, 1\}^k$. Then $|X[i_1, \dots, i_{d+1}]|^2 = |X[i'_1, \dots, i'_{d+1}]|^2$ is guaranteed once \mathbf{I}' satisfies (i). If $\mathbf{I} \cap \mathbf{I}' \neq \emptyset$, then there exists an $i'_a \in \mathbf{I} \cap \mathbf{I}'$ ($a \in \{1, \dots, d+1\}$). Thus $i_a \in \mathbf{I}$ and the i_a th row of $[C_1, \dots, C_k]$ is same as the i'_a th row of $[A_1, \dots, A_k]$. Based on the definitions of C_1, \dots, C_k , the i_a th and i'_a th rows of $[A_1, \dots, A_k]$ have the same entries at A_i for all $i \notin \mathbf{J}$ but different entries at A_i for all $i \in \mathbf{J}$. On the other hand, once the index pair $\{i_a, i'_a\} \subset \mathbf{I}$ is given, it uniquely determines the subset $\mathbf{J} \subset \{1, \dots, k\}$. Note that there are $2^k - 1$ possible nonempty \mathbf{J} but only $\frac{d(d+1)}{2}$ possible pairs in \mathbf{I} . Since $\frac{d(d+1)}{2} < 2^k - 1$, there is at least one \mathbf{J} such that there is no pair in \mathbf{I} corresponding to it. For such a \mathbf{J} , we must have $\mathbf{I} \cap \mathbf{I}' = \emptyset$. □

Proof of Theorem 5.1.6: Fixing any row index set $I = \{i_1, \dots, i_{d+1}\}$ of X such that $|X[i_1, i_2, \dots, i_{d+1}]|^2 > 0$, among all the $(d+1)$ -row fractional designs satisfying $p_i = 0, \forall i \notin \mathbf{I}$, $|X'WX|$ attains its maximum $\left(\frac{1}{d+1}\right)^{d+1} w_{i_1} \cdots w_{i_{d+1}} \times |X[i_1, i_2, \dots, i_{d+1}]|^2$ at \mathbf{p}_I satisfying $p_{i_1} = \cdots = p_{i_{d+1}} = \frac{1}{d+1}$. Given any other index set $\mathbf{I}' = \{i'_1, \dots, i'_{d+1}\}$ with minimally supported design $\mathbf{p}_{I'}$ satisfying $p_{i'_1} = \cdots = p_{i'_{d+1}} = \frac{1}{d+1}$, the loss of efficiency of \mathbf{p}_I with respect to $\mathbf{p}_{I'}$ given $\mathbf{w}_{I'} =$

$(w_1, \dots, w_{2^k})'$ is

$$\begin{aligned} R_{I'}(I) &= 1 - \left(\frac{\psi(\mathbf{p}_I, \mathbf{w}_{I'})}{\psi(\mathbf{p}_{I'}, \mathbf{w}_{I'})} \right)^{\frac{1}{d+1}} = 1 - \left(\frac{w_{i_1} \cdots w_{i_{d+1}} |X[i_1, \dots, i_{d+1}]|^2}{w_{i'_1} \cdots w_{i'_{d+1}} |X[i'_1, \dots, i'_{d+1}]|^2} \right)^{\frac{1}{d+1}} \\ &\leq 1 - \frac{a}{b} \cdot \left(\frac{|X[i_1, \dots, i_{d+1}]|^2}{|X[i'_1, \dots, i'_{d+1}]|^2} \right)^{\frac{1}{d+1}}. \end{aligned}$$

By Lemma S1.4, there always exists an index set $\mathbf{I}' = \{i'_1, \dots, i'_{d+1}\}$ such that $|X[i'_1, \dots, i'_{d+1}]|^2 = |X[i_1, \dots, i_{d+1}]|^2$ and $\mathbf{I} \cap \mathbf{I}' = \emptyset$. Let $\mathbf{w}_{I'} = (w_1, \dots, w_{2^k})'$ satisfy $w_i = b, \forall i \in \mathbf{I}'$ and $w_i = a, \forall i \in \mathbf{I}$ (here we assume (w_1, \dots, w_{2^k}) can take any point in $[a, b]^{2^k}$). Then the loss of efficiency of \mathbf{p}_I with respect to this $\mathbf{w}_{I'}$ is at least $1 - a/b$. If we choose $\mathbf{I} = \{i_1, \dots, i_{d+1}\}$ which maximizes $|X[i_1, \dots, i_{d+1}]|^2$, then the corresponding \mathbf{p}_I attains the minimum value $1 - a/b$ of the maximum loss in efficiency compared to other minimally supported designs. \square

We need two lemmas for the exchange algorithm for integer-valued allocations.

Lemma S1.5 *Let $g(z) = Az(m - z) + Bz + C(m - z) + D$ for real numbers $A > 0, B \geq 0, C \geq 0, D \geq 0$, and integers $m > 0, 0 \leq z \leq m$. Let Δ be the integer closest to $\frac{mA+B-C}{2A}$.*

(i) *If $0 \leq \Delta \leq m$, then $\max_{0 \leq z \leq m} g(z) = mC + D + (mA + B - C)\Delta - A\Delta^2$ at $z = \Delta$.*

(ii) *If $\Delta < 0$, then $\max_{0 \leq z \leq m} g(z) = mC + D$ at $z = 0$.*

(iii) *If $\Delta > m$, then $\max_{0 \leq z \leq m} g(z) = mB + D$ at $z = m$.*

Lemma S1.6 *Let $\mathbf{n} = (n_1, \dots, n_{2^k})'$, $W_n = \text{diag}\{n_1 w_1, \dots, n_{2^k} w_{2^k}\}$, $f(\mathbf{n}) = |X' W_n X|$. Fixing $1 \leq i < j \leq 2^k$, let*

$$\begin{aligned} f_{ij}(z) &= f(n_1, \dots, n_{i-1}, z, n_{i+1}, \dots, n_{j-1}, m - z, n_{j+1}, \dots, n_{2^k}) \\ &\triangleq Az(m - z) + Bz + C(m - z) + D, \end{aligned} \tag{S.5}$$

where $m = n_i + n_j$. Then (i) $D > 0 \implies B > 0$ and $C > 0$; (ii) $B > 0$ or $C > 0 \implies A > 0$; (iii) $f(\mathbf{n}) > 0 \implies A > 0$; (iv) $D = f(n_1, \dots, n_{i-1}, 0, n_{i+1}, \dots, n_{j-1}, 0, n_{j+1}, \dots, n_{2^k})$. (v) Suppose $m > 0$, then $A = \frac{2}{m^2} (2f_{ij}(\frac{m}{2}) - f_{ij}(0) - f_{ij}(m))$, $B = \frac{1}{m} (f_{ij}(m) - D)$, $C = \frac{1}{m} (f_{ij}(0) - D)$.

Exchange algorithm for real-valued allocations

Lemma S1.7 Let $g(z) = Az(e-z) + Bz + C(e-z) + D$ for nonnegative constants A, B, C, D, e . Define $\Delta = \frac{eA+B-C}{2A}$.

- (i) If $0 \leq \Delta \leq e$, then $\max_{0 \leq z \leq e} g(z) = eC + D + \frac{(eA+B-C)^2}{4A}$ at $z = \Delta$.
- (ii) If $\Delta < 0$, then $\max_{0 \leq z \leq e} g(z) = eC + D$ at $z = 0$.
- (iii) If $\Delta > e$, then $\max_{0 \leq z \leq e} g(z) = eB + D$ at $z = e$.

Lemma S1.8 Let $\mathbf{p} = (p_1, \dots, p_{2^k})'$, $f(\mathbf{p}) = |X'WX|$, and

$$\begin{aligned} f_{ij}(z) &:= f(p_1, \dots, p_{i-1}, z, p_{i+1}, \dots, p_{j-1}, e-z, p_{j+1}, \dots, p_{2^k}) \\ &\triangleq Az(e-z) + Bz + C(e-z) + D, \end{aligned}$$

where $1 \leq i < j \leq 2^k$ and $e = p_i + p_j$. Then (i) $D > 0 \implies B > 0$ and $C > 0$; (ii) $B > 0$ or $C > 0 \implies A > 0$; (iii) $f(\mathbf{p}) > 0 \implies A > 0$; (iv) $D = f(p_1, \dots, p_{i-1}, 0, p_{i+1}, \dots, p_{j-1}, 0, p_{j+1}, \dots, p_{2^k})$; (v) Suppose $e > 0$, then $A = \frac{2}{e^2} (2f_{ij}(\frac{e}{2}) - f_{ij}(0) - f_{ij}(e))$, $B = \frac{1}{e} (f_{ij}(e) - D)$, $C = \frac{1}{e} (f_{ij}(0) - D)$.

Exchange algorithm for maximizing $f(\mathbf{p}) = f(p_1, \dots, p_{2^k}) = |X'WX|$

1° Start with an arbitrary design $\mathbf{p}^{(0)} = (p_1^{(0)}, \dots, p_{2^k}^{(0)})'$ such that $f(\mathbf{p}^{(0)}) > 0$.

2° Set up a random order of (i, j) going through all pairs

$$\{(1, 2), (1, 3), \dots, (1, 2^k), (2, 3), \dots, (2^k - 1, 2^k)\}.$$

3° For each (i, j) , if $e := p_i^{(0)} + p_j^{(0)} = 0$, let $\mathbf{p}^{(1)} = \mathbf{p}^{(0)}$ and jump to 5°.

Otherwise, let

$$\begin{aligned} f_{ij}(z) &= f(p_1^{(0)}, \dots, p_{i-1}^{(0)}, z, p_{i+1}^{(0)}, \dots, p_{j-1}^{(0)}, e-z, p_{j+1}^{(0)}, \dots, p_{2^k}^{(0)}) \\ &= Az(e-z) + Bz + C(e-z) + D \end{aligned}$$

with nonnegative constants A, B, C, D determined by Lemma S1.8.

4° Define $\mathbf{p}^{(1)} = (p_1^{(0)}, \dots, p_{i-1}^{(0)}, z_*, p_{i+1}^{(0)}, \dots, p_{j-1}^{(0)}, e-z_*, p_{j+1}^{(0)}, \dots, p_{2^k}^{(0)})'$ where z_* maximizes $f_{ij}(z)$ with $0 \leq z \leq e$ (see Lemma S1.7). Note that $f(\mathbf{p}^{(1)}) = f_{ij}(z_*) \geq f(\mathbf{p}^{(0)}) > 0$.

5° Repeat 2° \sim 4° until convergence (no more increase in terms of $f(\mathbf{p})$ by any pairwise adjustment).

Theorem S1.7 *If the exchange algorithm converges, the converged \mathbf{p} maximizes $|X'WX|$.*

Proof of Theorem S1.7: Suppose the exchange algorithm converges at $\mathbf{p}^* = (p_1^*, \dots, p_{2^k}^*)'$. According to the algorithm, $|X'WX| > 0$ at \mathbf{p}^* . Without any loss of generality, assume $p_{2^k}^* > 0$. Let $\mathbf{p}_r^* = (p_1^*, \dots, p_{2^k-1}^*)$, $l_r(\mathbf{p}_r) = \log f_r(\mathbf{p}_r)$, and $f_r(\mathbf{p}_r) = f(p_1, \dots, p_{2^k-1}, 1 - \sum_{i=1}^{2^k-1} p_i)$. Then for $i = 1, \dots, 2^k - 1$, $\frac{\partial l_r}{\partial p_i} \Big|_{\mathbf{p}_r^*} = \frac{1}{f(\mathbf{p}^*)} \cdot \frac{\partial f_r}{\partial p_i} \Big|_{\mathbf{p}_r^*} = 0$, if $p_i^* > 0$; ≤ 0 , otherwise. Thus \mathbf{p}^* (or \mathbf{p}_r^*) locally maximizes $l(\mathbf{p})$ (or $l_r(\mathbf{p}_r)$), and \mathbf{p}^* attains the global maximum of $f(\mathbf{p})$ on S . \square

Similar to the lift-one algorithm, we may modify the exchange algorithm so that $\mathbf{p}^{(0)}$ won't be updated until all potential pairwise exchanges among p_i 's have been checked. It can be verified that the modified exchange algorithm must converge.