

Two-level orthogonal screening designs with 80, 96 and 112 runs: Construction and evaluation

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Outline

1. Tuberculosis inhibition experiment
2. Criteria to evaluate designs
3. Construction method
4. Results and discussion

Tuberculosis inhibition experiment

- Silva et al. (2016) conducted a study to develop a treatment that maximizes the percentage of inhibition of tuberculosis in infected human cells.
- The first stage of the study involved a screening experiment.
- 14 factors (drugs) at two levels (presence or absence).

Goal:

Identify the influential main effects and two-factor interactions of the factors.

For the experiment, the researchers chose an attractive two-level design with 14 factors and 128 runs.

It is a **strength-4** orthogonal design (Hedayat et al. 1999) which allows to estimate all main effects and all two-factor interactions with full precision.

In the end, however, only 8 main effects and 6 two-factor interactions were active.

Research question:

Could we have identified the active effects with a smaller design?

Literature review

Strength-3 orthogonal designs. Properties:

1. Main effects are **not correlated** with each other **nor** with two-factor interactions.
2. Pairs of two-factor interactions can be correlated.
3. Exist for run sizes which are multiples of 8.

Example:

- Regular fractional factorial designs of resolution IV.
- Folded-over Plackett-Burman designs (nonregular).

Catalogs of strength-3 designs with 16, 24, 32, 40 and 48 runs and up to 24 factors are available (Schoen et al., 2010).

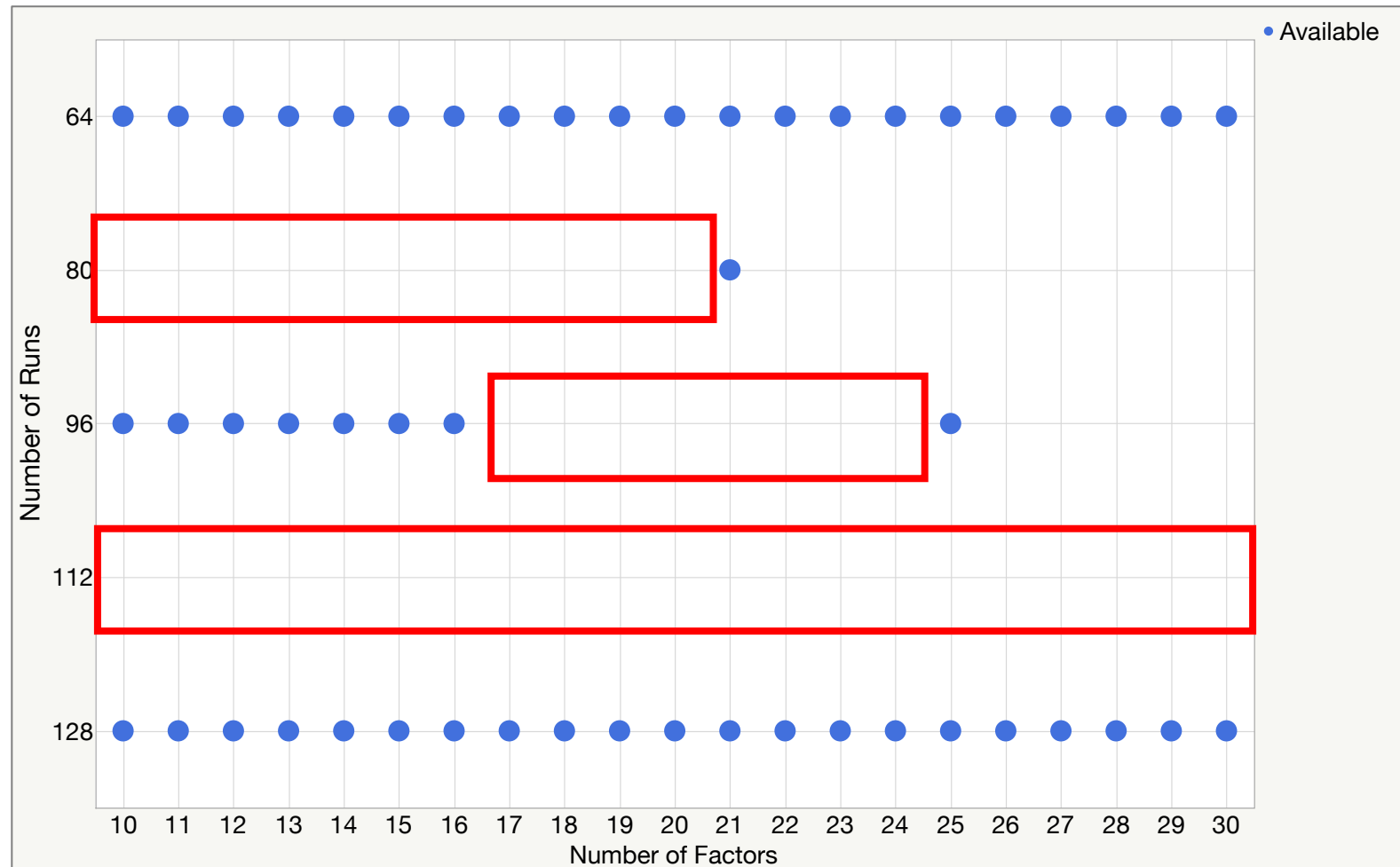
Available strength-3 designs with run sizes from 64 to 128

Chen et al. (1993); Xu & Wong (2007); Cheng et al. (2008); Vazquez & Xu (2019); Vazquez et al. (2019).

Cheng et al. (2008).

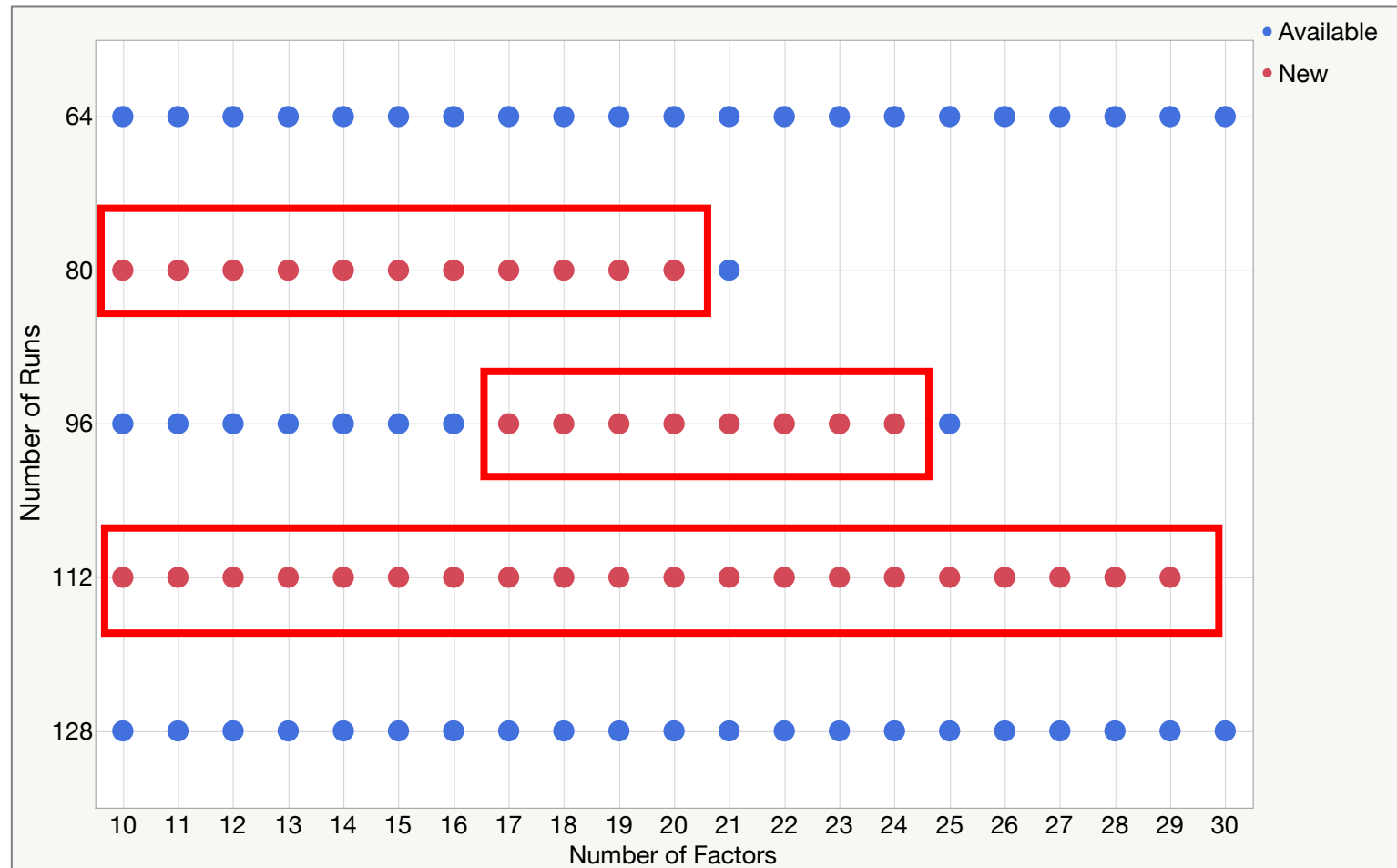
Cheng et al. (2008); Vazquez & Xu (2019).

Ryan & Bulutoglu (2010); Xu & Wong (2007); Vazquez & Xu (2019); Vazquez et al. (2019)



This talk

New strength-3 orthogonal designs which fill the gaps between the available designs.



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Evaluating two-level strength-3 designs

Example: Compare designs with 32 runs and 10 factors.

Regular resolution-IV design

[illegible]

Wu & Hamada
(2009)

Nonregular strength-3 design

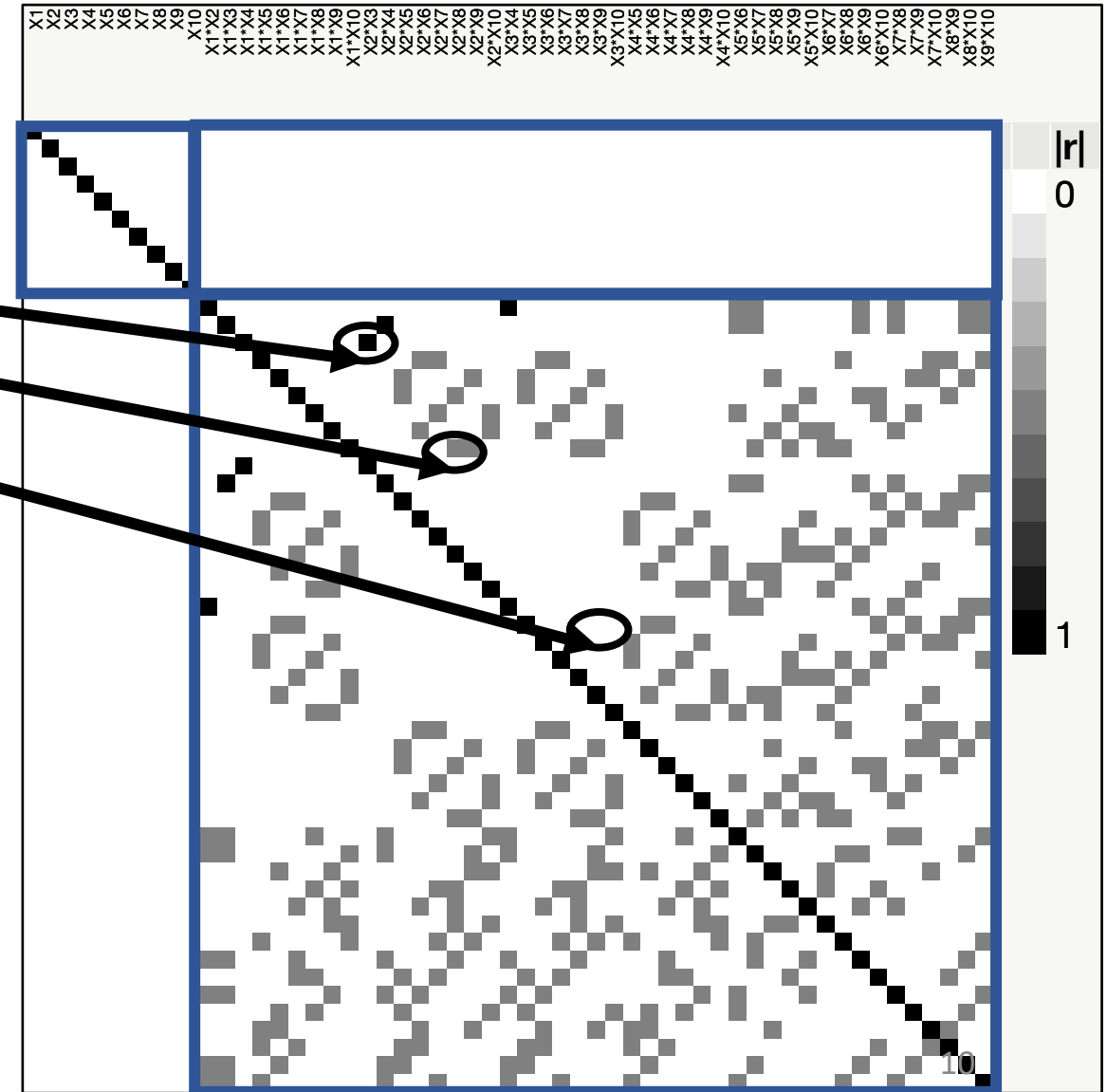
[illegible]

Schoen & Mee
(2010)

Color map on correlations

For 32-run designs, the absolute correlations can have values of 0, 0.5 or 1 (Deng & Tang, 1999).

Nonregular strength-3 design



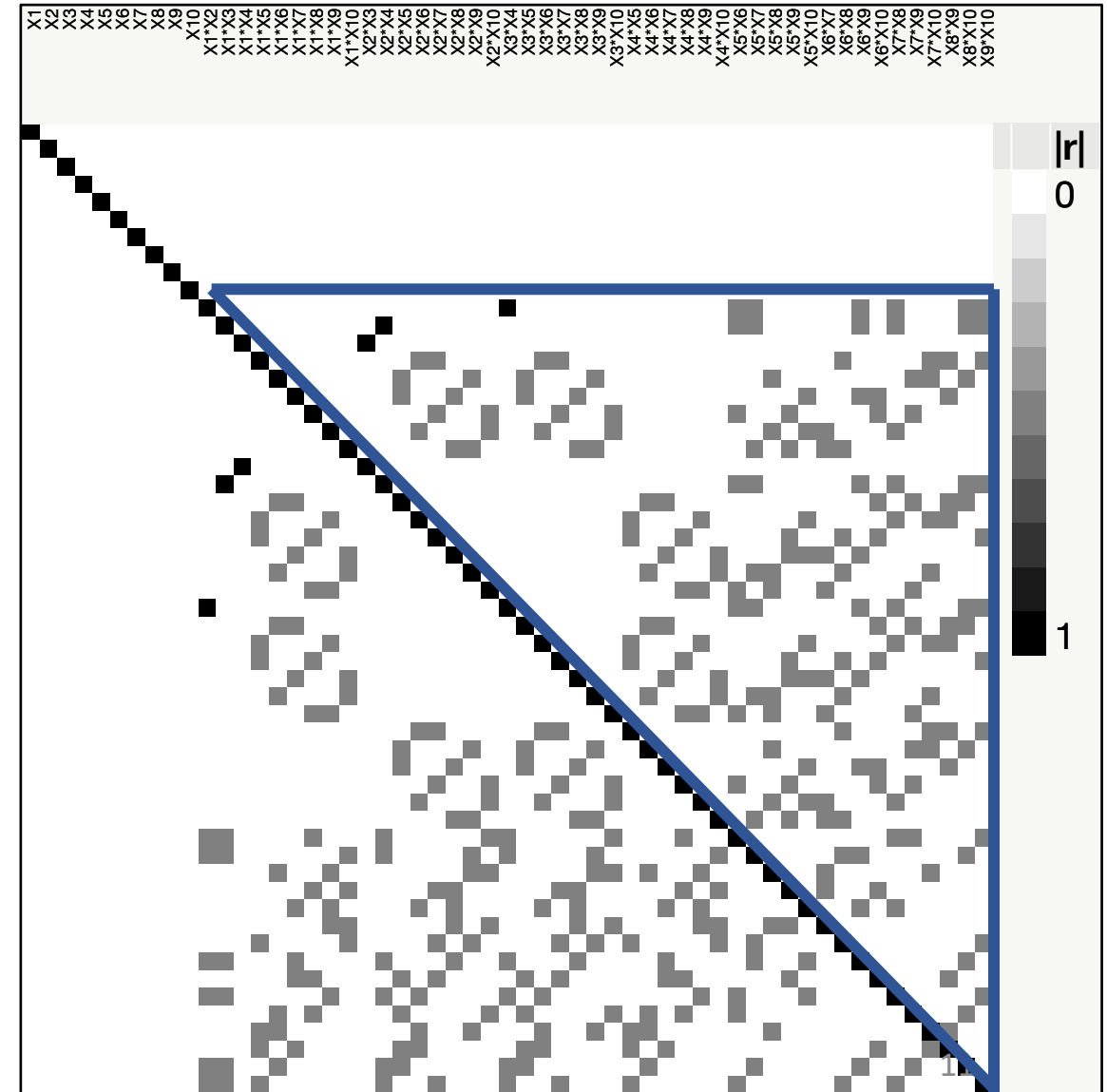
How can we summarize the correlation between pairs of two-factor interactions in strength-3 designs?

Nonregular strength-3 design

Two statistical criteria:

- The F_4 vector
- The B_4 value

(Deng & Tang, 1999; Tang & Deng 1999)



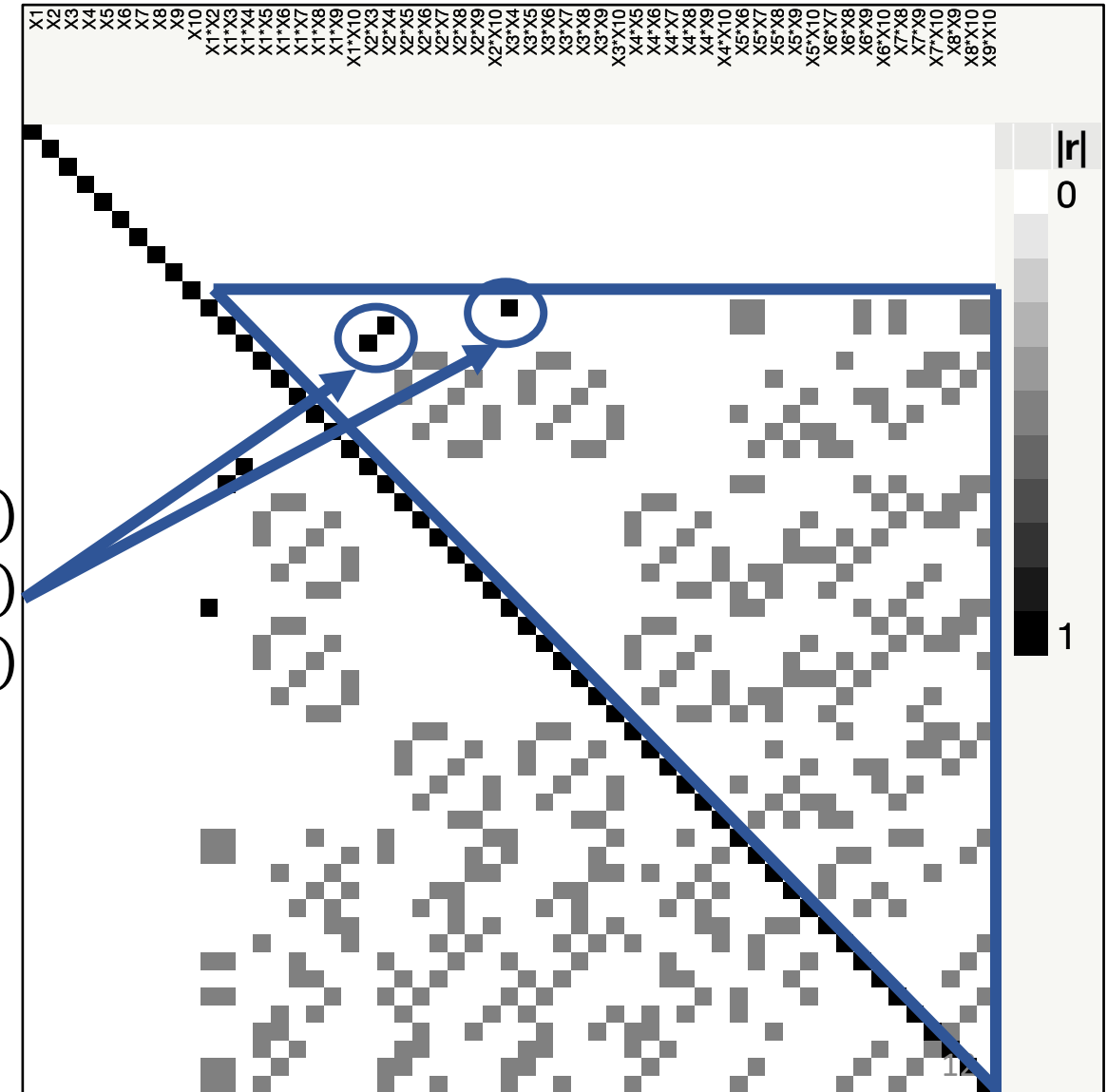
How can we summarize the correlation between pairs of two-factor interactions in strength-3 designs?

Nonregular strength-3 design

The F_4 vector has entries equal to the frequencies for the possible absolute correlation values between pairs of interactions, divided by three (Deng & Tang, 1999).

- $F_4(1, 0.5) = (1, 62)$
- 3 pairs with an abs. correlation of 1.
A good strength-3 design sequentially minimizes the F_4 vector.
- 186 pairs with an abs. correlation of 0.5.

(X_1X_2, X_3X_4)
 (X_1X_3, X_2X_4)
 (X_1X_4, X_2X_3)



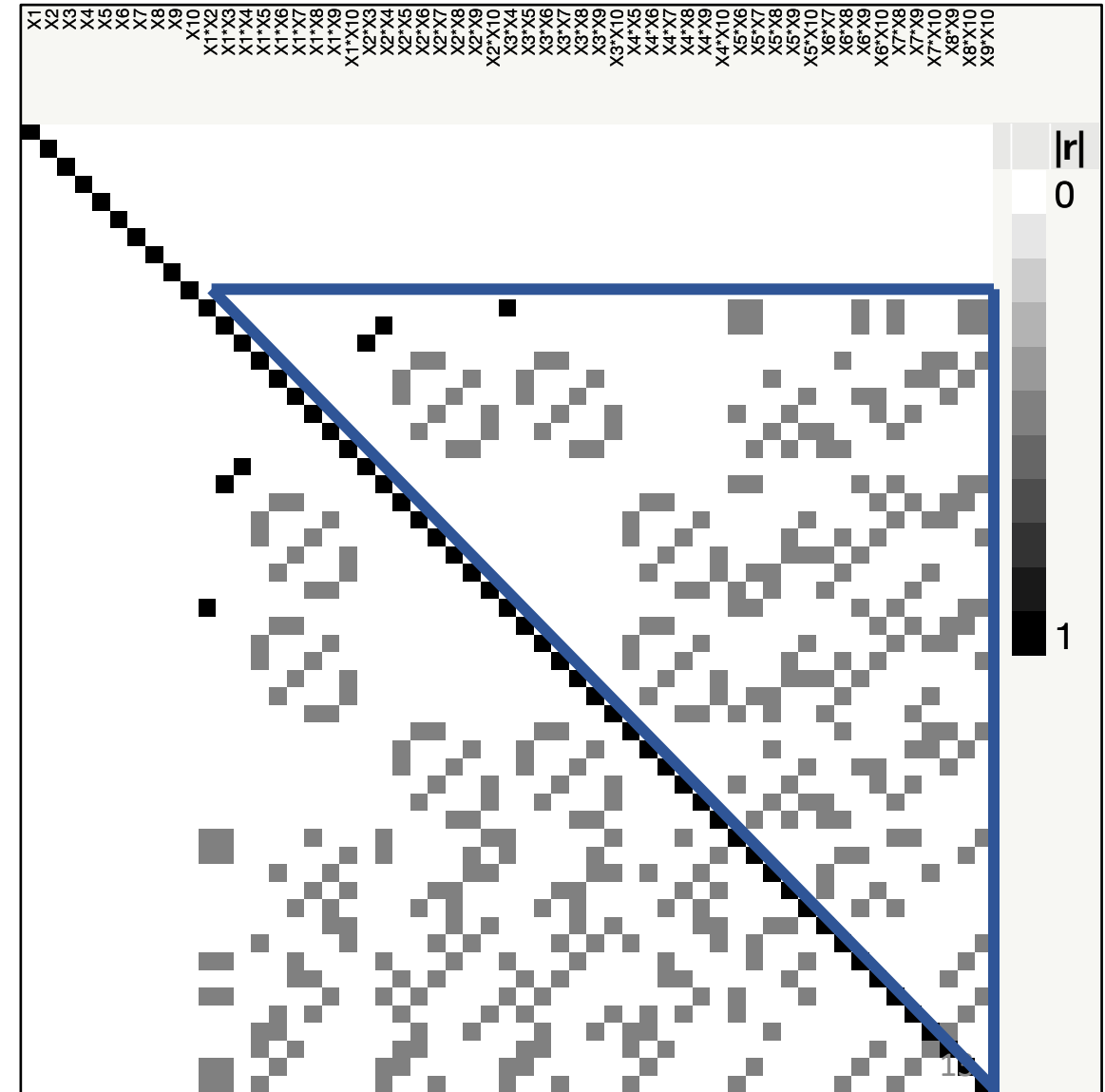
How can we summarize the correlation between pairs of two-factor interactions in strength-3 designs?

The B_4 value equals the sum of squared correlations between pairs of interactions divided by three (Tang & Deng, 1999).

- $B_4 = 1(1)^2 + 62(0.5)^2 = 16.5$

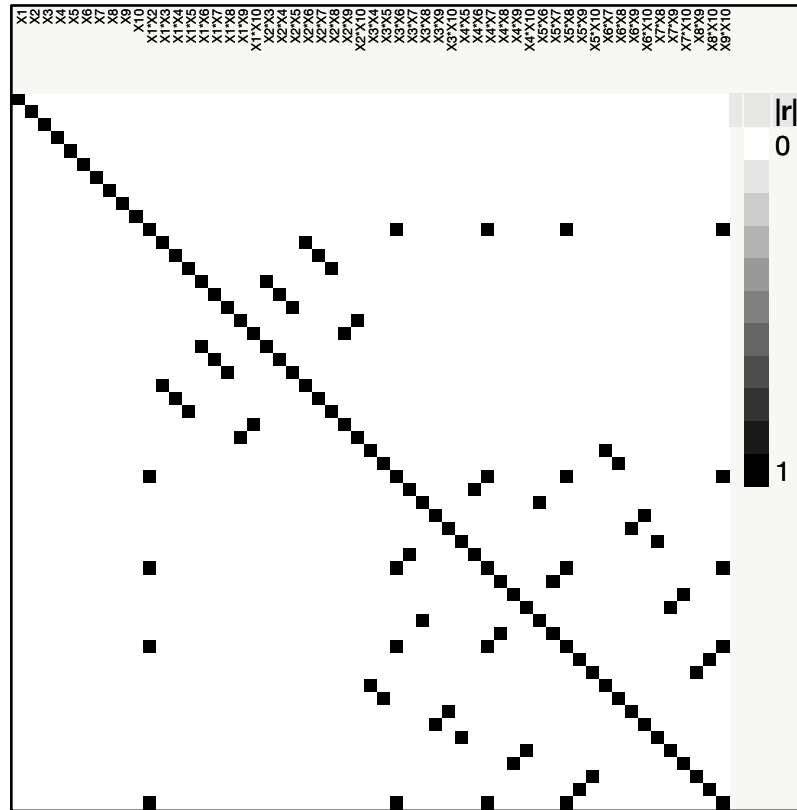
Ideally, the B_4 value of a strength-3 design is small.

Nonregular strength-3 design



Regular resolution-IV design

Example (cont.):

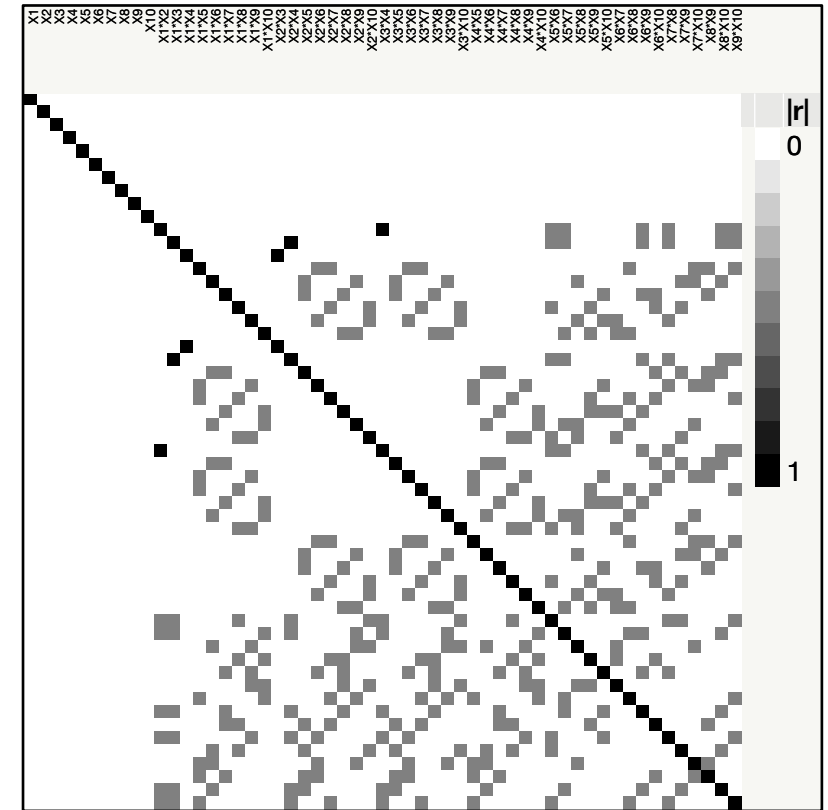


$$F_4(1, 0.5) = (10, 0)$$

30 pairs with an abs. correlation of 1.

$$B_4 = 10$$

Nonregular strength-3 design



$$F_4(1, 0.5) = (1, 62)$$

3 pairs with an abs correlation of 1.

$$B_4 = 16.5$$

Outline

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Construction by example

Goal: Construct a strength-3 design with 32 runs and 9 factors.

Step 1. Consider two good equally-sized strength-3 designs with 16 runs and 8 factors. Call them D_u and D_l .

Minimum aberration 2_{IV}^{8-4} design

$$D_u, D_l = \begin{bmatrix} -1 & -1 & -1 & -1 & -1 & -1 & -1 & -1 \\ -1 & -1 & -1 & 1 & -1 & 1 & 1 & 1 \\ -1 & -1 & 1 & -1 & 1 & -1 & 1 & 1 \\ -1 & -1 & 1 & 1 & 1 & 1 & -1 & -1 \\ -1 & 1 & -1 & -1 & 1 & 1 & -1 & 1 \\ -1 & 1 & -1 & 1 & 1 & -1 & 1 & -1 \\ -1 & 1 & 1 & -1 & -1 & 1 & 1 & -1 \\ -1 & 1 & 1 & 1 & -1 & -1 & -1 & 1 \\ 1 & -1 & -1 & -1 & 1 & 1 & 1 & -1 \\ 1 & -1 & -1 & 1 & 1 & -1 & -1 & 1 \\ 1 & -1 & 1 & -1 & -1 & 1 & -1 & 1 \\ 1 & -1 & 1 & 1 & -1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 & -1 & -1 & 1 & 1 \\ 1 & 1 & -1 & 1 & -1 & 1 & -1 & -1 \\ 1 & 1 & 1 & -1 & 1 & -1 & -1 & -1 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \end{bmatrix}$$

Construction by example

Goal: Construct a strength-3 design with 32 runs and 9 factors.

Step 1. Consider two good equally-sized strength-3 designs with 16 runs and 8 factors. Call them D_u and D_l .

Step 2. Construct the 32-run 9-factor concatenated design D .

$D =$

| | | | | | | | | | |
|----|----|----|----|----|----|----|----|----|-------|
| -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | D_u |
| -1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 | |
| -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 | |
| -1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 | |
| -1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 | |
| -1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 | |
| -1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 | |
| -1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 | |
| -1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 | |
| -1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 | |
| -1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 | |
| -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 | |
| -1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 | |
| -1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 | |
| -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 | |
| -1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | |
| 1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | D_l |
| 1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 | |
| 1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 | |
| 1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 | |
| 1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 | |
| 1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 | |
| 1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 | |
| 1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 | |
| 1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 | |
| 1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 | |
| 1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 | |
| 1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 | |
| 1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 | |
| 1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 | |
| 1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 | |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | |

Construction by example

Step 3. Use the Column-Change Variable Neighborhood Search algorithm (Vazquez et al., 2019) to sequentially minimize the F_4 vector or minimize the B_4 value of D .

- Two moves in D_l : (1) foldover columns
(2) swap columns

Example:

$$F_4(1, 0.5) = (14, 0)$$

$D =$

| | | | | | | | | | |
|----|----|----|----|----|----|----|----|----|-------|
| -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | D_u |
| -1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 | |
| -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 | |
| -1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 | |
| -1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 | |
| -1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 | |
| -1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 | |
| -1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 | |
| -1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 | |
| -1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 | |
| -1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 | |
| -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 | |
| -1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 | |
| -1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 | |
| -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 | |
| -1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | |
| 1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | D_l |
| 1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 | |
| 1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 | |
| 1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 | |
| 1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 | |
| 1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 | |
| 1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 | |
| 1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 | |
| 1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 | |
| 1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 | |
| 1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 | |
| 1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 | |
| 1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 | |
| 1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 | |
| 1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 | |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | |

Construction by example

Step 3. Use the Column-Change Variable Neighborhood Search algorithm (Vazquez et al., 2019) to **sequentially minimize the F_4 vector** or **minimize the B_4 value of D .**

- Two moves in D_l : (1) foldover columns
(2) swap columns

Example:

$$F_4(1, 0.5) = (7, 0)$$

Foldover column 8

$D =$

| | | | | | | | | |
|----|----|----|----|----|----|----|----|----|
| -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 |
| -1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 |
| -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 |
| -1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 |
| -1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 |
| -1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 |
| -1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 |
| -1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 |
| -1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 |
| -1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 |
| -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 |
| -1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 |
| -1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 |
| -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 |
| -1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1 | -1 |
| 1 | -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 |
| 1 | -1 | -1 | 1 | -1 | 1 | -1 | -1 | 1 |
| 1 | -1 | -1 | 1 | 1 | 1 | 1 | 1 | -1 |
| 1 | -1 | 1 | -1 | -1 | 1 | 1 | 1 | 1 |
| 1 | -1 | 1 | -1 | 1 | 1 | -1 | -1 | -1 |
| 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 | -1 |
| 1 | -1 | 1 | 1 | 1 | -1 | -1 | 1 | 1 |
| 1 | 1 | -1 | -1 | -1 | 1 | 1 | -1 | -1 |
| 1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 | 1 |
| 1 | 1 | -1 | 1 | -1 | -1 | 1 | 1 | 1 |
| 1 | 1 | -1 | 1 | 1 | -1 | -1 | -1 | -1 |
| 1 | 1 | 1 | -1 | -1 | -1 | -1 | -1 | 1 |
| 1 | 1 | 1 | -1 | 1 | -1 | 1 | 1 | -1 |
| 1 | 1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | -1 | 1 |

Construction by example

Step 3. Use the Column-Change Variable Neighborhood Search algorithm (Vazquez et al., 2019) to **sequentially minimize the F_4 vector** or **minimize the B_4 value of D .**

- Two moves in D_l : (1) foldover columns
(2) swap columns

Example:

$$F_4(1, 0.5) = (6, 0)$$

Foldover column 9

$D =$

| | | | | | | | | |
|----|----|----|----|----|----|----|----|----|
| -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 |
| -1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 |
| -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 |
| -1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 |
| -1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 |
| -1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 |
| -1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 |
| -1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 |
| -1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 |
| -1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 |
| -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 |
| -1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 |
| -1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 |
| -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 |
| -1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1 | 1 |
| 1 | -1 | -1 | -1 | 1 | -1 | 1 | -1 | -1 |
| 1 | -1 | -1 | 1 | -1 | 1 | -1 | -1 | -1 |
| 1 | -1 | -1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | 1 | -1 | -1 | 1 | 1 | 1 | -1 |
| 1 | -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 |
| 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 | 1 |
| 1 | -1 | 1 | 1 | 1 | -1 | -1 | 1 | -1 |
| 1 | 1 | -1 | -1 | -1 | 1 | 1 | -1 | 1 |
| 1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 | -1 |
| 1 | 1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 |
| 1 | 1 | -1 | 1 | 1 | -1 | -1 | -1 | 1 |
| 1 | 1 | 1 | -1 | -1 | -1 | -1 | -1 | -1 |
| 1 | 1 | 1 | -1 | 1 | -1 | 1 | 1 | 1 |
| 1 | 1 | 1 | 1 | -1 | 1 | -1 | 1 | 1 |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | -1 | -1 |

Construction by example

Step 3. Use the Column-Change Variable Neighborhood Search algorithm (Vazquez et al., 2019) to sequentially minimize the F_4 vector or minimize the B_4 value of D .

- Two moves in D_l : (1) foldover columns
(2) swap columns

Example:

$$F_4(1, 0.5) = (2, 16)$$

Swap columns 6 and 7

$D =$

| | | | | | | | | |
|----|----|----|----|----|----|----|----|----|
| -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 |
| -1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 |
| -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 |
| -1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 |
| -1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 |
| -1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 |
| -1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 |
| -1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 |
| -1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 |
| -1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 |
| -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 |
| -1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 |
| -1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 |
| -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 |
| -1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | -1 | -1 | -1 | -1 | -1 | 1 | 1 |
| 1 | -1 | -1 | -1 | 1 | 1 | -1 | -1 | -1 |
| 1 | -1 | -1 | 1 | -1 | -1 | 1 | -1 | -1 |
| 1 | -1 | -1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | 1 | -1 | -1 | 1 | 1 | 1 | -1 |
| 1 | -1 | 1 | -1 | 1 | -1 | 1 | -1 | 1 |
| 1 | -1 | 1 | 1 | -1 | 1 | -1 | -1 | 1 |
| 1 | -1 | 1 | 1 | 1 | -1 | -1 | 1 | -1 |
| 1 | 1 | -1 | -1 | -1 | 1 | 1 | -1 | 1 |
| 1 | 1 | -1 | -1 | 1 | -1 | 1 | 1 | -1 |
| 1 | 1 | -1 | 1 | -1 | 1 | -1 | 1 | -1 |
| 1 | 1 | -1 | 1 | 1 | -1 | -1 | -1 | 1 |
| 1 | 1 | 1 | -1 | -1 | -1 | -1 | -1 | -1 |
| 1 | 1 | 1 | -1 | 1 | 1 | -1 | 1 | 1 |
| 1 | 1 | 1 | 1 | -1 | -1 | 1 | 1 | 1 |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | -1 | -1 |

Construction by example

Step 3. Use the Column-Change Variable Neighborhood Search algorithm (Vazquez et al., 2019) to **sequentially minimize the F_4 vector or minimize the B_4 value of D .**

- Two moves in D_l : (1) foldover columns
(2) swap columns

Example:

$$F_4(1, 0.5) = (0, 24)$$

Swap columns 7 and 8

$D =$

| | | | | | | | | |
|----|----|----|----|----|----|----|----|----|
| -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 |
| -1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 |
| -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 |
| -1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 |
| -1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 |
| -1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 |
| -1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 |
| -1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 |
| -1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 |
| -1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 |
| -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 |
| -1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 |
| -1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 |
| -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 |
| -1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | -1 | -1 | -1 | -1 | 1 | -1 | 1 |
| 1 | -1 | -1 | -1 | 1 | 1 | -1 | -1 | -1 |
| 1 | -1 | -1 | 1 | -1 | -1 | -1 | 1 | -1 |
| 1 | -1 | -1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | 1 | -1 | -1 | 1 | 1 | 1 | -1 |
| 1 | -1 | 1 | -1 | 1 | -1 | -1 | 1 | 1 |
| 1 | -1 | 1 | 1 | -1 | 1 | -1 | -1 | 1 |
| 1 | -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 |
| 1 | 1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 |
| 1 | 1 | -1 | -1 | 1 | -1 | 1 | 1 | -1 |
| 1 | 1 | -1 | 1 | -1 | 1 | 1 | -1 | -1 |
| 1 | 1 | -1 | 1 | 1 | -1 | -1 | -1 | 1 |
| 1 | 1 | 1 | -1 | -1 | -1 | -1 | -1 | -1 |
| 1 | 1 | 1 | -1 | 1 | 1 | 1 | -1 | 1 |
| 1 | 1 | 1 | 1 | -1 | -1 | 1 | 1 | 1 |
| 1 | 1 | 1 | 1 | 1 | 1 | -1 | 1 | -1 |

Construction by example

Step 3. Use the Column-Change Variable Neighborhood Search algorithm (Vazquez et al., 2019) to **sequentially minimize the F_4 vector** or **minimize the B_4 value of D .**

- Two moves in D_l : (1) foldover columns
(2) swap columns

Example:

$$F_4(1, 0.5) = (0, 24)$$

$D^* =$

| | | | | | | | | |
|----|----|----|----|----|----|----|----|----|
| -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 | -1 |
| -1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 | 1 |
| -1 | -1 | -1 | 1 | -1 | 1 | -1 | 1 | 1 |
| -1 | -1 | -1 | 1 | 1 | 1 | 1 | -1 | -1 |
| -1 | -1 | 1 | -1 | -1 | 1 | 1 | -1 | 1 |
| -1 | -1 | 1 | -1 | 1 | 1 | -1 | 1 | -1 |
| -1 | -1 | 1 | 1 | -1 | -1 | 1 | 1 | -1 |
| -1 | -1 | 1 | 1 | 1 | -1 | -1 | -1 | 1 |
| -1 | 1 | -1 | -1 | -1 | 1 | 1 | 1 | -1 |
| -1 | 1 | -1 | -1 | 1 | 1 | -1 | -1 | 1 |
| -1 | 1 | -1 | 1 | -1 | -1 | 1 | -1 | 1 |
| -1 | 1 | -1 | 1 | 1 | -1 | -1 | 1 | -1 |
| -1 | 1 | 1 | -1 | -1 | -1 | -1 | 1 | 1 |
| -1 | 1 | 1 | -1 | 1 | -1 | 1 | -1 | -1 |
| -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 | -1 |
| -1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | -1 | -1 | -1 | -1 | 1 | -1 | 1 |
| 1 | -1 | -1 | -1 | 1 | 1 | -1 | -1 | -1 |
| 1 | -1 | -1 | 1 | -1 | -1 | -1 | 1 | -1 |
| 1 | -1 | -1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 1 | -1 | 1 | -1 | -1 | 1 | 1 | 1 | -1 |
| 1 | -1 | 1 | -1 | 1 | -1 | -1 | 1 | 1 |
| 1 | -1 | 1 | 1 | -1 | 1 | -1 | -1 | 1 |
| 1 | -1 | 1 | 1 | 1 | -1 | 1 | -1 | -1 |
| 1 | 1 | -1 | -1 | -1 | 1 | -1 | 1 | 1 |
| 1 | 1 | -1 | -1 | 1 | -1 | 1 | 1 | -1 |
| 1 | 1 | -1 | 1 | -1 | 1 | 1 | -1 | -1 |
| 1 | 1 | -1 | 1 | 1 | -1 | -1 | -1 | 1 |
| 1 | 1 | 1 | -1 | -1 | -1 | -1 | -1 | -1 |
| 1 | 1 | 1 | -1 | 1 | 1 | 1 | -1 | 1 |
| 1 | 1 | 1 | 1 | -1 | -1 | 1 | 1 | 1 |
| 1 | 1 | 1 | 1 | 1 | 1 | -1 | 1 | -1 |

Outline

1. Tuberculosis inhibition experiment
2. Criteria to evaluate designs
3. Construction method
4. Results and discussion

Alternative 14-factor designs for the tuberculosis inhibition experiment

First entry of F_4 vector

| Parent Designs (D_u and D_l) | Run Size | Maximum Absolute Correlation Among Interactions | # Pairs of Interactions Involved | B_4 value |
|---|----------|---|--|-------------|
| Good 40-run strength-3 designs (Schoen et al., 2010). | 80 | 0.60 | 3 | 17.56 |
| | | 0.40 | 48 | 19.16 |
| Good 48-run strength-3 designs (Schoen and Mee, 2012). | 96 | 0.34 | 63 | 8.33 |
| | | 0.34 | 24 | 10.06 |
| 56-run strength-3 designs obtained by dropping columns from the folded-over 28-run Plackett-Burman design... | 112 | 0.43 | 3 | 11.00 |
| | | 0.29 | 129 | 11.86 |

Alternative 14-factor designs for the tuberculosis inhibition experiment

First entry of F_4 vector

| Parent Designs | Run Size | Maximum Absolute Correlation Among Interactions | # Pairs of Interactions Involved | B_4 value |
|--|----------|---|----------------------------------|-------------|
| Good 48-run strength-3 designs (Schoen and Mee, 2012). | 96 | 0.34 | 63 | 8.33 |
| | | 0.34 | 24 | 10.06 |
| Benchmark strength-3 design | | | | |
| Vazquez and Xu (2019) | 96 | 0.34 | 639 | 23.67 |

Simulation study

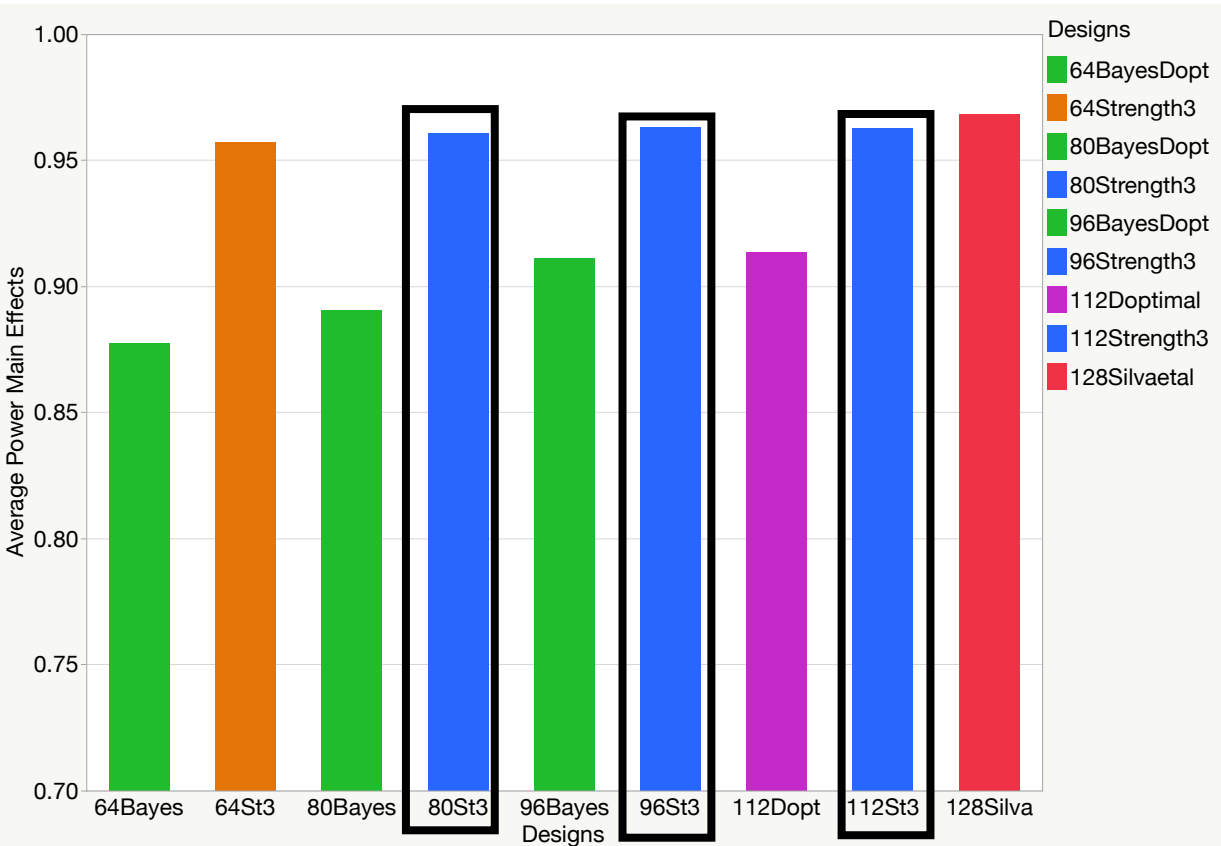
- Using simulations, we evaluate the performance of our 14-factor designs. We restrict our attention to those with best F_4 vector.
- We also consider two-level benchmark designs in the literature:
 - Strength-3 design with 64 runs (Xu & Wong, 2007)
 - Bayesian D-optimal designs with 64, 80, 96 and 112 runs (DuMouchel & Jones, 1994).
 - D-optimal design with 112 runs (Atkinson et al., 2007).
 - Strength-4 design with 128 runs (Silva et al. , 2016).

Simulation protocol

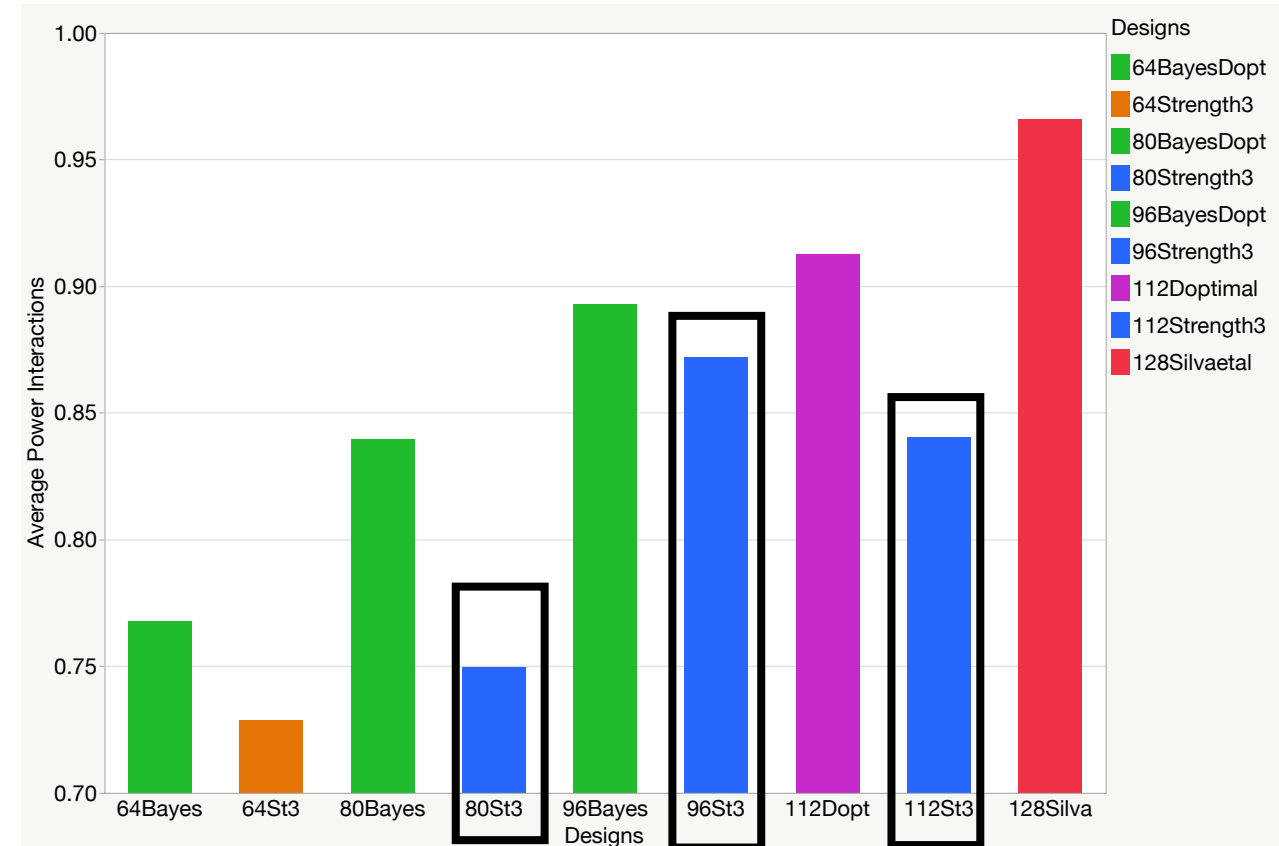
1. We randomly selected 8 main effects and 6 two-factor interactions (satisfying weak effect heredity) as active.
2. We obtained coefficients for the active effects by adding a 1 to an exponentially distributed random number. A '+' or '-' was randomly assigned.
3. The coefficients for the inactive effects were drawn from $N(0, 0.25^2)$.
4. We simulated response vectors with residuals following $N(0, 1)$.
5. We used the Dantzig selector (Candes & Tao, 2007) to identify the active effects.

1,000 simulations

Main Effects



Two-Factor Interactions



Power: Proportion of active effects that are successfully detected.

(Mee, Schoen & Edwards, 2017; Eendebak & Schoen, 2017; Vazquez & Xu, 2019)

Discussion

- For active effects larger than 2, our simulations show that all 80-, 96- and 112-run designs have powers close to 1.
- Using our methodology, we generated a new collection of attractive two-level strength-3 designs with 80, 96 and 112 runs and up to 29 factors.
- Constructing good two-level strength-3 designs with 72, 88, 104 and 120 runs is an interesting topic for future research.

Vazquez, A. R., Schoen, E. D., and Goos, P. (2021). Two-level orthogonal screening designs with 80, 96 and 112 runs, and up to 29 factors. *Journal of Quality Technology*. Published online.