

Central Composite Designs - Axial Points

- A design is called **rotatable** if $\text{Var}(\hat{y}(\mathbf{x}))$ depends only on $\|\mathbf{x}\| = (x_1^2 + \dots + x_k^2)^{1/2}$,
estimate $\rightarrow [x^(X^T X)^{-1} x^*] \hat{\sigma}^2$*
RSM interested in \hat{y} *$(1, x_1, \dots, x_k, x_1 x_2, \dots, x_1 x_k)$*

that is, if the accuracy of prediction of the response is the same on a sphere around the center of the design.

- For a central composite design whose factorial portion is a 2^{k-p} design of resolution V, it can be shown that rotatability is guaranteed by choosing

It causes orthogonality between LME \leftrightarrow LME, 2fi \leftrightarrow 2fi, LME \leftrightarrow 2fi,

$$\alpha = \sqrt[4]{n_f}. \quad (5)$$

\approx # of cube pts.

- Eqn. (5) serves as a useful guide even when the factorial portion does not have resolution V.

Central Composite Designs - Axial Points



- The rotatability criterion should not, however, be strictly followed.

- Its definition depends on how the x_i variables are coded. This lack of invariance with respect to the coding scheme makes it less attractive as a theoretical criterion.
- Because rotatability often requires that the factorial portion has resolution V, use of a smaller composite design whose factorial portion has a lower resolution has not been popular. However, some of the central composite designs whose factorial portions have resolution V or higher have excessive degrees of freedom.

MA 2⁶⁻²

have resolution IV

- For $k = 6$, the smallest design that satisfies the rotatability requirement is a 2^{6-1}_{VI} design for the factorial portion. *run size: $2^{6-1} + 2 \times 6 + 1 = 45$*

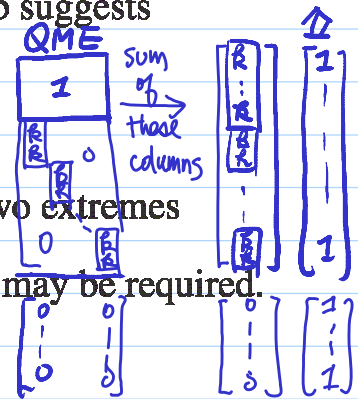
- The total run size of the central composite design is 45, which has 17 additional degrees of freedom over the minimal run size.

- Near rotatability** will be a more reasonable and practical criterion.

Central Composite Designs - Center Points

$$\widehat{\text{Var}}(\hat{y}_x) = \underbrace{[x^*{}^T (X^T X)^{-1} x^*]}_{\text{matrix}} \cdot \underbrace{\sigma^2}_{\text{variance}}$$

- The center points can
 - provide information on the pure error and → replicates
 - help stabilize the variance of the predicted response \hat{y} .
- For $\alpha = \sqrt{k}$, at least one center point is required for the estimability of the parameters in the second-order model. (Why?) Otherwise the variance of $\hat{y}(\mathbf{x})$ becomes infinite.
- To stabilize the prediction variance, the rule of thumb suggests
 - 3 to 5 runs at the center point when $\alpha \approx \sqrt{k}$, cube pts
 - 1 or 2 runs at the center point when $\alpha \approx 1$, axial pts
 - 2 to 4 runs should be considered between these two extremes
- To estimate the error variance, more than 4 or 5 runs may be required.



✓ **Reading:** textbook, 10.7

Box-Behnken Designs

- A family of three-level second-order designs
- Combining two-level factorial designs with balanced (or partially balanced) incomplete block designs in a particular manner
- Illustration: A BIBD with 3 treatments and 3 blocks of size 2

	Treatment		Factors
Block	1 2 3		x_1 x_2 x_3
1	× × ○	1	±1 ±1 0
2	× ×	2	±1 0 ±1
3	× ×	3	0 ±1 ±1
		4	0 0 0

1 within
1 between
block
comparison.

run size comparison

$$R = \begin{array}{cc|c} 3, & 12+nc & 12+nc \\ 4, & 6 \times 2^2 + nc & 24+nc \\ 5, & 10 \times 2^2 + nc & 40+nc \end{array}$$

- run size = $3 \times 2^2 + n_c$

- Why is it a second-order design?

25c QME

$x_1 x_2$	x_1^2	x_2^2	x_3^2
$\begin{bmatrix} 1 \\ 1 \\ -1 \\ -1 \end{bmatrix}$	$\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$	$\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$	$\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$
~ 0	~ 0	~ 0	~ 0
~ 0	~ 0	~ 0	~ 0
~ 0	~ 0	~ 0	~ 0

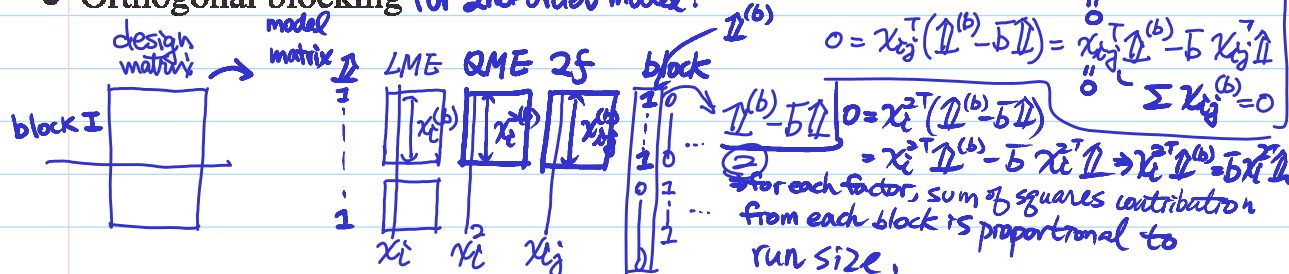
LME ← why estimable?
2fi ← why not aliased?
OME ←

Box-Behnken Designs - Some Properties

- Only require three levels of each factor (c.f., CCDs require 5 levels when $\alpha \neq 1$)
- All the design points (except the center) have length two, i.e., they all lie on the same sphere \leftarrow *Spherical design.*
 - These designs are particularly used for spherical regions
 - Because of the spherical property, there should be at least three to five runs at the center point

- The design for $k = 4$ is rotatable and the other designs are nearly rotatable

- Orthogonal blocking for 2nd-order model.



Box-Behnken Designs - Orthogonal Blocking

- For $k = 4, 5$, the designs can be arranged in orthogonal blocks
- $k = 4$ can be arranged in 3 blocks

BIBD

1	2	3	4
x	x		
x		x	
x			x
	x	x	
	x		x
		x	x

6 block

± 1	± 1	0	0
0	0	± 1	± 1
0	0	0	0
± 1	0	0	± 1
0	± 1	± 1	0
0	0	0	0
± 1	0	± 1	0
0	± 1	0	± 1
0	0	0	0

LME \rightarrow OK

2FI $\rightarrow x_1 x_2^{(b)}$ OK, $x_1 x_3^{(b)} = 2$ OK.

QME \rightarrow OK.

- $k = 5$, can be arranged in 2 blocks

BIBD

1	2	3	4	5
x	x			
x		x		
x			x	
	x	x		
	x		x	
		x	x	
		x		x
			x	x

± 1	± 1	0	0	0
0	0	± 1	± 1	0
0	± 1	0	0	± 1
± 1	0	0	± 1	0
0	0	0	± 1	± 1
0	0	0	0	0
± 1	0	± 1	0	0
0	± 1	0	0	± 1
0	0	± 1	0	± 1
0	0	0	0	0

Check (exercise)

LME

2FI

QME

Uniform Shell Designs

- The designs have $k^2 + k + 1$ points, which consist of
 - $k^2 + k$ points uniformly spaced on the k -dimensional sphere, and
 - one center point
- For $k = 2$, it can be described by a regular hexagon plus the center point

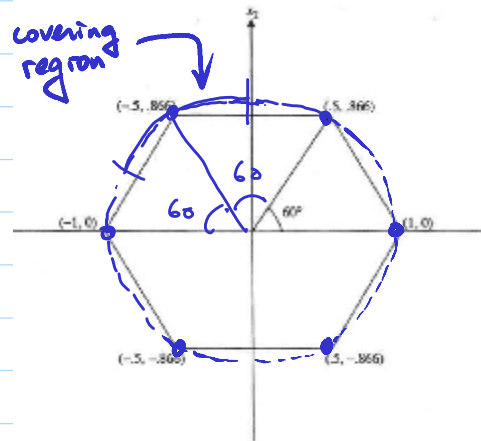


Figure 10: Uniform shell design, $k=2$

- For $k = 3$, it is identical to the Box-Behnken design

Uniform Shell Designs

- Uniformity property
 - Covering region of a design point: collection of points on the sphere that are closer to this design point than to the other design points.
 - Uniformity ensures that the covering region of each design point has the same spherical area.

- Run size comparison

k	$k^2 + k + 1$	$(k+1)(k+2)/2$
2	7	6
3	13	10
4	21	15
5	31	21

- For $k > 5$, its run size is much larger than $(k+1)(k+2)/2$
- For $2 \leq k \leq 5$, the designs are given in Table 10C.1 (textbook)
 - These designs can be rotated ^{without} affecting the uniformity property. However, rotation will likely increase the number of levels.
 - In the table, the design points minimize the number of the levels
- A disadvantage of spherical designs: lack of coverage of the region between the origin and the spherical shell
 - remedy: choose 2 concentric spherical shells of distinctly different radii.

Analysis Strategies for Multiple Responses *eg.*

RPD \rightarrow location model
dispersion model

- Multiple response analysis starts with building a regression fitted model for each response separately
- Some analysis strategies *study each response separately & find compromise solution.*
 - Overlaid contour plots (if there are few responses and only two or three important input variables.) *max min \hat{y}_1 subject to $a < \hat{y}_2 < b$
 $c < \hat{y}_3$*
 - Constrained optimization (if, among the responses, one is of primary importance)
 - Desirability function (if the goal is to find an optimal balance among several response characteristics)
- The desirability function approach transform each predicted response \hat{y}_i to a desirability value d_i , where $0 \leq d_i \leq 1$
 - The value of the **desirability function** d increases as the “desirability” of the corresponding response increases.