Choices for Alpha in a Split-Plot CCD

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Central Composite Design



Common Choices for Alpha

Cuboidal Region (Face Centered Cube)

 $\alpha = 1$

• Rotatable (Prediction Variance)

$$\alpha = \sqrt[4]{F}$$

• Orthogonal Blocking (Sequential Design)

$$\alpha = \sqrt{\frac{F(2k + n_{0A})}{2(F + n_{0F})}}$$

Split-Plot CCD

- There are two types of factors
 - Hard-to-change factors
 - Easy-to-change factors
- There is now alpha and beta axial distances

WP	W ₁	W ₂	S ₁	S ₂	WP	W ₁	W ₂	S ₁	S ₂
	-1	-1	-1	-1		α	0	0	0
1	-1	-1	1	-1	6	α	0	0	0
	-1	-1	-1	1		α	0	0	0
	-1	-1	-1	1		α	0	0	0
	1	-1	-1	-1		0	-α	0	0
2	1	-1	1	-1	7	0	-α	0	0
	1	-1	-1	1		0	-α	0	0
	1	-1	-1	1		0	-α	0	0
	-1	1	-1	-1		0	α	0	0
3	-1	1	1	-1	8	0	α	0	0
	-1	1	-1	1		0	α	0	0
	-1	1	-1	1		0	α	0	0
	1	1	-1	-1		0	0	-β	0
4	1	1	1	-1	9	0	0	β	0
	1	1	-1	1		0	0	-β	0
	1	1	-1	1		0	0	β	0
	-α	0	0	0		0	0	0	-β
5	-α	0	0	0	10	0	0	0	β
	-α	0	0	0		0	0	0	-β
	-α	0	0	0		0	0	0	β

Cuboidal WP Design

- Using alpha=1 makes less settings for the hardto-change factors
- Can be used to create minimum whole plot designs
- Beta is not constrained here

Rotatability

- Consider the model $\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \mathbf{M}\boldsymbol{\gamma} + \boldsymbol{\varepsilon}$
- Consider partitioning the design matrix as

$$\mathbf{X} = \begin{bmatrix} \mathbf{1}_{1} & \mathbf{W}_{Q_{1}} & \mathbf{S}_{Q_{1}} & \mathbf{W}_{F_{1}} & \mathbf{S}_{F_{1}} \\ \mathbf{1}_{2} & \mathbf{W}_{Q_{2}} & \mathbf{S}_{Q_{2}} & \mathbf{W}_{F_{2}} & \mathbf{S}_{F_{2}} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ \mathbf{1}_{m} & \mathbf{W}_{Q_{m}} & \mathbf{S}_{Q_{m}} & \mathbf{W}_{F_{m}} & \mathbf{S}_{F_{m}} \end{bmatrix} = \begin{bmatrix} \mathbf{W}^{*} & \mathbf{S}_{F} \end{bmatrix}$$

• Then: $\mathbf{X}'\Sigma^{-1}\mathbf{X} = \begin{bmatrix} \frac{1}{\sigma^2 + n\sigma_{\delta}^2} (\mathbf{W}'^*\mathbf{W}^*) & \mathbf{0} \\ \mathbf{0} & \frac{1}{\sigma^2} (\mathbf{S}'_F\mathbf{S}_F) \end{bmatrix}$

Rotatability

 Let's look at the pure second-order moment for the ith subplot factor. A rotatable design would require

$$\begin{bmatrix} ii \end{bmatrix}_{sp}^{S_F} = \begin{bmatrix} ii \end{bmatrix}_{sp}^{W^*}$$

However, there is a multiplier of $\frac{1}{\sigma^2}$ in S_F and $\frac{1}{\sigma^2 + n\sigma_{\delta}^2}$ in W*.

Unless $\sigma_{\delta}^2 = 0$, we cannot achieve exact rotatability.

- We propose a concept of partial rotatability. It has similar properties, and it is more logical in the split-plot case.
- In a two-strata rotatable design, there are two axial distances instead of one. The whole plot axial distance, ρ_W^2 and the second is the subplot axial distance, ρ_s^2
- A two-strata rotatable design will allow all the points with same ρ_w^2 and ρ_s^2 to have same prediction variance.

- In a rotatable design, $Var[\hat{y}(\mathbf{x}_0)] = f(\rho_W^2 + \rho_S^2)$ which is a function of the overall distance.
- In a two-strata rotatable design, $Var[\hat{y}(\mathbf{x}_0)] = f(\rho_W^2, \rho_S^2)$ a function of the two separate distances.
- Therefore, the stronger assumption in rotatable designs is being relaxed to accommodate the split-plot structure.

We define the two-strata rotatable designs as:

In a k-factor split-plot design with k_1 WP factors and k_2 SP factors, let z_i be the *i*th WP factor and x_j be the *j*th SP factor. If for any two design runs

$$t_{0} = (z_{10}, z_{20}, \dots, z_{k1,0}, x_{10}, \dots, x_{k2,0}) \text{ and } t_{1} = (z_{-11}, z_{21}, \dots, z_{k1,1}, x_{11}, \dots, x_{k2,1}) \text{ satisfying}$$

$$\sum_{i=1}^{k_{1}} z_{i0}^{2} = \sum_{i=1}^{k_{1}} z_{i1}^{2} = \rho_{W}^{2} \text{ and } \sum_{j=1}^{k_{2}} x_{j0}^{2} = \sum_{j=1}^{k_{2}} x_{j1}^{2} = \rho_{S}^{2}$$

the corresponding prediction variances $Var[\hat{y}(t_0)] = Var[\hat{y}(t_1)]$ are equal, the design is called a two-strata rotatable split-plot design.

• The values of alpha and beta are

WP	W ₁	W ₂	S ₁	s ₂
	0	0	-β	0
9	0	0	β	0
	0	0	-β	0
	0	0	β	0
	0	0	0	-β
10	0	0	0	β
	0	0	0	-β
	0	0	0	β

$$\alpha = \sqrt[4]{\frac{F}{n}}$$
 and $\beta = \sqrt[4]{\frac{2(1+nd)F}{n}}$

WP	W ₁	W ₂	S ₁	S ₂
	0	0	-β	0
9	0	0	β	0
	0	0	0	-β
	0	0	0	β

$$\alpha = \sqrt[4]{\frac{F}{n}}$$
 and $\beta = \sqrt[4]{\frac{4F}{n}}$

General conditions for these designs



where r = # alphas in a column and a = # betas

- What if we had a CRD?
- Then d=0, n=1, r=2, a=2

$$\alpha = \sqrt[4]{\frac{2F}{r}} = \sqrt[4]{F} \text{ and } \beta = \sqrt[4]{\frac{2(1+nd)F}{a}} = \sqrt[4]{F}$$
$$\alpha = \sqrt[4]{\frac{2F}{r}} = \sqrt[4]{F} \text{ and } \beta = \sqrt[4]{\frac{2F}{a}} = \sqrt[4]{F}$$

• So they all reduce to the CRD rotatable value

- Conditions can be generalized to other combinations of HTC and ETC factors
- Unbalance in the WPs with axial runs can be handled, then there are subscripts on the *n*
- Only condition is $\mathbf{1'S}_{D_i} = \mathbf{0'}$

• Condition for Orthogonal Blocking:

For any column of X, the average in block i is the equal to the average in block j is equal to the overall average

 Khuri (1992) and Goos (2002) show that for an OB design, OLS=GLS for a balanced fixed block design

- Extend this concept to SP-CCD
- Not going to show proof
- Block sizes do not have to be equal, but the whole plot sizes need to be equal

WP	W ₁	W ₂	S ₁	S ₂	WP	W ₁	W ₂	S ₁	S ₂
	-1	-1	-1	-1		α	0	0	0
1	-1	-1	1	-1	6	α	0	0	0
	-1	-1	-1	1		α	0	0	0
	-1	-1	-1	1		α	0	0	0
	1	-1	-1	-1		0	-α	0	0
2	1	-1	1	-1	7	0	-α	0	0
	1	-1	-1	1		0	-α	0	0
	1	-1	-1	1		0	-α	0	0
	-1	1	-1	-1		0	α	0	0
3	-1	1	1	-1	8	0	α	0	0
	-1	1	-1	1		0	α	0	0
	-1	1	-1	1		0	α	0	0
	1	1	-1	-1		0	0	-β	0
4	1	1	1	-1	9	0	0	β	0
	1	1	-1	1		0	0	0	-β
	1	1	-1	1		0	0	0	β
	-α	0	0	0		0	0	0	0
5	-α	0	0	0	10	0	0	0	0
	-α	0	0	0		0	0	0	0
	-α	0	0	0		0	0	0	0

- Sequential Design will have WPs 1-4 in Block 1 and WPs 5-9 in Block 2
- For OB, trying to solve for the number of WP 10s to add into Blocks 1 & 2 as well as find the values for alpha and beta
- Only need to focus on quadratics

- F = # factorial points
- $k_1 = #$ WP factors
- $k_2 = #$ SP factors
- n_{c1} = # overall center runs added to B1
- $n_{c2} = #$ overall center runs added to B2
- r = # alphas in a column
- a = number of betas in a column

Orthogonal Blocking $\frac{F}{F+n_{c1}} = \frac{r\alpha^{2}}{k_{1}r+k_{2}a+n_{c2}} = \frac{r\alpha^{2}+F}{k_{1}r+k_{2}a+n_{c2}+F+n_{c1}}$ $\frac{F}{F+n_{c1}} = \frac{a\beta^{2}}{k_{1}r+k_{2}a+n_{c2}} = \frac{a\beta^{2}+F}{k_{1}r+k_{2}a+n_{c2}+F+n_{c1}}$

$$\alpha^{2} = \frac{(k_{1}r + k_{2}a + n_{c2})F}{r(F + n_{c1})}$$
$$\beta^{2} = \frac{(k_{1}r + k_{2}a + n_{c2})F}{a(F + n_{c1})}$$

- There is a tradeoff between the size of the overall design, size of the blocks and reasonable values for alpha and beta
- With n_{c1} =8 and n_{c2} =4, our example yields:

 $\alpha = 1.414, \beta = 2.828$

• With one replicate of the beta WP, gives

 $\alpha = 1.53, \beta = 2.16$

Summary

- When conducting a CCD, the axial distance must be chosen
- In a SP-CCD, there are now two distances
- Simple solution is to use a distance of 1
- Cannot achieve the general definition of rotatability but can use partial rotatability
- OLS-GLS designs can be OB