

## Complex Aliasing

- Alias matrix

- True model:  $\mathbf{Y} = \mathbf{X}_1\boldsymbol{\beta}_1 + \mathbf{X}_2\boldsymbol{\beta}_2 + \boldsymbol{\varepsilon}$ , with  $E(\boldsymbol{\varepsilon}) = \mathbf{0}$  and  $Var(\boldsymbol{\varepsilon}) = \sigma^2\mathbf{I}$ , where  $\mathbf{I}$  is the identity matrix.
- Working model:  $\mathbf{Y} = \mathbf{X}_1\boldsymbol{\beta}_1 + \boldsymbol{\varepsilon}$ , where  $\mathbf{X}_1$  is nonsingular.
- The ordinary least squares estimator of  $\boldsymbol{\beta}_1$  under the working model is

$$\hat{\boldsymbol{\beta}}_1 = (\mathbf{X}_1^T \mathbf{X}_1)^{-1} \mathbf{X}_1^T \mathbf{Y}$$

- Under the true model,

$$\begin{aligned} E(\hat{\boldsymbol{\beta}}_1) &= (\mathbf{X}_1^T \mathbf{X}_1)^{-1} \mathbf{X}_1^T E(\mathbf{Y}) \\ &= (\mathbf{X}_1^T \mathbf{X}_1)^{-1} \mathbf{X}_1^T (\mathbf{X}_1\boldsymbol{\beta}_1 + \mathbf{X}_2\boldsymbol{\beta}_2) \\ &= \boldsymbol{\beta}_1 + (\mathbf{X}_1^T \mathbf{X}_1)^{-1} \mathbf{X}_1^T \mathbf{X}_2\boldsymbol{\beta}_2 \end{aligned}$$

where  $(\mathbf{X}_1^T \mathbf{X}_1)^{-1} \mathbf{X}_1^T \mathbf{X}_2\boldsymbol{\beta}_2$  is the bias of  $\hat{\boldsymbol{\beta}}_1$  under the true model.

- We call  $L = (\mathbf{X}_1^T \mathbf{X}_1)^{-1} \mathbf{X}_1^T \mathbf{X}_2$  the **alias matrix** because it provides the aliasing coefficients for the estimate  $\hat{\boldsymbol{\beta}}_1$ .

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## Complex Aliasing

- Alias matrix - example

- Consider the  $OA(12, 2^7)$  used in the cast fatigue experiment.
- Suppose that  $\mathbf{X}_1$  consists of the 7 main effects.
- Suppose that  $\mathbf{X}_2$  consists of the  $\binom{7}{2} = 21$  two-factor interactions.
- The true model is unidentifiable.
- Under the working model, the main effects estimates are uncorrelated and unbiased. The main effect estimate (say, of  $A$ ) is

$$\hat{A} = \bar{y}(A = +) - \bar{y}(A = -)$$

- Under the true model,

$$\begin{aligned} E(\hat{A}) &= A - \frac{1}{3}BC - \frac{1}{3}BD - \frac{1}{3}BE + \frac{1}{3}BF - \frac{1}{3}BG \\ &\quad + \frac{1}{3}CD - \frac{1}{3}CE - \frac{1}{3}CF + \frac{1}{3}CG + \frac{1}{3}DE \\ &\quad + \frac{1}{3}DF - \frac{1}{3}DG - \frac{1}{3}EF - \frac{1}{3}EG - \frac{1}{3}FG \end{aligned}$$

- For the 12-run design, each main effect has the same property as that for factor  $A$  in that it is partially aliased with the 2-factor interactions not involving the main effect and the aliasing coefficients are  $1/3$  or  $-1/3$ .

## Complex Aliasing

- Two effects are said to be **partially aliased** if the absolute value of their aliasing coefficient is strictly between 0 and 1.
- The **complex aliasing** refers to the enormous number of effects in  $\mathbf{X}_2$  (e.g., 2-factor interactions) that are partially aliased with the effects in  $\mathbf{X}_1$  (e.g., main effects).
- By comparison, for  $2^{k-p}$  or  $3^{k-p}$  regular designs,
  - the factorial effects are either orthogonal (i.e., uncorrelated) or fully aliased (i.e., with aliasing coefficient 1 or  $-1$ ) with each other, and
  - the number of (full) aliases is much smaller.

❖ Reading: textbook, 9.1

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## Traditional Analysis Strategy

- For designs with complex aliasing, it is difficult to disentangle the large number of aliased effects and to interpret their significance.
- For this reason, such designs were traditionally used for *screening factor effects* only, i.e., to estimate factor main effects but not their interactions.
- These designs are therefore referred to as **screening designs**.
- The validity of the main effect estimates for screening designs depends on the assumption that the interaction effects are negligible, which is often questionable in many practical situations.
- It suggests the need for strategies that allow interactions to be entertained.

❖ Reading: textbook, 9.2

## Simplification of Complex Aliasing

- The complex aliasing pattern can be greatly simplified under the assumption of *effect sparsity*.
  - Effect sparsity: number of relatively important effects (e.g., main effects and two-factor interactions) is small.
  - By employing the effect sparsity and effect hierarchy principles, one can assume that only few main effects and even fewer two-factor interactions are relatively important.
- With this assumption, the complex aliasing pattern of a design can be greatly simplified. This simplification and the partial aliasing of effects make it possible to estimate some interactions.

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## Simplification of Complex Aliasing

- Consider the hypothetical example.
  - Suppose that in the cast fatigue experiment only  $A$ ,  $F$ ,  $G$ ,  $AB$ , and  $FG$  are significant.
  - If the seven main effect estimates were obtained under the main-effect-only model, their expectations in LNp.4-26 are simplified as follows:

$$E(\hat{A}) = A - \frac{1}{3}FG$$

$$E(\hat{B}) = -\frac{1}{3}FG$$

$$E(\hat{C}) = -\frac{1}{3}AB + \frac{1}{3}FG$$

$$E(\hat{D}) = -\frac{1}{3}AB + \frac{1}{3}FG \tag{1}$$

$$E(\hat{E}) = -\frac{1}{3}AB - \frac{1}{3}FG$$

$$E(\hat{F}) = F + \frac{1}{3}AB$$

$$E(\hat{G}) = G - \frac{1}{3}AB$$