

#### **Panel Discussion Points**

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#### "Classical Designs are Optimal"

- All two-level fractional factorial designs are optimal by every variance criterion – given an appropriate model...
- Latin Square, BIBs etc. are also D-optimal for the additive model...
- If all the specific design requirements are absolutely standard, then use textbook designs.
- But...

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#### Elegance vs. Practical Necessity

- Would you rather specify the number of runs you can do instead of having the sample size dictated to you?
- Do you ever need to avoid certain combinations of factor settings?
- Would you like to be able to augment a design without doubling the number of runs you have already done?
- Do you ever want to do response surface experiments with categorical factors?



### "Alphabet Soup" Criticism

- Optimal designs are limiting because they focus on one number (D, A, I, G)
- "Good" designs need to have several desirable characteristics.

## Box and Draper 14 Points

- 1. Generate satisfactory distribution of information.
- 2. Make unbiased predictions.
- 3. Detect lack of fit.
- 4. Allow transformations to be estimated.
- 5. Allow for blocking.
- 6. Allow for sequential design.
- 7. Provide and internal estimate of error.
- 8. Be robust to outliers and other violations of assumptions.

## Box and Draper 14 Points (cont.)

- 9. Require a minimum number of experimental runs.
- 10. Provide simple data patterns.
- **11**.Ensure simplicity of calculation.
- 12.Behave well when errors occur in the x settings.
- **13.**Don't use too many levels in the x settings
- 14. Provide a check on the constancy of variance assumption.



## **Conflicting Advice**

- Require a minimum number of experimental runs.
  versus
- 7. Provide an internal estimate of error.
- 3. Detect of lack of fit.
- 14. Check constancy of variance assumption.



### Alternative

Multi-criterion optimization.

Example: Model robust design

Li & Nachtsheim (2000) Technometrics



Matters of Style...

One size fits all.

versus

Build to suit.



### My own style...

- Learn as much as possible about the unique setting of each problem.
- 2. Ask questions to tease out constraints.
  - 1. Factor combination restrictions
  - 2. Block size restrictions
  - 3. Need to group runs to make fewer changes in some factors
  - 4. Budget
- 3. Rely on software to design and analyze a study that matches the requirements.

# Summary

- 1. Classical designs are optimal.
- 2. Classical designs are *not* flexible.
- 3. Optimal designs *can* make principled trade-offs among competing objectives.
- 4. It is preferable to build a design to suit the problem rather than alter the problem to match a design *and this is possible given modern software*.



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