# Experimental Design

Part 3: Statistical Design of Experiments



# Today's Quote

"Learning without thinking is useless. Thinking without learning is dangerous."

— Confucius

- Statistical Design of Experiments
  - Broad field you could take multiple classes in this
  - Today: a brief overview
- In any general process or system
  - Input
  - Controllable factors
  - Uncontrollable factors
  - Outputs
- Goals
  - •
  - •

#### Approaches

#### Best Guess Approach

- Can work with great deal of knowledge about system or if you get very lucky
- Disadvantages
  - May require many "best guesses" if at first you don't succeed
  - May get acceptable result, but not likely \_\_\_\_\_\_

#### One Factor at a Time Approach

- Set baseline levels for each controllable factor and vary one at a time
- Problem \_\_\_\_\_

#### Factorial Approach

Factors varied together, systematically

- Usually abbreviated as "DoE"
- Typical applications of DoE
  - Screen to determine important factors in a process
  - Process development
  - Process optimization
- Limitations of DoE approach

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## Basic Principles of DoE

#### Replication

- Repetition of basic experiment
- Necessary for \_\_\_\_\_\_\_

#### Randomization

Allocation of experimental material and order of experiments performed randomly

## Blocking

- A portion of experimental material that should be more homogeneous than the entire set of material
- [E.g., comparing the effect of two diets on N persons Block 1: All females, Block 2: All males; Block 1 + Block 2 = N.]

## Factorial Design

- Typically, most efficient for effect of 2 or more factors on output variable
- Complete trial or replicant of experiment at all levels of the factors
- Example: 2 factors A and B
  - a levels of A and b levels of B
  - *ab* total treatment levels
  - Factors said to be crossed
  - For 2 factors at 2 levels (high +; low -) 2<sup>2</sup> factorial

| <b>EXPERIMENT</b> | A | В |
|-------------------|---|---|
| 1                 | - | - |
| 2                 | + | - |
| 3                 | - | + |
| 4                 | + | + |

## Factorial Design

Factorial design leads to a regression model:

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_{12} x_1 x_2 + \varepsilon$$

- Parameters to be determined are  $\beta$ 's
  - Note the interaction  $\beta_{12}$ 
    - Interaction occurs when the effect of one variable depends on the level of another variable
- $x_1$  represents A,  $x_2$  represents B
- Random, unexplained error is arepsilon
- Model accuracy from \_\_\_\_\_\_
  - Residual is difference between model prediction and measured data point
  - Other statistical measures of error can also be calculated

# Sample Size

- How many replicants?
  - As # or replicants increases, likelihood of error decreases
  - Decrease in error probability is nonlinear
  - Dependent on
    - \_\_\_\_\_
    - •
    - •

# Blocking

Blocking provides some advantages although experiments may be designed to be randomized.

- Example
  - Clinical trials for a vaccine on 10,000 randomly selected volunteers
  - Separate total sample into two blocks of equal size male and female
  - Each becomes a block; choose members randomly within each block
  - Advantage: May provide data on whether vaccine efficacy is the same for males and females.
- Other common blocks
  - Time e.g., run N experiment per day for M days how do the results change (or do they)?
  - Individuals e.g., you run one set (replicant), someone else runs another set –do results change?
- Regression model with blocking term

$$y = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_{12} x_1 x_2 + \delta + \varepsilon$$

• Residual error associated with the blocking is  $\delta$ 

## **Blocking: Example**

- The results from a randomized experiment conducted at the American Economic Review on the effects of double-blind versus single-blind peer reviewing on acceptance rates and referee rating indicate that acceptance rates are lower and referees are more critical when the reviewer is unaware of the author's identity.
- These patterns are not significantly different between female and male authors. (What are the blocks?)
- Authors at top-ranked universities and at colleges and low-ranked universities are largely unaffected by the different reviewing practices. (What are the blocks?)
- The authors at near-top-ranked universities and at nonacademic institutions have lower acceptance rates under double-blind reviewing. (What are the blocks?)

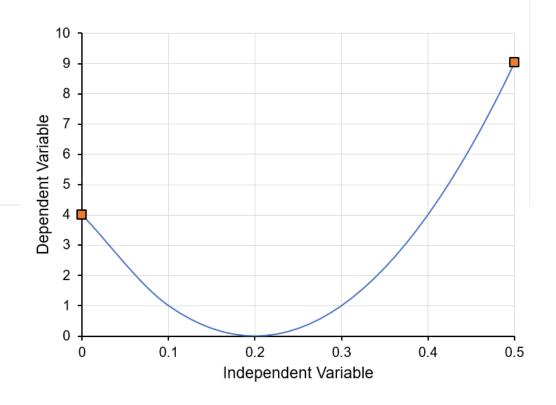
## Design Considerations

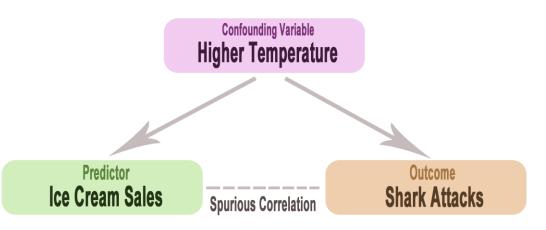
#### Curvature

- Linearity of factor effects assumed in 2<sup>k</sup> design
- Add center points high, medium, low
  - Check for curvature
  - Minimize additional experiments

## Confounding variables

 Observed correlation between A and B could be caused by a confounding variable C





### Full factorial design

- 2<sup>2</sup> design: 4 experiments per replicate
- 2<sup>6</sup> design: 64 experiments per replicate
- Numbers increase rapidly

## Fractional factorial Design

- Assume main effects and low order interactions most important
- Design such that fraction can be projected onto specific subset of factors
- Design such that combination of runs of 2 or more fractions can be assembled into a larger design

Useful as \_\_\_\_\_\_

#### Conclusions

- Several software packages available to set up and analyze statistically designed experiments
- Not widely done in academia little fundamental insights
- Extremely important in industry
- You may want to consider taking a class in this area if you haven't already!

#### For further reading:

- Douglas C. Montgomery, Design and Analysis of Experiments, 10<sup>th</sup> ed., John Wiley & Sons, Inc., New York, 2020
- Robert L. Mason, Richard F. Gunst, and James L. Hess, *Statistical Design and Analysis of Experiments With Applications to Engineering and Science 2<sup>nd</sup> ed.,* John Wiley & Sons, Inc., 2003 (Available through Knovel)

# The End