

Lecture 8

Introduction to Design of Experiments

Reading: Oehlert 2010 Chapters 1, 2; DAE 2017 Chapters 1, 2

DSA 8020 Statistical Methods II

Background and
Definitions

Fundamental
Principles of
Experimental Design

History of
Experimental Design

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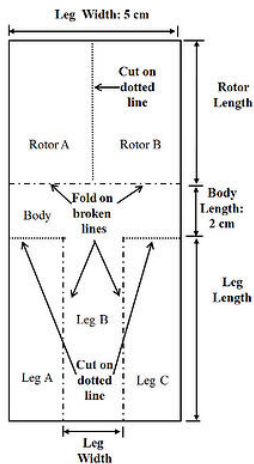
1 Background and Definitions

2 Fundamental Principles of Experimental Design

3 History of Experimental Design

Paper Helicopter Experiment

Suppose we want to investigate how long a paper helicopter can fly in the air.



Some of potential determining factors:

- Paper type (light, medium, and heavy)
- Rotor length (7.5cm or 8.5cm)
- Leg length (7.5cm or 12 cm)
- Leg width (3.2cm or 5cm)

Source:

<https://blog.minitab.com/en/learning-design-of-experiments-with-paper-helicopters-and-minitab>

Steps for Planning, Conducting and Analyzing an Experiment

- State the problem of interest
- Select the response variable and determine the factor(s)
- Choose the design and conduct the experiment
- Perform statistical analysis
- Draw conclusions

- **Specific question:** How do battery types vary with respect to life-per-unit cost?
- **Response:** Time (per unit cost) to exhaust battery under standard load
- **Comparative:** Difference between 4 battery types
- **Controlled:** All compared using the same device
- **Replication:** Four batteries of each type tested

- **Factor:** Variable whose influence upon the response variable. Settings of factor are called **levels**
- **Treatments:** The procedures (a set of values for all factors) used for comparison
- **Experimental units:** Objects on which treatments are applied
- **Measurement units:** Objects on which the response is measured. **These may differ from the experimental units**
- **Randomization:** Using a known probabilistic mechanism to assign treatments to experimental units
- **Experimental error:** Variation in response outcomes (modeled as random)

Main Elements of An Experiment

An **experiment** applies **treatments** to **experimental units** and measures **responses**.

- Want to learn about **treatments** (e.g., dose of drug; nano-tech coating for a fabric)
- **Responses** tell us how the treatment worked (patient get better; stain resistance)
- Experimenter **assigns** treatments to **experimental units** (e.g., a patient; a bolt of fabric)

- Perhaps the most important concept in statistical design
- The **experimental unit** is the unit (subject, plant, pot, animal) which is randomly assigned to a treatment
- The experimental unit *defines the unit to be replicated to increase degrees of freedom*

Experimental Units vs Measurement Units

If a group of “units” must have the same treatment, they are likely measurement units (MUs) rather than experimental units (EUs)

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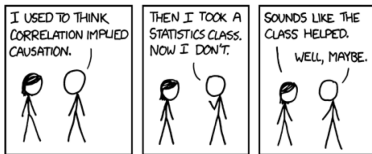


- Different food placed in tanks containing the fish. Fish are not the EUs



Observational vs. Experimental Studies

- An **observational study** has the same triple of treatment, unit, and response, but one **observes** the assignment of treatments to units (e.g., human health studies on cigarette smoke and adverse health effects)
- What makes an **experimental study** special is **control**. The experimenter gets to control the assignment of treatments to the experimental units
- Experiments can make **causal inference** while observational studies find **association**



Source: Slide 5 at <http://users.stat.umn.edu/~gary/classes/5303/lectures/Introduction.pdf>

Why Designed Experiments?

- Design for direct comparison of treatments
- Design to reduce bias in comparisons (avoid systematic errors)
- Design to reduce variability (be precise)
- Experiments support causal inference

Fundamental Principles: Replication, Randomization, and Blocking

- Each treatment is applied to (experimental) units that are representative of the population
- Enable the estimation/quantification of **experimental error** using standard deviation
- Decrease variance of estimates and increase the power to detect significant differences: for independent y_i 's,

$$\text{Var}\left(\frac{1}{n} \sum_{i=1}^n y_i\right) = \frac{1}{n} \text{Var}(y_1)$$

Use of a chance mechanism such as random number generators to assign treatments to (experimental) units. It has the following advantages:

- Protect against latent variables or “lurking” variables
- Reduce influence of subjective bias in treatment assignments (e.g., clinical trials)
- Ensure validity of statistical inference

A **block** refers to a collection of homogeneous units. Effective blocking: larger between-block variations than within-block variations.

Examples: hours, batches, lots.

- Run and compare treatments within the same blocks to eliminate block-block variation and reduce variability of treatment effects estimates
- Block what you can and randomize what you cannot

A Brief History of Experimental Design: Agricultural Era

- R. A. Fisher and his co-workers, Rothamsted Agricultural Experimental Station (1930, England)
- Introduced statistical experimental design and data analysis. Summarized the fundamental principles: **replication**, **randomization**, and **blocking**
- Factorial designs, ANOVA



"To consult the statistician after an experiment is finished is often merely to ask him to conduct a post mortem examination. He can perhaps say what the experiment died of."

Ronald Fisher

A Brief History of Experimental Design: Industrial Eras

- The first industrial era, 1951 - late 1970s
 - Process modeling and optimization
 - G. E. P. Box & K. B. Wilson and coworkers in chemical industries and other processing industries
 - Empirical modeling, response surface methodologies, central composite design
- The second industrial era, late 1970s - 1990
 - Quality improvement and variation reduction
 - G. Taguchi and robust parameter design

- Popular outside statistics, and an indispensable tool in many scientific/engineering endeavors
- New challenges:
 - Large and complex experiments, e.g., screening design in pharmaceutical industry, experimental design in biotechnology
 - **Computer experiments:** efficient ways to model complex systems based on computer simulation
 - ...

These slides cover:

- Basic concepts of design of experiments (DOE):
- A brief history of DOE
- Fundamental principles: randomization, blocking, replication

- Completely Randomized Designs
- Randomized Complete Block Designs, Factorial Designs, and Split-Plot Designs
- Random and Mixed Effects Models
- Computer Experiments