

# Chapter 5: One-way ANOVA and Randomized Experiments - Part 1

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# Key Topics

- ▶ Overview of ANOVA
- ▶ The One-Way Randomized Experiment
- ▶ Fitting One-Way ANOVA Model

# Overview of ANOVA

- ▶ Simple Linear Regression and Multiple Linear Regression:
  - ▶ Response variable: continuous
  - ▶ Predictor variable(s): at least one continuous variable
- ▶ In this chapter we will look into models with
  - ▶ Continuous response
  - ▶ Categorical predictor(s)
- ▶ Examples:
  - ▶ Weight gain among rats vs. source of protein in diet (beef, cereal or pork)
    - ▶ An experiment
  - ▶ Difference in average commute time among four cities (Boston, Houston, Minneapolis and Washington)
    - ▶ Observational study
- ▶ From an overall context we will be investigating whether there is a mean difference in response for each category

## Example 5.1 - Fat Rats: A Randomized Experiment

- ▶ Objective: to study the effect of three different high-protein diets on weight gain in baby rats
- ▶ Response: weight gain (continuous / quantitative)
- ▶ Predictor: sources of protein in high-protein diets
  - ▶ beef
  - ▶ cereal
  - ▶ pork
- ▶  $n = 30$ : 10 on each type of protein diet

## Fat rats: a randomized experiment. . .

- ▶ Is there a difference in weight gain depending on the different sources of diet?
- ▶ If different, how big is the difference?
- ▶ Are all groups different or subset of groups different?
- ▶ We can answer all these questions by performing an ANOVA

## Example 5.2 - Teen Pregnancy and the Civil War

- ▶ Response: Pregnancy rate (number of pregnancies per 1000 teenage girls) for each of 50 U.S. states in 2010
- ▶ Predictor: role of the state in the U.S. civil war
  - ▶ C: confederate state
  - ▶ B: border state
  - ▶ U: union state
  - ▶ O: other state
- ▶ Objective: investigate whether teen pregnancy differ depending on the role of the state during civil war (1860)

# One-way Randomized Experiment

- ▶ Simplest experimental structure for an ANOVA model
- ▶ Has a single categorical predictor, and we create the data for this choice of model by randomly assigning units to the categories of that variable
- ▶ One-way ANOVA is a regression model when we have one categorical predictor and continuous response

# One-way Randomized Experiment - Terminology

- ▶ Some terminology in Experimental Designs (example: Fat rats)
  - ▶ EXPERIMENTAL UNIT: each rat who get randomly assigned to one of the protein sources
  - ▶ EXPLANATORY FACTOR(S): one factor - source of protein
  - ▶ LEVELS/TREATMENTS: different levels of the factor (beef, cereal, pork)
  - ▶ BALANCED DESIGN: when each level/treatment gets assigned to same number of experimental units (10 rats per level)
  - ▶ RESPONSE VARIABLE: measurement of each experimental unit after the treatment applied (weight gain of rats)
- ▶ One-way ANOVA can also be used with observation studies as long as the observations are gathered randomly



# Fitting One-Way ANOVA model

- ▶ For quantitative response  $Y$  and single categorical predictor,

$$Y = \mu_i + \epsilon$$

or

$$Y = \mu + \alpha_i + \epsilon$$

- ▶  $\mu$ : Grand mean
- ▶  $\alpha_i$ : treatment effect for  $i^{th}$  group mean
- ▶ Random error:  $\epsilon \sim N(0, \sigma_\epsilon)$
- ▶ Mean for the  $i^{th}$  group:  $\mu_i = \mu + \alpha_i$

# Fitting One-Way ANOVA Model

- ▶ The goal of ANOVA is to compare the averages of each category

Response = Grand Average + Treatment effect + Residual

- ▶ Objective is to answer following questions
  1. What is the overall average: Grand mean ( $\hat{\mu} = \bar{y}$ )
  2. How far is each average from overall average: treatment effect ( $\hat{\alpha}_i = \bar{y}_i - \bar{y}$ )
  3. How far is each response from its group average: residual ( $y_i - \bar{y}_i$ )

# ANOVA Table

► For  $k$  groups

Source	df	SS	MS	F
Groups	$k-1$	$SS_{Groups}$	$MS_{Groups} = \frac{SS_{Groups}}{k-1}$	$F = \frac{MS_{Groups}}{MSE}$
Residual	$n-k$	$SSE$	$MSE = \frac{SSE}{n-k}$	
Total	$n-1$	$SS_{Total}$		

Source	df	SS	MS	F
Groups	$k-1$	$\sum(\bar{y}_i - \bar{y})^2$	$MS_{Groups} = \frac{SS_{Groups}}{k-1}$	$F = \frac{MS_{Groups}}{MSE}$
Residual	$n-k$	$\sum(y_i - \bar{y}_i)^2$	$MSE = \frac{SSE}{n-k}$	
Total	$n-1$	$\sum(y_i - \bar{y})^2$		