

# Lecture 19

## Introduction to Statistics

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## Terminology

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## Population and Sample

Population: All individuals (or items) of interest.

- Typically, we want to learn *something* about the population
- Usually impossible to get information from entire population

Sample: A subset of the population

- Since samples are much smaller than population, it possible to actually get information about the sample
- This information is called "*data*" or "*sample data*".

Statistics: (as a field)

- Use probability to learn about the real world (population) from data (sample).
- Assuming random mechanism generated the data allows us to use probability.

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## Random Variables and Observations

Population "measurements":

- $X_1, X_2, \dots, X_n \stackrel{iid}{\sim} f_X(x)$
- $X_i$  represents (theoretical) measurements from the population.
- $f_X(x)$  represents the population distribution.

Observations:

- $x_1, x_2, \dots, x_n$  are the observed data (or realization of the random variables).
- $x_i$  are actual measurements from the sample.

Use the observed values  $(x_1, \dots, x_n)$  to learn about the population.

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## Example

Example 1: A machine fills bottles of water. We are interested in the amount of water filled in the bottles.

- $X_i$  = amount of water filled in bottle  $i$  for  $i = 1, \dots, n$
- $X_i$  follows *some* distribution (with *some* parameters).
- But, its impossible to measure *every* bottle that the machine fills

So take a *sample* of  $n$  bottles from the machine, and measure the amount of water in them.

- Gives observed values  $x_1, \dots, x_n = (500.01, \dots, 499.80, )$
- Use this information to understand how much water the machine fills in general (population)

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## Drawing Samples

Typically, we assume a *simple random sample (SRS)* is drawn from the population to create our sample

- All subsets of same size are equally likely to be chosen
- Guarantees the sample is representative of population
  - leads to good inferences
- If not, we will introduce *bias* in our sample
  - inferences can be way off from the truth
  - leads to untrustworthy results

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# Descriptive Statistics

## Descriptive Statistics

Once we have obtained data from the sample, what comes next?

Descriptive Statistics: Describe/summarize key features of the data

- Graphics → visualize the data, describe shape, etc



- Numbers → numerical summaries of quantities of interest

"Typical value"      median       $\bar{x}$

how spread apart  
is my data?

range  
= max - min

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No conclusions are made yet. We just want an idea of what the data looks like.

# Inference

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## Estimation

Inferential Statistics: Draw conclusions about the population/distribution that generate the data.

1. Estimation: Estimate the parameters of the probability distribution that generated the data
  - In probability portion of the course, we assumed we knew the parameters of the distribution to answer questions
    - Ex: Get average of 5 hits per hour to a website.  $X = \#$  of hits in next hour.  $X \sim Pois(5)$ . What is  $P(X < 3)$ ?
  - In statistics, parameters are unknown and need to be estimated by the data
    - Confidence intervals
    - Hypothesis testing

## Prediction

2. Prediction: Estimate parameters of a data model, then use model to predict values for new observations

Example 2:  $X = \text{ACT score}$ ;  $Y = \text{Freshman GPA}$

We model relationship as between  $X$  and  $Y$  as:

$$Y = f(X) + \epsilon$$

Use data to learn about the form of  $f(X)$ , "fit" a model, and then we can predict the GPA of a new student based on their ACT score.

$$\hat{Y}_{new} = \hat{f}(x)$$