



# Lecture 18

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Decisions & Uncertainty

# Weekly Goals

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- Monday
    - Comparing distributions
    - Hypothesis tests and p-values
  - **Today**
    - Making decisions with incomplete information
    - Error probabilities
  - Friday
    - A/B testing
    - Permutation Test
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# Announcements

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- This Week
    - Ramesh's office hours cancelled today
    - Lab attendance is optional (Lab 6 now due Fri 11:59PM)
    - Homework 6 due tomorrow
  - Next Week
    - Tutoring sections via Google Hangouts
    - Midterm Review Lab attendance will be optional
    - Midterm on March 13th, 7PM
      - **Midterm Review Cancelled**
        - Video/Slides uploaded before Wednesday
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# Decisions and Uncertainty

# Incomplete Information

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- We are trying to choose between two views of the world, based on data in a sample.
  - It is not always clear whether the data are consistent with one view or the other.
  - Random samples can turn out quite extreme. It is unlikely, but possible.
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# **Review: Terminology**

# Testing Hypotheses

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- A test chooses between two views of how data were generated
  - The views are called **hypotheses**
  - The test picks the hypothesis that is better supported by the observed data
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# Null and Alternative

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The method only works if we can simulate data under one of the hypotheses.

- **Null hypothesis**
    - A well defined chance model about how the data were generated
    - We can simulate data under the assumptions of this model – “under the null hypothesis”
  - **Alternative hypothesis**
    - A different view about the origin of the data
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# Test Statistic

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- The statistic that we choose to simulate, to decide between the two hypotheses

Questions before choosing the statistic:

- What values of the statistic will make us lean towards the null hypothesis?
  - What values will make us lean towards the alternative?
    - Preferably, the answer should be just “high” or just “low”. Try to avoid “both high and low”.
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# Prediction Under the Null Hypothesis

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- Simulate the test statistic under the null hypothesis; draw the histogram of the simulated values
  - This displays the **empirical distribution of the statistic under the null hypothesis**
  - It is a prediction about the statistic, made by the null hypothesis
    - It shows all the likely values of the statistic
    - Also how likely they are (**if the null hypothesis is true**)
  - The probabilities are approximate, because we can't generate all the possible random samples
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# Conclusion of the Test

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Resolve choice between null and alternative hypotheses

- Compare the **observed test statistic** and its empirical distribution under the null hypothesis
- If the observed value is **not consistent** with the distribution, then the test favors the alternative (“data is more consistent with the alternative”)

Whether a value is consistent with a distribution:

- A visualization may be sufficient
  - If not, there are conventions about “consistency”
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# **Another Example**

# The Problem

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- Large(-ish) Statistics class divided into 12 discussion sections
  - Graduate Student Instructors (GSIs) lead the sections
  - After the midterm, students in Section 3 notice that the average score in their section is lower than in others
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# The GSI's Defense

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## **GSI's position (Null Hypothesis):**

- If we had picked my section at random from the whole class, we could have got an average like this one.

## **Alternative:**

- No, the average score is too low. Randomness is not the only reason for the low scores.

(Demo)

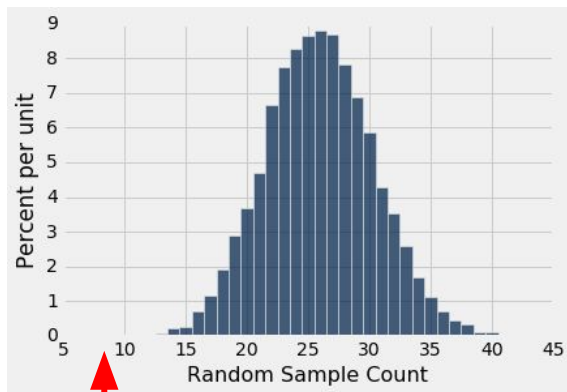
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# **Statistical Significance**

# Tail Areas

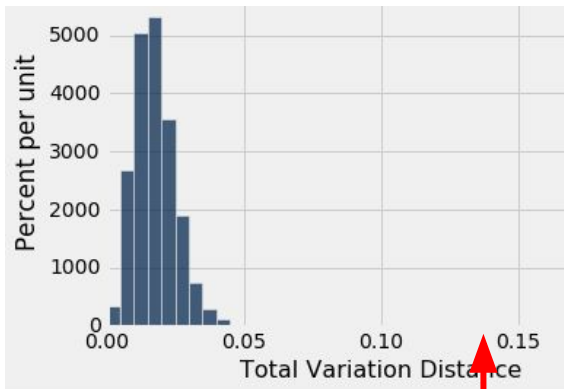
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Alabama Jury



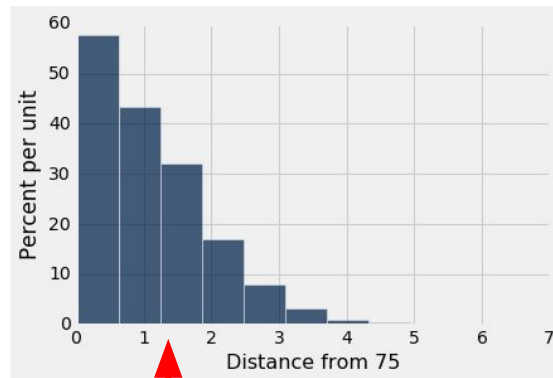
Observed Number (8)

Alameda Jury



Observed TVD (0.14)

Pea Plants



Observed Distance (1.32)



# Conventions About Inconsistency

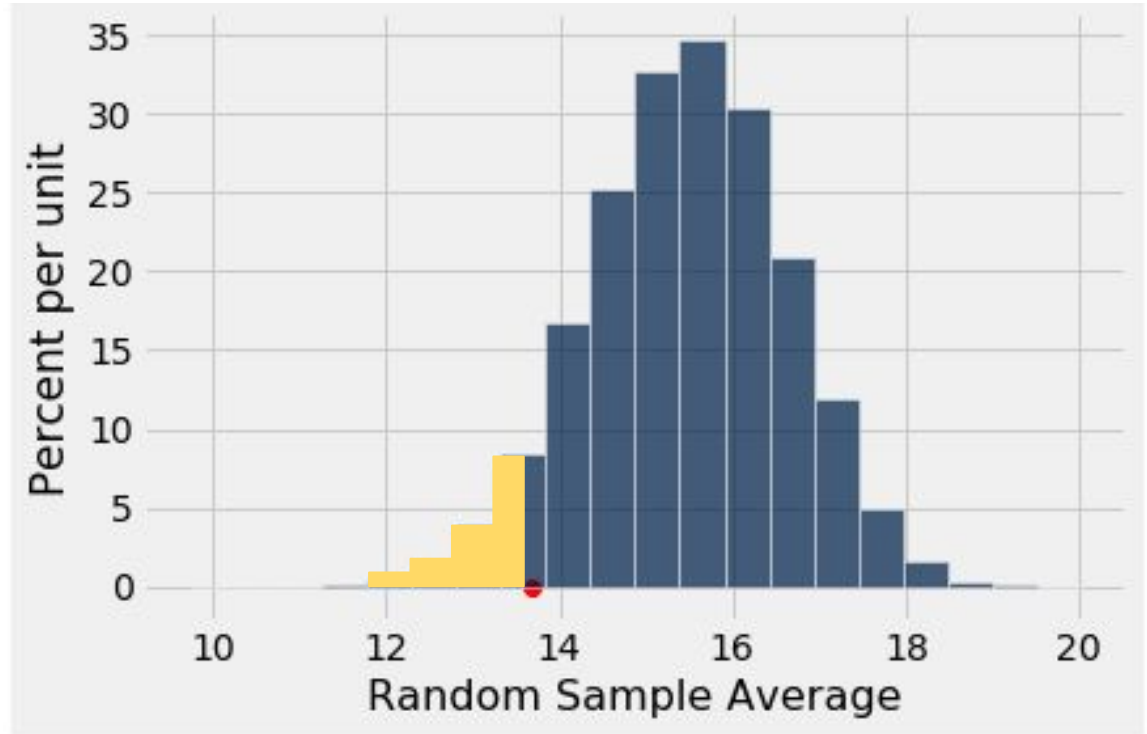
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- **“Inconsistent with the null”:** The test statistic is in the tail of the empirical distribution under the null hypothesis
  - **“In the tail,” first convention:**
    - The area in the tail is less than 5%
    - The result is “statistically significant”
  - **“In the tail,” second convention:**
    - The area in the tail is less than 1%
    - The result is “highly statistically significant” (Demo)
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# The P-Value as an Area

Empirical distribution  
of the test statistic  
under the null  
hypothesis

The red dot is the  
observed statistic.



# Definition of the $P$ -value

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Formal name: **observed significance level**

The  $P$ -value is the chance,

- under the null hypothesis,
  - that the test statistic
  - is equal to the value that was observed in the data
  - or is even further in the direction of the alternative.
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# How We've Tested Thus Far

# Hypothesis Testing Review

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- **One Category** (*ex: percent of flowers that are purple*)
  - Test Statistic (1): `empirical_percentage`
  - Test Statistic (1): `abs(empirical_percentage - null_percentage)`
  - How to Simulate: `sample_proportions(n, null_dist)`
- **Multiple Categories** (*ex: ethnicity distribution of jury panel*)
  - Test Statistic: `tvd(empirical_dist, null_dist)`
  - How to Simulate: `sample_proportions(n, null_dist)`
- **Numerical Data** (*ex: scores in a lab section*)
  - Test Statistic: `empirical_mean`
  - How to Simulate: `population_data.sample(n, with_replacement=False)`