# BC COMS 1016: Intro to Comp Thinking \& Data Science 

 Significant Testing (P-values) $\&$A/B Testing $\frac{\square+1}{\square}$

## Announcements

- Lab 06 - Inference and the Death Penalty
- Due Monday 03/28
- HW06 - Testing Hypotheses
- Due Thursday 03/31



## Tail Areas

Alabama Jury


## Alameda Jury



## Tail Areas

Alabama Jury


Observed Number (8)

## Alameda Jury



## Not so clear example



Conventions About Inconsistency

- "Inconsistent with the null": The test statistic is in the tail of the empirical distribution under the null hypothesis


## Not so clear example



- "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"


## Observed significance level (aka P-value)

## 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 tail


## Example

Scenario: After the midterm, students in a MW lab (of 27 students) noticed that their scores were on average lower than the rest of the class.

Question:

Potential Answers:
Null Hypothesis: The average score of the students in the lab is like the average score of the same number of students picked at random from the class

Alternative Hypothesis: No, the average is too low

## Example

Scenario: After the midterm, students in a MW lab noticed that their scores were on average lower than the rest of the class.

## Question:

Did the 27 students do lower by chance?

## Potential Answers:

Null Hypothesis: The average score of the students in the lab is like the average score of the same number of students picked at random from the class

Alternative Hypothesis: No, the average is too low
Statistic to measure:
The average score per section (27 students)

Assessing a Model

- Choose a statistic to measure the "discrepancy" between model and data
- Average score per 27 students
- Simulate the statistic under the model's assumptions
- np.average(scores_only.sample(27, with_replacement=False))
- Compare the data to the model's predictions:
- Draw a histogram of simulated values of the statistic
- Compute the observed statistic from the real sample


## Histogram of simulated values \& observed statistic

## Is the observed statistic consistent with the histogram?



## Compute the p-value

## 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 tail



## Compute the p-value

## Probability $(A)=\frac{\text { number of outcomes that make A happen }}{\text { total number of outcomes }}$



## Compute the p-value

## A = the sampled statistic was less than or equal to the observed statistic



## Compute the p-value

$P(A)=$ (the number of times the sampled statistic was less than the observed statistic) divided by the number of samples


## Compute the p-value

$P(A)=$
sum(sample averages $\leq$ observed averages)
50K


## Compute the p-value

$$
\begin{aligned}
& P(A)=0.05682 \approx 5 \% \\
& P(A)=0.05682 \approx 5 \%
\end{aligned}
$$

## Compute the p-value

Area to the left of the gold line: 5\%



## Terminology

- Compare values of sampled individuals in Group A with values of sampled individuals in Group B.
- Question: Do the two sets of values come from the same underlying distribution?
- Answering this question by performing a statistical test is called A/B testing.


## The Groups and the Questions

- Random sample of mothers of newborns. Compare:
A. Birth weights of babies of mothers who smoked during pregnancy
B. Birth weights of babies of mothers who didn't smoke
- Question: Could the difference be due to chance alone?


## Null Hypothesis:

- In the population, the distributions of the birth weights of the babies in the two groups are the same. (They are different in the sample just due to chance.)
Alternative Hypothesis:
- In the population, the babies of the mothers who smoked weigh less, on average, than the babies of the non-smokers


## Test Statistic

## Group A: non-smokers <br> Group B: smokers

## Statistic:

- Difference between average weights:
- Group B average - Group A average

Negative values of this statistic favor the alternative

If the null is true, all rearrangements of labels are equally likely

Permutation Test:

- Shuffle all birth weights
- Assign some to Group A and the rest to Group B
- Key: keep the sizes of Group A and Group B that same from before
- Find the difference between the two shuffled groups
- Repeat


## Random Permutations

- tbl.sample(n)

Table of $n$ rows picked randomly with replacement

- tbl.sample()
- Table with same number of rows as original tbl,
- picked randomly with replacement
- tbl.sample(n, with_replacement = False)
- Table of $n$ rows picked randomly without replacement
- tbl.sample(with_replacement = False)
- All rows of tbl, in random order



## Hypothesis Testing Review

1 Sample: One Category (e.g. percent of black male jurors)

- Test Statistic: empirical_percent, abs(empirical_percent - null_percent)
- How to Simulate: sample_proportions(n, null_dist)

1 Sample: Multiple Categories (e.g. ethnicity distribution of jury panel)

- Test Statistic: tvd(empirical_dist, null_dist)
- How to Simulate: sample_proportions(n, null_dist)

1 Sample: Numerical Data (e.g. scores in a lab section)

- Test Statistic: empirical_mean, abs(empirical_mean - null_mean)
- How to Simulate: population_data.sample(n, with_replacement=False)

2 Samples: Numerical Data (e.g. birth weights of smokers vs. non-smokers)

- Test Statistic: group_a_mean - group_b_mean,

```
• group___mean - group_a_mean, abs(group_a_mean - group__b_mean)
```

- How to Simulate: empirical_data.sample(with_replacement=False)

