

# STAT 311: Hypothesis Testing

Y. Samuel Wang

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# Logistics

- Homework 6 posted
- Lab 6 posted as well

# Hypothesis Testing Procedure

A hypothesis test consists of the following steps

- 1 Determine the Null and Alternative hypotheses
- 2 Determine the Null distribution
- 3 Gather data / calculate a test statistic
- 4 Calculate a p-value
- 5 Draw conclusions

# P-values

P-values (probability values) explicitly quantify **the probability of a test statistic as (or more) extreme as the one I actually did observe if the null hypothesis is true.**

- A smaller p-value denotes stronger evidence against the null hypothesis
- Typically a cut-off of .05 is used for statistical significance
- The cut off is called the **level** or **size** of the hypothesis test

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P-values do **not** mean- “How likely is the null hypothesis?”

- Correct:  $P(\text{Data}|\text{Null Hypothesis})$
- Incorrect:  $P(\text{Null Hypothesis}|\text{Data})$

If the p-value is large we do not “accept” the null hypothesis, we simply **“fail to reject”**

# Details about Hypothesis Testing

The specifics of a hypothesis test depend on the details

Parameter	$H_0$	$H_A$	Test Statistic	Null Distribution
$p$	$p = p_0$	$p \neq p_0$	$\hat{p}$	$\mathcal{N}\left(p_0, \frac{p_0(1-p_0)}{n}\right)$
$p_1 - p_2$	$p_1 - p_2 = 0$	$p_1 - p_2 \neq 0$	$\hat{p}_1 - \hat{p}_2$	$\mathcal{N}\left(0, \hat{p}_0(1 - \hat{p}_0)\left(\frac{1}{n_1} + \frac{1}{n_2}\right)\right)$ $\hat{p}_0 = \frac{\hat{p}_1 n_1 + \hat{p}_2 n_2}{n_1 + n_2}$
$\mu$	$\mu = \mu_0$	$\mu \neq \mu_0$	$\sqrt{n} \frac{\bar{x} - \mu_0}{s_x}$	$T_{n-1}$
$\mu_1 - \mu_2$	$\mu_1 - \mu_2 = 0$	$\mu_1 - \mu_2 \neq 0$	$\frac{(\bar{x}_1 - \bar{x}_2) - 0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$	$T_{\min(n_1-1, n_2-1)}$

For the  $T$  distributions, if the degrees of freedom are greater than 30, we can use a normal distribution, since the difference is negligible

## Example: Single proportion

Each month, the employees of Alaska Airlines receive a \$50 bonus, if at least 78% of all flights are on-time. On Feb 2, 2016, Alaska Airlines had  $X$  arrivals at Sea-Tac Airport. Of the  $X$  arrivals,  $Y$  arrived on time. Based on our sample, do we have reason to believe that the true proportion of all on-time arrivals at Sea-Tac is less than 78%?

$$H_0 : p = .78$$

$$H_A : p < .78$$

## Example: Single proportion

Each month, the employees of Alaska Airlines receive a \$50 bonus, if at least 78% of all flights are on-time. On Feb 2, 2016, Alaska Airlines had 146 arrivals at Sea-Tac Airport. Of the 146 arrivals, 99 arrived on time<sup>1</sup>. Based on our sample, do we have reason to believe that the true proportion of all Alaska arrivals at Sea-Tac which are on-time is less than 78%?

$$H_0 : p = .78$$

$$H_A : p < .78$$

Based on our sample, we have  $\hat{p} = .68$

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<sup>1</sup>http:

[//transborder.bts.gov/xml/ontimesummarystatistics/src/index.xml](http://transborder.bts.gov/xml/ontimesummarystatistics/src/index.xml)



## Example: Single proportion

Under the null hypothesis,  $p = .78$ ,

$$\hat{p} \sim \mathcal{N}(p_0, p_0(1 - p_0)/n) = \mathcal{N}(.78, 0.0011) =$$

Notice that we use  $p_0$  for the null distribution and not  $\hat{p}$



## Example: Single proportion

Since the p-value is less than .05, we would reject the null hypothesis that the true proportion of on-time arrivals is .78.

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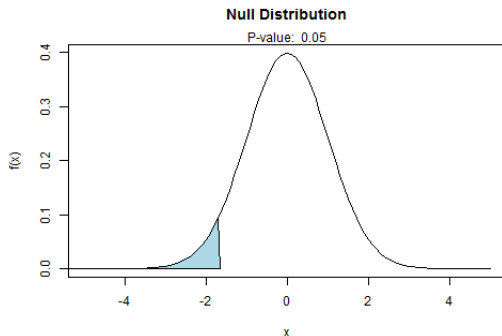
Since the p-value is less than .05, we would reject the null hypothesis that the true proportion of on-time arrivals is .78.

Can we generalize this conclusion?

- Is this date representative of other days in February?
- Is Sea-Tac representative of all Alaska arrivals?

# Rejection Region

Instead of looking at a p-value of an observed value, we could've just first decided a cut-off, then translate that back into a length. For instance, if we decide on a cut-off of .05, that means we would reject the null if the test statistic is less than  $-1.645$ .



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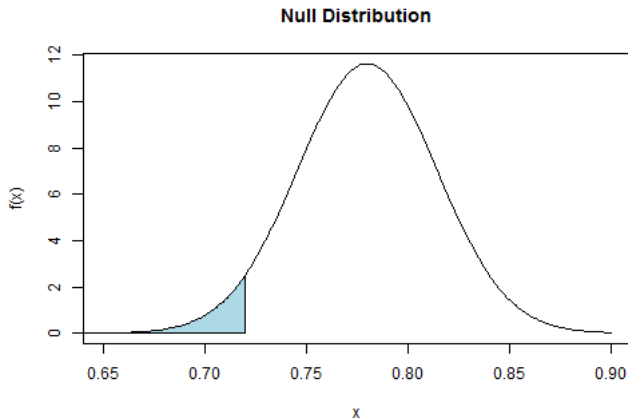
$$\frac{\hat{p} - p_0}{\sqrt{p_0(1 - p_0)/n}} = -1.645$$

$$\Rightarrow \hat{p} = p_0 - 1.645\sqrt{p_0(1 - p_0)/n} = .72$$

Thus, an equivalent procedure to rejecting if the p-value is less than .05, is rejecting if  $\hat{p}$  is less than .72. The region of the null distribution less than .72 is called the **Rejection Region**.

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## Example: Difference in proportions

A 2015 Pew Research Poll investigated the internet and social media habits of Americans<sup>2</sup>. One particular question asked respondents whether or not they used auto-deleting messaging apps (ie Snapchat). One might be interested in whether the proportion of men and women which use the these apps differs.

$$H_0 : p_{men} - p_{women} = 0$$

$$H_A : p_{men} - p_{women} \neq 0$$

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<sup>2</sup><http://www.pewinternet.org/2015/08/19/mobile-messaging-and-social-media-2015/>



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Note that this is a two sided test, so the p-value will be  $P(|\hat{p}_{men} - \hat{p}_{women}| > \text{Observed})$  assuming the Null Hypothesis is true

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## Example: Difference in proportions

A 2015 Pew Research Poll investigated the internet and social media habits of Americans. One particular question asked respondents whether or not they used auto-deleting messaging apps (ie Snapchat). One might be interested in whether the proportion of men and women which use the these apps differs. The released data gives  $\hat{p}_{men} = .17$  and  $\hat{p}_{women} = .18$ . It does not give the sample size for each sub-population, but the total sample size is 1907, so let's assume that's evenly divided with  $n_{men} = 953$  and  $n_{women} = 954$ .

## Example: Difference in proportions

Under the null hypothesis, where  $p_{men} = p_{women}$ ,

$$\hat{p}_{men} - \hat{p}_{women} \sim \mathcal{N}\left(0, \hat{p}_0(1 - \hat{p}_0) \left(\frac{1}{n_1} + \frac{1}{n_2}\right)\right)$$

where  $\hat{p}_0 = \frac{\hat{p}_1 n_1 + \hat{p}_2 n_2}{n_1 + n_2}$ . Note that this null distribution actually doesn't specify what the common value is, just that they are equal. In this case,  $\hat{p}_0 = .175$  so the null distribution is

$$\mathcal{N}(0, 0.0174^2)$$

which gives a z-score of

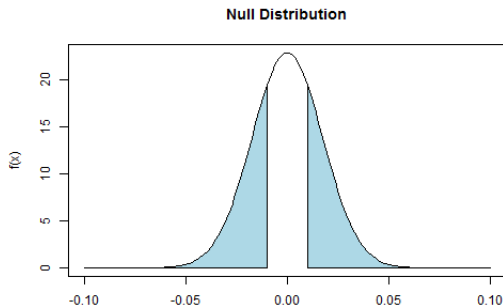
$$\frac{.01 - 0}{.0174} = .575$$

so the p-value is  $P(|Z| > .575) = 0.282 \times 2$

## Example: Difference in proportions

Because it is a two sided test, we actually “double” the p-value we see in the table. This is because we are only interested in the absolute value  $|\hat{p}_{men} - \hat{p}_{women}|$ .

$$\begin{aligned} &P(|\hat{p}_{men} - \hat{p}_{women}| > .01 | \text{Null Hypothesis}) \\ &= P(|Z| > .575) = P(Z > .575) \times 2 = 0.282 \times 2 \end{aligned}$$



## Example: Difference in proportions

Because the p-value is greater than .05, we fail to reject the Null hypothesis that there is no difference between the proportion of men and women who use auto-delete messaging services

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Note that we **do not** accept the Null hypothesis! We simply do not have evidence to reject it.

## Example: Single Mean

In 2013, a class action suit was brought against Subway, claiming that the footlong subs were not actually 12 inches long. Subway eventually agreed to settle the lawsuit for \$525,000. Let  $\mu$  denote the average sub length.

$$H_0 : \mu = 12$$

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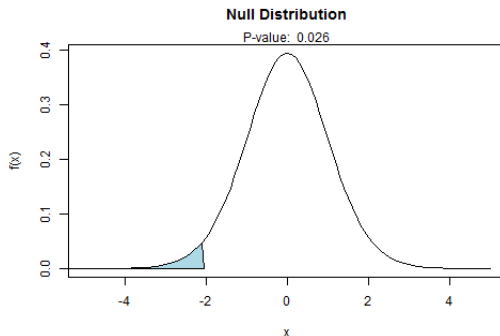
$$H_A : \mu < 12$$

Suppose we measure 25 sandwiches and get an average length of 11.91 inches and  $s_x = .22$  which yields  $\sqrt{n} \frac{\bar{x} - \mu_0}{s_x} = -2.05$

We know that  $\sqrt{n} \frac{\bar{x} - \mu_0}{s_x} \sim T_{24}$ , so we can find the probability of a value as or more extreme as -2.05 from a  $T_{24}$ .



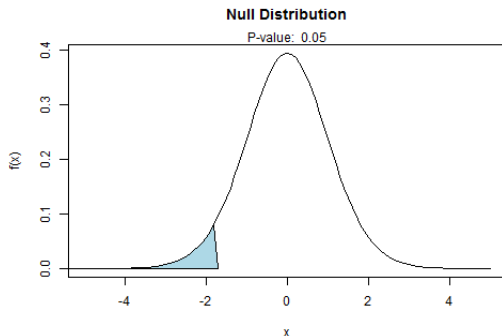
# Example: Single Mean



Under the null hypothesis, we would only expect to see a test statistic as or more extreme than -2.05 about .026 of the time. Thus, in a hypothesis test with level .05, we would reject the null hypothesis that the average length of the footlong subs is actually 12 inches long.

# Rejection Region

Instead of looking at a p-value of an observed value, we could've just first decided a cut-off, then translate that back into a length. For instance, if we decide on a cut-off of .05, that means we would reject the null if the test statistic is less than  $-1.71$ .



The area to the left of  $-1.71$  is called the **Rejection Region**.