

# Stat 201: Introduction to Statistics

Significance Tests – for Mean  
Differences

# Telling Which Parameter We're After

- As statisticians, or data scientists, it's our job to hear a problem and decide what we're after
  - We call the parameter of interest the **target parameter**

Parameter	Point Estimate	Key Phrase	Type of Data
$\mu_1 - \mu_2$	$\bar{x}_1 - \bar{x}_2$	Mean Difference	Quantitative
$\rho_1 - \rho_2$	$\hat{p}_1 - \hat{p}_2$	Difference of Proportion, percentage, fraction, rate	Qualitative (Categorical)

# Note!

- We use the usual approach of confidence intervals and hypothesis testing on the mean difference  $\mu_d = \mu_1 - \mu_2$  as we did in chapters 7-9
  - Our data becomes  $x_{d_i} = x_{1_i} - x_{2_i}$  and we are interested in making inference on  $\mu_d$ .

# Hypothesis Test for Mean Difference: Step 1

- State Hypotheses:
  - **Null hypothesis:** that the population mean equals some  $\mu_o$ 
    - $H_o: \mu_d = \mu_1 - \mu_2 \leq \mu_o$  (one sided test)
    - $H_o: \mu_d = \mu_1 - \mu_2 \geq \mu_o$  (one sided test)
    - $H_o: \mu_d = \mu_1 - \mu_2 = \mu_o$  (two sided test)
  - **Alternative hypothesis:** What we're interested in
    - $H_a: \mu_1 - \mu_2 > \mu_o$  (one sided test)
    - $H_a: \mu_1 - \mu_2 < \mu_o$  (one sided test)
    - $H_a: \mu_1 - \mu_2 \neq \mu_o$  (two sided test)

# Hypothesis Test for Mean Difference: Step 2

- Check the assumptions
  - The difference variable must be quantitative
  - The data are obtained using randomization
  - Samples are **independent**
  - The differences are from the normal distribution
    - If  $n_1 > 30$  and  $n_2 > 30$
    - If either  $n < 30$  we need to check a histogram to ensure the differences are normally distributed

# Hypothesis Test for Mean Difference:

## Step 3: Case One With known $\sigma_1$ & $\sigma_2$

- Calculate Test Statistic
  - The test statistic measures how different the sample proportion we have is from the null hypothesis
  - We calculate the z-statistic by assuming that  $\mu_{d_0}$  is the population mean difference

$$z^* = \frac{((\bar{x}_1 - \bar{x}_2) - \mu_0)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

# Hypothesis Test for Mean Differences: Step 4: Case One With known $\sigma_1$ & $\sigma_2$

- Determine the P-value
  - The P-value describes how unusual the sample data would be if  $H_0$  were true.

Alternative Hypothesis	Probability	Formula for the P-value
$H_a: \mu_1 - \mu_2 > \mu_o$	Right tail	$P(Z > z^*)$
$H_a: \mu_1 - \mu_2 < \mu_o$	Left tail	$P(Z < z^*)$
$H_a: \mu_1 - \mu_2 \neq \mu_o$	Two-tail	$2 * P(Z < - z^* )$

# Hypothesis Test for Mean Difference:

## Step 3: Case Two With unknown $\sigma_1 = \sigma_2$

- Calculate Test Statistic
  - The test statistic measures how different the sample proportion we have is from the null hypothesis
  - We calculate the t-statistic by assuming that  $\mu_{d_0}$  is the population mean difference

$$t^* = \frac{((\bar{x}_1 - \bar{x}_2) - \mu_0)}{\sqrt{s_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}} \text{ where } s_p^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}$$



# Hypothesis Test for Mean Difference:

## Step 4: Case Two With unknown $\sigma_1 = \sigma_2$

- Determine the P-value
  - The P-value describes how unusual the sample data would be if  $H_0$  were true.

Alternative Hypothesis	Probability	Formula for the P-value
$H_a: \mu_1 - \mu_2 > \mu_o$	Right tail	$P(T > t^*)$
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# Hypothesis Test for Mean Difference:

## Step 3: Case Three With unknown $\sigma_1 \neq \sigma_2$

- Calculate Test Statistic
  - The test statistic measures how different the sample proportion we have is from the null hypothesis
  - We calculate the t-statistic by assuming that  $\mu_{d_0}$  is the population mean difference

$$t^* = \frac{((\bar{x}_1 - \bar{x}_2) - \mu_0)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

# Hypothesis Test for Mean Difference:

## Step 4: Case Three With unknown $\sigma_1 \neq \sigma_2$

- Determine the P-value
  - The P-value describes how unusual the sample data would be if  $H_0$  were true.

Alternative Hypothesis	Probability	Formula for the P-value
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# Hypothesis Test for Mean Difference:

## Step 5

- Summarize the test by reporting and interpreting the P-value
  - Smaller p-values give stronger evidence against  $H_0$
- If  $p\text{-value} \leq (1 - \textit{confidence}) = \alpha$ 
  - Reject  $H_0$ , with a p-value = \_\_\_\_\_, we have sufficient evidence that the alternative hypothesis might be true
- If  $p\text{-value} > (1 - \textit{confidence}) = \alpha$ 
  - Fail to reject  $H_0$ , with a p-value = \_\_\_\_\_, we do not have sufficient evidence that the alternative hypothesis might be true

# Example

- According to a NY Times article a survey conducted showed that 22 men averaged 3 hours of housework per day with a standard deviation of .85 and 49 women averaged 6 hours of housework per day with a standard deviation of 1.3
- Test with 90% confidence interval that the true population difference of means is more males than females

# Example

- State Hypotheses:

$$- H_o: \mu_d = \mu_1 - \mu_2 \leq 0$$

$$- H_a: \mu_d = \mu_1 - \mu_2 > 0$$

# Example

- Check the assumptions
    1. Each sample must be obtained through randomization
    2. Samples are **independent**
    3. The differences are from the normal distribution
      - If  $n_1 < 30$  &  $n_2 > 30$   
**AND**
        - We don't know the populations follow the normal distribution
- \*Proceed with caution\***

# Example

- Calculate Test Statistic

$$\begin{aligned} t^* &= \frac{((\bar{x}_1 - \bar{x}_2) - \mu_0)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}} \\ &= \frac{((3 - 6) - 0)}{\sqrt{\frac{.85^2}{22} + \frac{1.3^2}{49}}} \\ &= -11.56151 \end{aligned}$$



# Example

First we solve for  $v$ :

$$v = \frac{\left(\frac{.85^2}{22} + \frac{1.3^2}{49}\right)^2}{\frac{\left(\frac{.85^2}{22}\right)^2}{22-1} + \frac{\left(\frac{1.3^2}{49}\right)^2}{49-1}} = 59.5402 \approx 59$$

# Example

- Determine the P-value
  - The P-value describes how unusual the sample data would be if  $H_0$  were true.

$$\begin{aligned} & 1 - P(T < t^*) \\ &= 1 - P(T < -11.56151) \\ &\approx 1 - 0 = 1 \end{aligned}$$

# Example

- Summarize the test by reporting and interpreting the P-value

$$1 > (1 - .9) = .1$$

We fail to reject  $H_0$ , we do not have sufficient evidence to suggest that males do more housework than females

Summary!

# Sampling Distribution for the Sample Mean Summary

Shape, Center and Spread of Population	Shape of sample	Center of sample	Spread of sample
Populations are normal with means $\mu$ and standard deviations $\sigma$ .	Regardless of the sample size $n$ , the shape of the distribution of the sample mean is normal	$\mu_d = \mu_1 - \mu_2$	$\sigma_d = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$
Population are not normal with means $\mu$ and standard deviations $\sigma$ .	As the sample size $n$ increases, the distribution of the sample mean becomes approximately normal	$\mu_d = \mu_1 - \mu_2$	$s_d = \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$

# Hypothesis Testing for $\mu$ known $\sigma_1$ and $\sigma_2$

Step One:	<p>(i) <math>H_0: \mu_1 - \mu_2 = \mu_0</math> &amp; <math>H_a: \mu_1 - \mu_2 \neq \mu_0</math></p> <p>(ii) <math>H_0: \mu_1 - \mu_2 \geq \mu_0</math> &amp; <math>H_a: \mu_1 - \mu_2 &lt; \mu_0</math></p> <p>(iii) <math>H_0: \mu_1 - \mu_2 \leq \mu_0</math> &amp; <math>H_a: \mu_1 - \mu_2 &gt; \mu_0</math></p>
Step Two:	<ol style="list-style-type: none"> <li>1. Quantitative</li> <li>2. <i>Random Sample</i></li> <li>3. <math>n &gt; 30</math> OR the population is bell shaped</li> </ol>
Step Three:	$z^* = \frac{((\bar{x}_1 - \bar{x}_2) - \mu_0)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$
Step Four:	<p>(i) <math>H_a: \mu_1 - \mu_2 \neq \mu_0 \rightarrow</math> p-value = <math>2 * P(Z &lt; - z^* )</math></p> <p>(ii) <math>H_a: \mu_1 - \mu_2 &lt; \mu_0 \rightarrow</math> p-value = <math>P(Z &lt; z^*)</math></p> <p>(iii) <math>H_a: \mu_1 - \mu_2 &gt; \mu_0 \rightarrow</math> p-value = <math>P(Z &gt; z^*) = 1 - P(Z &lt; z^*)</math></p>
Step Five:	<p>If p-value <math>\leq (1 - \text{confidence}) = \alpha</math>  <math>\rightarrow</math> Reject <math>H_0</math></p> <p>If p-value <math>&gt; (1 - \text{confidence}) = \alpha</math>  <math>\rightarrow</math> Fail to Reject <math>H_0</math></p>

# Hypothesis Testing for $\mu$ unknown $\sigma_1 = \sigma_2$

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