## Elementary Statistics – Inference for Numerical Data Pt. 2

Dr. Ab Mosca (they/them)

#### Plan for Today

- Inference for numerical variables
  - Paired means

# Inference for Two Independent Means

### Confidence Interval for Difference Between Two Independent Means

For confidence intervals, we use  $s_1$  and  $s_2$  as the best guess of  $\sigma_1$  and  $\sigma_2$ , so

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

For degrees of freedom, use the smaller of  $n_1-1$ ,  $n_2-1$ 

We use SE to compute margin of error for our confidence interval:  $\left((\bar{x}_1 - \bar{x}_2) - t_{df}^* SE, (\bar{x}_1 - \bar{x}_2) + t_{df}^* SE\right)$ 

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You suspect WSU and SC students have different numbers of siblings. You perform an experiment to statistically test this suspicion. You sample 100 WSU students and find on average they have 3 siblings with s=0.5. You sample 110 SC students and find on average they have 1 sibling with s=0.75. Calculate a 95% CI for  $\mu_{SC}-\mu_{WSU}$  from your  $\bar{x}_{SC}$  and  $\bar{x}_{WSU}$ .

# Inference for Two Independent Means

### Hypothesis Test for Difference Between Two Independent Means

For a hypothesis test, our null hypothesis will be that there is no difference between means.

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Then,  $T=\frac{(\bar{x_1}-\bar{x_2})-0}{SE}$  , and df is the smaller of  $n_1-1$  ,  $n_2-1$ 

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You perform a hypothesis test with  $H_0$ :  $\mu_{SC} - \mu_{WSU} = 0$ ,  $H_A$ :  $\mu_{SC} - \mu_{WSU} \neq 0$ . Finish the hypothesis test. Calculate T, find the p-value, and compare to an  $\alpha$  of 0.05.

## Inference for Dependent Means

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But what if observations are not independent?

Sometimes, dependency cannot be addresses through a statistical method. However, one specific type of dependency, *pairing*, can be with tools we already know.

#### Paired Means

Paired data result from a specific type of experimental structure.

In this type of experiment:

- The observational unit is paired across two levels of the explanatory variable
- For each observational unit, quantitative measures are made on each of the two levels of the explanatory variable. The two measurements are subtracted and only the difference is retained

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Are the following data paired? If yes, identify the observational unit, explanatory variable, measurement, and response variable.

- Compare pre- (beginning of semester) and post-test (end of semester) scores of students.
- Assess gender-related salary gap by comparing salaries of randomly sampled men and women.
- Compare artery thicknesses at the beginning of a study and after 2 years of taking Vitamin E for the same group of patients.
- Assess effectiveness of a diet regimen by comparing the before and after weights of subjects.

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Our sample statistic ( $\bar{x}_{\text{diff}}$ ) represents our best guess for the true population parameter, ( $\mu_{diff}$ ). We know this best guess is not perfect; we expect error (variability) due to the sampling process.

Because we can't know the truth directly, we infer the truth via:

- 1. A confidence interval
- 2. A hypothesis test

In either case, we need the sampling distribution for  $\bar{x}_{diff}$ .

We can approximate it via the central limit theorem as long as:

- 1. The sample's observations are independent
- 2. The sample size is **large enough**,  $n \ge 30$ , or clearly normally distributed with no outliers

When these conditions are met, variability of  $\bar{x}_{diff}$  is well described by:

$$SE(\bar{x}_{diff}) = \frac{best\ guess\ of\ \sigma_{diff}}{\sqrt{n_{diff}}}$$

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We typically use  $s_{diff}$  (sample variance), as the best guess for  $\sigma_{diff}$  (population variance). However, this is less precise with small samples.

As a solution, we use the t-distribution to model the sampling distribution of  $\bar{x}_{diff}$ 

For confidence intervals, we use  $\bar{x}_{diff}$  as the best guess of  $\mu_{diff}$ , and  $s_{diff}$  as the best guess of  $\sigma_{diff}$ , so

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 $t_{df}^{st}$  is calculated from a specified percentile on the t-distribution with df.

Ex. 5<sup>th</sup> percentile of a for a 95% confidence

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## Inference for a Single Mean

We use SE to compute margin of error for our confidence interval:  $(\bar{x}_{diff} - t^*SE, \bar{x}_{diff} + t^*SE)$ 

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You suspect that on average WSU students work more during the summer than during the semester. You perform an experiment to statistically test this suspicion. You sample 100 students and calculate the difference in average number of hours worked per week between the summer and semester to be 30, with s=1.5. Calculate a 95% CI for  $\mu_{diff}$  from your  $\bar{x}_{diff}$ .

#### **Hypothesis Test for Paired Means**

When the conditions are met so that the distribution of  $\bar{x}_{\text{diff}}$  can be modeled with a t-distribution, variability of  $\bar{x}_{\text{diff}}$  is well described by:

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Because we are using the t-distribution, we will need a T-score to find our p-value:

$$T = \frac{\bar{x}_{diff} - \mu_{0diff}}{SE}$$

Degrees of freedom,  $df = n_{diff} - 1$ 

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Perform a hypothesis test for  $H_0$ :  $\mu_{diff} = 0$ ,  $H_A$ :  $\mu_{diff} \neq 0$ . Use  $\alpha = 0.05$ .

#### In-Class Activity 3

Start on the next in class activity (ico3)

Instructions are on the course website. You will have the remainder of class today and all of next class to complete the activity.

Ask questions as they come up!