

Elementary Statistics – Inference for Numerical Data Pt. 2

Dr. Ab Mosca (they/them)

Plan for Today

- Final project
- Inference for numerical variables
 - Paired means
- Note: We will work on Quiz 5 in class on Thursday

Final Project

- Any questions?

Warm Up: 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 σ_1 and σ_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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Warm Up: Inference for Two Independent Means

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} .

Warm Up: 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 α 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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Inference for Paired Means

Our sample statistic (\bar{x}_{diff}) represents our best guess for the true population parameter, (μ_{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

Inference for Paired Means

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**, $n \geq 30$, or clearly normally distributed with no outliers

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

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

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We typically use s_{diff} (sample variance), as the best guess for σ_{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}

Confidence Interval for Paired Means

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

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

Ex. 5th percentile of a for a 95% confidence

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

Inference for a Single Mean

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 μ_{diff} from your \bar{x}_{diff} .

Hypothesis Test for Paired Means

When the conditions are met so that the distribution of \bar{x}_{diff} can be modeled with a t-distribution, variability of \bar{x}_{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 $\alpha = 0.05$.

Quiz 5

Open up Quiz 5 on PLATO

Find a group to work with. Everyone in the group must participate (I will be checking).

Go through the quiz; for each problem write down the problem and your work to solve it on paper (write your name on the paper, too!). **At the end of class on Thursday, you will turn in the pages showing your work.**

Your grade on Quiz 5 will be updated based on what you turn in from class. In addition to answering the questions correctly, you will get credit for working with your group to solve the problems and asking questions.

I will come around to check-in, but also please get my attention if you have a question!