

04/18/2025

Decision Error

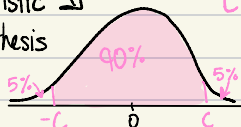
Hypothesis Testing

- 1) Form a hypothesis: H_0
- 2) Collect data
- 3) Create test statistic: $t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$
- 4) Make a decision: reject or fail to reject the null hypothesis

• We assume that the null is true

$$t = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$$

Distribution of t-statistic assuming the null hypothesis is true



Ex: CI for 90%

• The C values bound 90% of the area under the curve, the middle 90% of t-statistics if the null hypothesis is true.

- If your t-statistic is outside the CI you reject the null hypothesis. However there is a chance that your null hypothesis is true and you just have an extreme t-statistic: this is called Type I Error.
- If you increase your confidence, you decrease your type I error rate

	H_0 True	H_0 False
Reject	Type I Error α	Correct 1- β Power
Fail to reject	Correct 1- α Confidence	Type II Error β

Type I Error

- Reject H_0 when H_0 was true
- "False positive"
- You cannot minimize type I and type II error at the same time

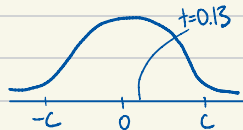
Type II Error

- Fail to reject when H_0 is false
- "Missed opportunity"

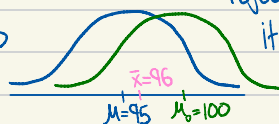
Example

- Assume the true value of the mean is $\mu = 95$
- Sample: $\bar{x} = 96$, $s = 20$, $n = 15$

t-distribution for $\mu = 95$ $\frac{\bar{x} - 95}{s/\sqrt{n}}$



what if we were testing $\mu_0 = 100$



We would still fail to reject $\mu_0 = 100$ even though it is untrue. This is a type 2 error

What causes type II error?

- distance between μ and μ_0 - "Effect Size"

- Large σ

- Smaller sample size (n)

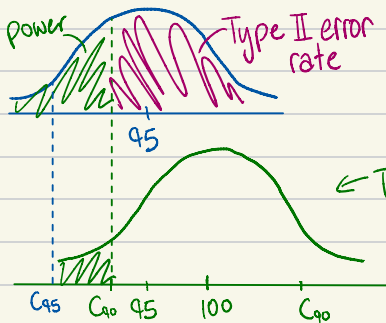
↳ With enough evidence you can always reject μ_0 because it will never be exactly identical to μ , but you have to ask yourself, is this difference meaningful?

(Ex: a weightloss drug helps people lose $\frac{1}{4}$ pound)

Example:

Assume the true value of the mean is $\mu = 95$

Here is the distribution of \bar{x} :



Note:

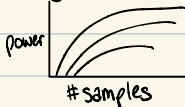
$H_0: \mu = 100$ and H_0 is false

← T-distribution for $\mu = 100$

- Increasing your confidence reduces your Type I Error rate but increases your Type II Error rate.

Power Analysis

- Try 10 different effect sizes and it creates a graph of your power distribution



Read through last 6 slides