

# **STAT 3011 Discussion 015**

## **Week 10**

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# Two Types of Confidence Intervals

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We will learn two main types of confidence intervals:

## 1. Population Proportion ( $p$ )

Used when dealing with categorical data (yes/no, success/failure)

## 2. Population Mean ( $\mu$ )

Used when dealing with quantitative data (heights, weights, scores)

Both follow the general form:

$$\text{Point Estimate} \pm \text{Margin of Error}$$

# Confidence Interval for Population Proportion ( $p$ )

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**When to use:** Estimating proportion of successes in population

**Formula:**

$$\hat{p} \pm z_{\alpha/2} \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$$

**Steps:**

1. Find  $\alpha = 1 -$  confidence level (e.g., for 95% CI:  $\alpha = 0.05$ )
2. Calculate sample proportion:  $\hat{p} = \frac{\text{number of successes}}{n}$
3. Check conditions:  $n\hat{p} \geq 15$  and  $n(1 - \hat{p}) \geq 15$
4. Find  $z_{\alpha/2}$  using `qnorm(1-\alpha/2)`
5. Calculate standard error:  $SE = \sqrt{\frac{\hat{p}(1 - \hat{p})}{n}}$
6. Calculate margin of error:  $moe = z_{\alpha/2} \times SE$
7. Construct CI:  $\hat{p} \pm moe$

# Confidence Interval for Population Mean ( $\mu$ )

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**When to use:** Estimating average value in population

**Formula:**

$$\bar{x} \pm t_{\alpha/2, df} \frac{s}{\sqrt{n}}$$

**Steps:**

1. Find  $\alpha = 1 -$  confidence level (e.g., for 95% CI:  $\alpha = 0.05$ )
2. Calculate sample mean:  $\bar{x} = \frac{\sum x_i}{n}$
3. Calculate sample standard deviation:  $s = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n-1}}$
4. Find degrees of freedom:  $df = n - 1$
5. Find  $t_{\alpha/2, df}$  using  $qt(1-\alpha/2, df=n-1)$
6. Calculate standard error:  $SE = \frac{s}{\sqrt{n}}$
7. Calculate margin of error:  $moe = t_{\alpha/2, df} \times SE$
8. Construct CI:  $\bar{x} \pm moe$

# Proportion vs Mean

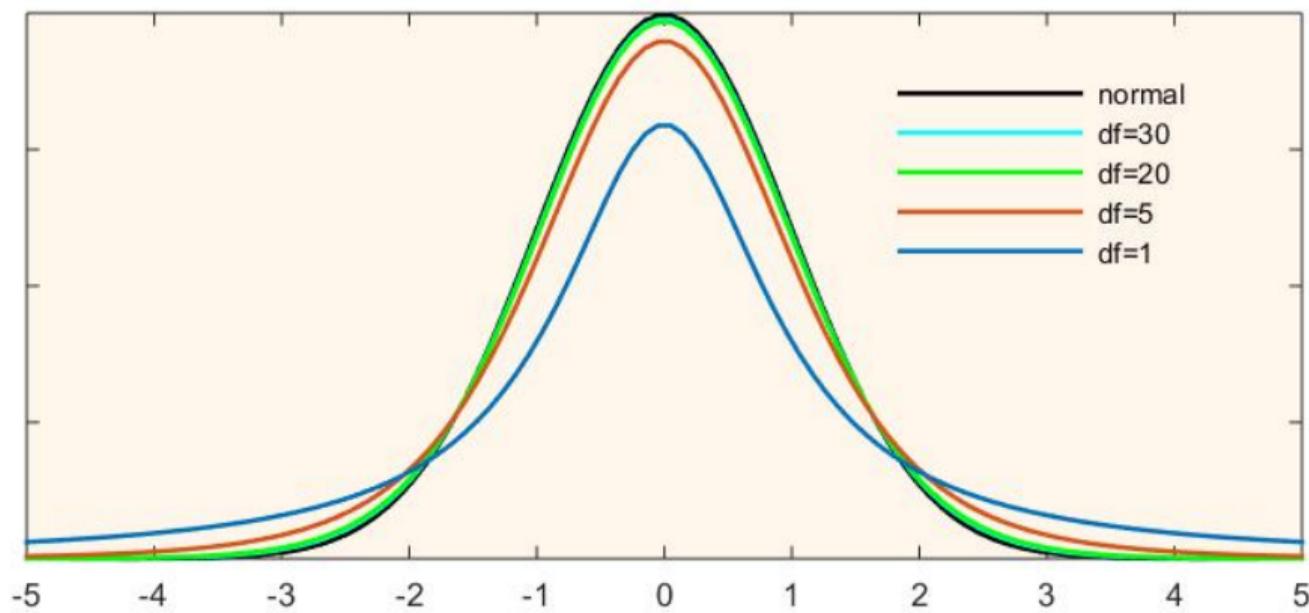
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	Proportion	Mean
Parameter	$p$	$\mu$
Point Estimate	$\hat{p}$	$\bar{x}$
Distribution	Normal	t-distribution
Multiplier	$z_{\alpha/2}$	$t_{\alpha/2, df}$
Standard Error	$\sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$	$\frac{s}{\sqrt{n}}$

**Important Note:** The t-distribution is robust to non-normality, in the absence of outliers.

## Normal vs Student's t-Distribution

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*Image credit: jmp.com*

## R commands:

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### To Find Multipliers:

- $z_{\alpha/2}$ : `qnorm(1-\alpha/2)` or `qnorm(\alpha/2, lower = FALSE)`
- $t_{\alpha/2, df}$ : `qt(1-\alpha/2, df=n-1)` or `qt(\alpha/2, df=n-1, lower = FALSE)`

### To Calculate Entire CI for Proportion:

- `prop.test(x=58, n=389, conf.level=0.95, alternative="two.sided", correct=FALSE)`

### To Calculate Entire CI for Mean:

- `t.test(x = data, conf.level=0.95, alternative="two.sided", correct=FALSE)`

# Interpreting 95% Confidence Intervals

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## Three Valid Interpretations

### 1. Long-run frequency:

"If we repeated this procedure many times, 95% of calculated CIs would contain the true parameter."

### 2. Single sample probability:

"There's a 95% probability that this procedure produces an interval containing the true parameter."

### 3. Statistical significance:

"Values within the 95% CI are not significantly different from our estimate at  $\alpha=0.05$ ."

*Note: The probability statement is about the method, not any particular interval.*

# What Confidence Intervals Don't Mean

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## Frequent Misconceptions

- **Not:** "There's a 95% chance the true value is in this interval"  
*(The interval either contains it or doesn't - no probability after calculation)*
- **Not:** "95% of our data points are in this interval"  
*(It's about the parameter estimate, not individual observations)*
- **Not:** "If we repeat the study, there's 95% chance new estimate will be in this interval"  
*(New intervals will center on new estimates)*

# Writing Hypotheses: $H_0$ and $H_a$

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## Null Hypothesis ( $H_0$ )

- Status quo claim
- Always contains " $=$ "
- Example:  $H_0: \mu = \mu_0$

## Alternative ( $H_a$ )

The claim we hope to find evidence for or against.

Three possible forms:

- $\mu < \mu_0$  (left-tailed)
- $\mu > \mu_0$  (right-tailed)
- $\mu \neq \mu_0$  (two-tailed)

