

# STAT 3011 Discussion 015

## Lab 10: Hypothesis Testing

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# Five Elements

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1. **Assumptions:** Conditions that must be verified
2. **Hypotheses:** Null ( $H_0$ ) and alternative ( $H_a$ )
3. **Test Statistic:** Calculated from sample data
4. **p-value:** Probability measure
5. **Conclusion:** Statistical decision

# Element 1: Assumptions

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## For Population Proportion $p$

1. Random sample
2.  $n \cdot p_0 \geq 15$  AND  $n \cdot (1 - p_0) \geq 15$

## For Population Mean $\mu$

1. Random sample
2. Sample is drawn from a Normal population,  
OR sample is sufficient large to assume normality ( $n \geq 30$ ).

# Element 2: Hypothesis Formulation

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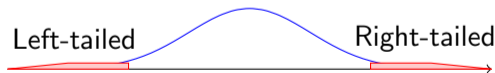
## Null Hypothesis ( $H_0$ )

- Status quo claim
- Always contains “=”
- Example:  $H_0: \mu = \mu_0$
- The statement we want to **disprove**.

## Alternative ( $H_a$ )

Three possible forms:

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



# Element 3: Test Statistic

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## One-Sample t-Statistic (Mean)

$$t^* = \frac{\bar{x} - \mu_0}{s/\sqrt{n}} \sim t_{df=n-1}$$

$$df = n - 1$$

- $\bar{x}$ : Sample mean
- $s$ : Sample standard deviation
- $n$ : Sample size
- $\mu_0$ : Hypothesized population mean
- Follows a standard T-Distribution

## One-Sample z-Statistic (Proportion)

$$z^* = \frac{\hat{p} - p_0}{\sqrt{p_0(1 - p_0)/n}} \sim N(0, 1)$$

- $\hat{p}$ : Sample proportion
- $p_0$ : Hypothesized population proportion
- $n$ : Sample size
- Follows a Standard Normal Distribution

# Element 4: Understanding & Calculating p-values

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## What is a p-value?

- Probability of observing a test statistic *as or more extreme* than what we got,
- **assuming** the null hypothesis  $H_0$  is true.

## Means (t distribution):

- $H_a : \mu > \mu_0$ :  
`pt(t*, df=n-1, lower.tail=FALSE)`
- $H_a : \mu < \mu_0$ :  
`pt(t*, df=n-1, lower.tail=TRUE)`
- $H_a : \mu \neq \mu_0$ :  
`2 * pt(abs(t*), df=n-1, lower.tail=FALSE)`

## Proportions (Normal):

- $H_a : p > p_0$ :  
`pnorm(z*, lower.tail=FALSE)`
- $H_a : p < p_0$ :  
`pnorm(z*, lower.tail=TRUE)`
- $H_a : p \neq p_0$ :  
`2 * pnorm(abs(z*), lower.tail=FALSE)`

# Element 5: Making Conclusions

## Decision Rule

Compare p-value to significance level  $\alpha$ :

$$\text{p-value} \leq \alpha \Rightarrow \text{Reject } H_0$$

$$\text{p-value} > \alpha \Rightarrow \text{Fail to reject } H_0$$

## Common $\alpha$ levels

- 0.01 (1%)
- 0.05 (5%)
- 0.10 (10%)

Never say "Accept  $H_0$ "

We either reject or fail to reject - never prove the null!

# Conducting Hypothesis Tests in R

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## One-Sample t-Test:

```
t.test(x,                # Data vector
       mu = ,            # Null value mu_0
       alternative = ,    # "two.sided", "less", "greater"
       conf.level = )    # 1 - alpha
```

## One-Sample Proportion Test:

```
prop.test(x,              # Number of successes
          n,              # Sample size
          p = ,           # Null proportion p_0
          alternative = ,  # "two.sided", "less", "greater"
          conf.level = ,  # 1 - alpha
          correct = FALSE)
```

# Type I & Type II Errors

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Truth	Fail to Reject $H_0$	Reject $H_0$
$H_0$ is true	Correct Decision	Type I Error ( $\alpha$ )
$H_a$ is true	Type II Error ( $\beta$ )	Correct Decision

**Type I Error:** Rejecting a true  $H_0$

**Type II Error:** Failing to reject a false  $H_0$

Note: P(Type II error) increases as P(Type I error) decreases.

Questions?