# INTRODUCTION TO THE *t* STATISTIC

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### *Z***-SCORES REVISITED**

- Basic Assumptions
  - M approximates μ
  - \*  $\sigma_M$  estimates how well *M* approximates  $\mu$
  - ✤ Quantifies inferences made about the population
- > The Problem with z-scores
  - \* Requires knowledge of  $\sigma$

### **ESTIMATE OF STANDARD ERROR**

estimate of the standard error when the population standard deviation is unknown; estimate of standard difference between M and  $\mu$ 

$$s_M = \sqrt{\frac{s^2}{n}}$$

### **DEGREES OF FREEDOM**

number of scores in a sample that are independent and free to vary

- > Larger  $df \rightarrow$  better  $s^2$  represents  $\sigma^2$
- $\succ$  df associated with  $s^2$  describes how well t represents z
  - ↔ Larger  $df \rightarrow$  better *t* statistic approximates *z* statistic

### t STATISTIC

statistic used to test hypotheses about unknown population  $\mu$  when the value of  $\sigma$  is unknown

May be positive or negative (absolute value = magnitude)
No upper- or lower- bound values

$$z = \frac{M-\mu}{\sigma_M} = \frac{M-\mu}{\sqrt{\frac{\sigma^2}{n}}} \qquad \qquad t = \frac{M-\mu}{s_M} = \frac{M-\mu}{\sqrt{\frac{s^2}{n}}}$$

### t distribution



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### *t* **DISTRIBUTION: PROPORTIONS**

	0.25	Pro 0.10	oportion in 0.05	One Tail 0.025	0.01	0.005
Proportion in Two Tails Combined       df     0.50     0.20     0.10     0.05     0.02     0.01						0.01
1	1.000	3.078	6.314	12.706	31.821	63.657
2	0.816	1.886	2.920	4.303	6.965	9.925
3	0.765	1.638	2.353	3.182	4.541	5.841
4	0.741	1.533	2.132	2.776	3.747	4.604
5	0.727	1.476	2.015	2.571	3.365	4.032
6	0.718	1.440	1.943	2.447	3.143	3.707

#### **INTRODUCTION TO THE T STATISTIC**

### **HYPOTHESIS TESTING WITH** *t*

#### Assumptions

- Values/Observations in sample are independent
- Sampled population is normal
- Process of Hypothesis Testing
  - ↔ Start with population with unknown  $\mu$  and  $\sigma^2$
  - ✤ Goal: Sample to determine effect (if any) of treatment
  - $H_0$ : the treatment had no effect
  - \*  $s^2$  and  $\sigma_s$  computed from sample data

$$t = \frac{M - \mu}{s_M}$$

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### **HYPOTHESIS TESTING WITH** *t*

- $\succ$  Step 1: State H<sub>0</sub> and H<sub>1</sub>
- Step 2: Locate critical region
  - Compute *df* and refer to *t* distribution table
- Step 3: Calculate Test Statistic
  - ✤ Calculate sample  $s^2$
  - $\diamond$  Compute  $s_M$
  - Compute t statistic
- > Step 4: Make decision regarding  $H_0$

$$s^2 = \frac{SS}{df}$$
  $s_M = \sqrt{\frac{s^2}{n}}$   $t = \frac{M-\mu}{s_M}$ 

### **EFFECT SIZES**

#### Cohen's d

 $\rightarrow r^2$ 

$$d = \frac{M - \mu}{s}$$

$$r^2 = \frac{t^2}{t^2 + df}$$

•  $r^2 = 0.01 - \text{Small Effect}$ 

- $r^2 = 0.09 \text{Medium Effect}$
- $r^2 = 0.25 \text{Large Effect}$

## INFLUENCE OF n AND s ON $s_M$

- > Larger  $s_M \rightarrow$  smaller values of t (closer to zero)
- > Any factor that increases  $s_M$  will reduce likelihood of rejecting  $H_0$

#### ➤ Large Variance

- ✤ Difference less likely to be significant
- Scores widely scattered, so harder to see consistent patterns in data
- Reduced effect size

#### Large Sample

- Difference more likely to be significant
- $\diamond$  smaller  $s_M$