

# Confidence Intervals

## Hypothesis testing

### Confidence Intervals

$$\text{estimate} \pm t_c \times \hat{SE}$$

error bounds

$$t_c \times \hat{SE}$$

$$\bar{y} \pm t_c \times \hat{SE}_M \quad \dots \quad \hat{\delta} \pm t_c \times \hat{SE}_D$$

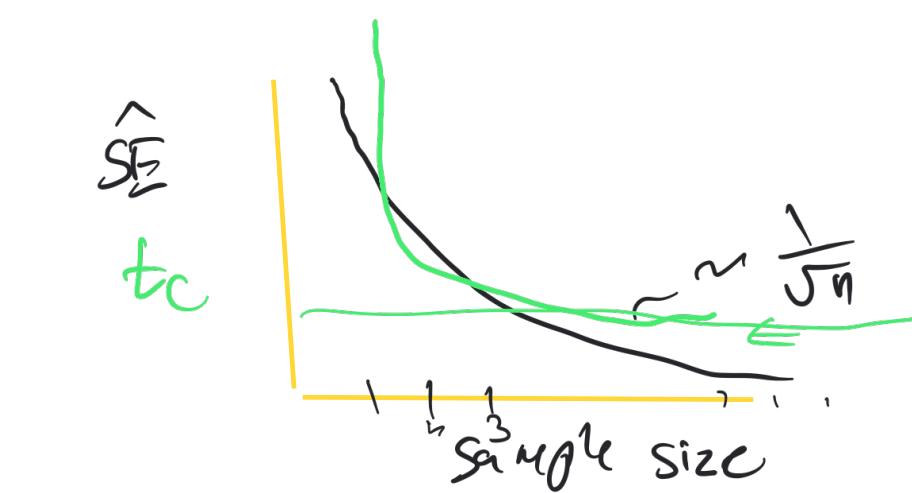
size  $\rightarrow 95\% \rightarrow \alpha = 0.05$  → coverage

$$t_c = t_{\frac{\alpha}{2}, df} = qt(\frac{\alpha}{2}, df, \text{low.tail=F}) \quad \leftarrow df = \text{sample size} - 2$$

$$\hat{SE} = \hat{SE}_D = \sqrt{\frac{s_2^2}{n_2} + \frac{s_1^2}{n_1}}$$

S<sup>2</sup> experimental design  
n's reduce  $\sigma^2$   
n's sample size

$$(S^2) \rightarrow \hat{\sigma}^2 \rightarrow \underline{\sigma^2}$$

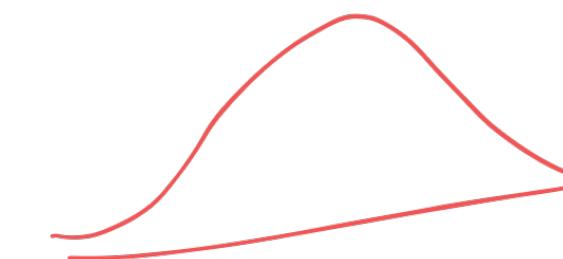


### Diminishing Returns

Sample size	Int. size
2	9.5
4	1.65
8	0.85
16	0.55

$\rightarrow 5.6x$   
 $\rightarrow 1.9x$   
 $\rightarrow 1.6x$

$$Y_i \sim N(\mu, \sigma^2)$$



Is the reference pop'n Normal?

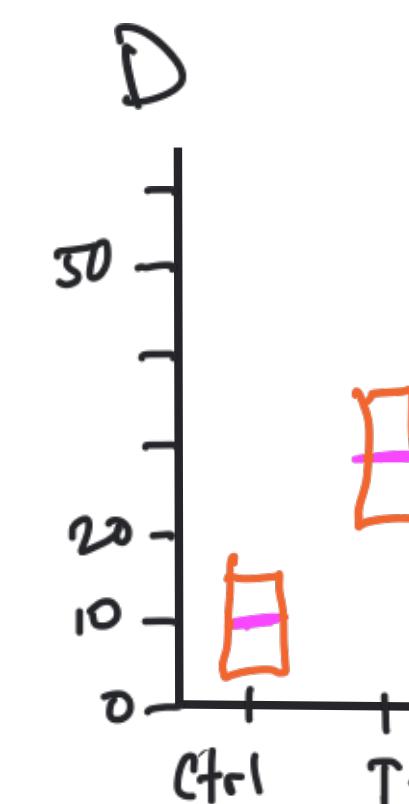
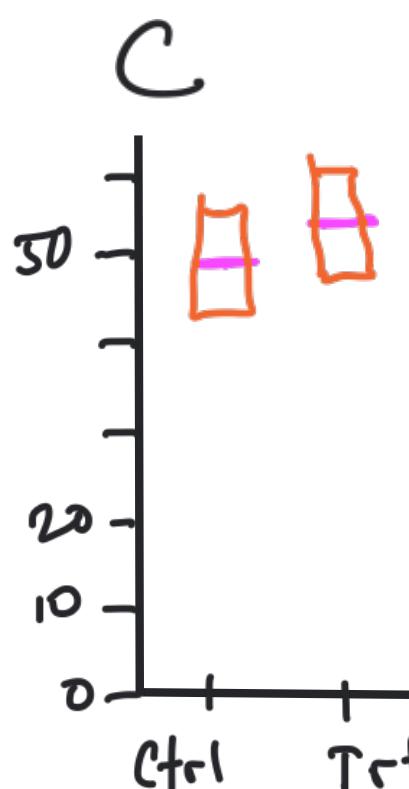
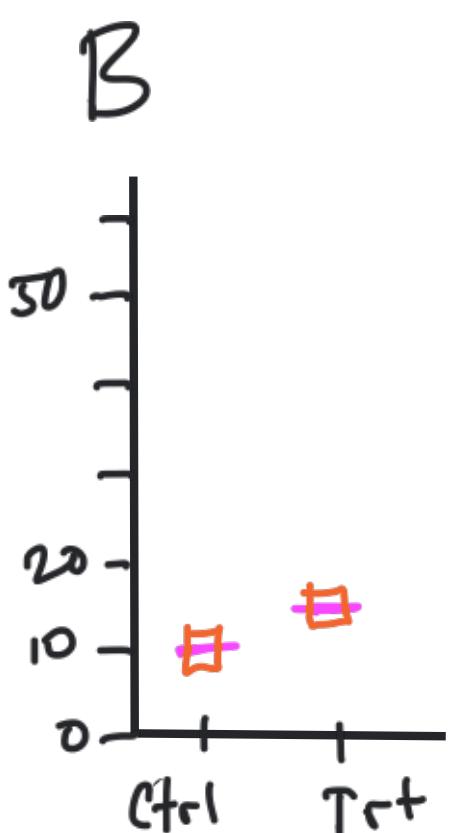
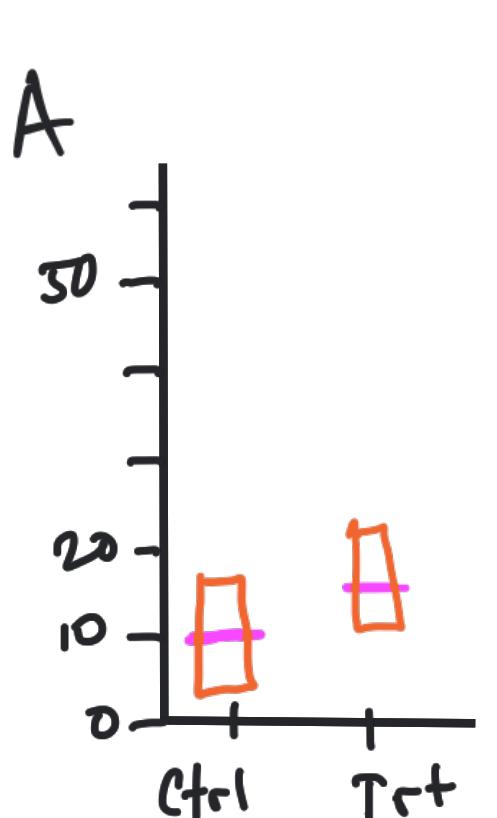
- 1) many are.
- 2)  $\frac{\bar{Y} - \mu}{SEM}$  or  $\frac{\bar{Y} - \delta}{S_{\bar{Y}}} \sim N(0, 1)$  if  $n$  is large

Central limit theorem.

- 3) use our data to check:

- histogram
- Q-Q plots
- ~~t-test for Normality~~

$$n_c = n_t = 32$$



- which effect is largest?
- which effect is most important?
- which effect is most significant?

t-test

p-value

Hypothesis testing

$\delta$  5  
 $\sigma$  8  
 $SE_D$  2

5  
2  
0.5

5  
8  
2

$$+ \frac{\delta}{M_C} = \frac{5}{10} = 0.5$$

$$0.5$$

$$\frac{\delta}{S_D} = 0.1$$

$$2$$

$$\frac{\delta}{SE_D} = \frac{5}{2} = 2.5$$

$$\frac{5}{0.5} = 10$$

$$\frac{\sum}{2} = 0.5 \quad \frac{20}{2} = 10$$

## Confidence Interval

- bounds on error
- how good is our estimate?

Estimate:  $\hat{\delta} = \bar{y}_L - \bar{y}_R$

2) SED :  $\sqrt{\frac{s_L^2}{n_L} + \frac{s_R^2}{n_R}}$

$$\hat{\delta} \pm t_c \times SED$$

cf function

$\alpha, df$

## T-test

- could the TRUE effect be = 0?

Null hypothesis

$$H_0: \delta = 0$$

Statistic

$$\hat{t} = \frac{(\hat{\delta} - 0)}{SED}$$

- observed  
effect  
typical  
error

$|\hat{t}| > 1$  - effect is bigger than  
a typical error

- bigger  $E$ , more confident  $\delta \neq 0$

P-value:  $P(|t| > |\hat{t}|)$

our data  
when  $\delta$  is  
unknown

from random effect

~~when  $\delta = 0$~~

## T-test

calculated  $\frac{\hat{s}}{SE\hat{p}} = \hat{t}$

bigger  $|\hat{t}| \Rightarrow$  smaller p-value

$\Rightarrow$  less plausible that  $s=0$

$H_0$  is TRUE

What do we do w/ T-test?

- Making a decision Y/N      ↗ both  
→ weighing evidence  
- more / less significant

We just consider one of being wrong

		Accept	Reject - Decide $\delta \neq 0$
		$\delta = 0$	
$H_0$	TRUE	✓	X Type I error
	false	X Type II error	✓
	$\delta \neq 0$		

↳  $\delta$  might be 0

false positive  
 $\alpha$

$\alpha$  = Prob of Reject when  $\delta = 0$   
 $\beta$  = Prob of Accept when  $\delta \neq 0$

lost opportunity

false negative

$\beta$

## Decision strategy

1) choose  $\alpha$   $\rightarrow$  calculate  $t_c$   $\leftarrow t_{\alpha_2, df}$   $gt$

if  $|t| > t_c \Rightarrow$  reject H<sub>0</sub>  $\delta \neq 0$

if  $|t| < t_c \Rightarrow$  Accept H<sub>0</sub>.  $\delta$  could be 0.

2) calculate p-value.

$p \leq \alpha \Rightarrow$  Reject

$p > \alpha \Rightarrow$  accept