

Recall: we can estimate a *95% confidence interval* for an unknown population mean  $\mu$  by using the sample mean  $\bar{X}$  and the population standard deviation  $\sigma$ :

$$\bar{X} \pm 1.96 \cdot SE$$

or equivalently

$$\bar{X} \pm 1.96 \cdot \frac{\sigma}{\sqrt{N}}$$

What if we are not given  $\sigma$ ? Can we use our estimate  $\hat{\sigma} = \sqrt{\frac{SS}{N-1}}$ ?

Well, yes – sort of – but we have to adjust the 1.96 part.

Why?

- 1.96 is used because for a normal distribution, 95% of sample means fall between  $-1.96 \cdot SE$  and  $1.96 \cdot SE$ . This assumes  $\sigma$  is known/given.
- if estimating  $\sigma$  via  $\hat{\sigma}$ , we get a  $t$ -distribution for the distribution of sample means. The exact shape of this distribution *depends on sample size*

In light of this, let's define a *generalized* confidence interval:

$$\bar{X} \pm t_{df}^* \cdot SE$$

where

1. the exact value of  $t_{df}^*$  depends on sample size
  - defined as the value of  $t$  which leaves 5% of distribution in the tails (both tails combined) – sometimes called the *critical value* of the  $t$ -distribution
2. the formula for SE depends on design:
  - for single group (or paired samples), we have  $SE = \frac{\hat{\sigma}}{\sqrt{N}}$
  - for independent samples, we have  $SE = \hat{\sigma}_p \sqrt{\frac{1}{N_1} + \frac{1}{N_2}}$

Example 1 (single group design): A sample of 25 people is given a treatment. After treatment, we find  $\bar{X} = 22.2$  with  $SS = 384$ . Construct a 95% confidence interval for  $\mu$ , the population mean for the treatment group.

For independent groups designs, the goal is to estimate the "mean difference"  $\mu_1 - \mu_2$ . The resulting CI is

$$(\bar{X}_1 - \bar{X}_2) \pm t_{df}^* \cdot \hat{\sigma}_p \sqrt{\frac{1}{N_1} + \frac{1}{N_2}}$$

Example 2: Does watching educational TV as a kid predict better high school grades?

Educational TV	No Educational TV
$N_1 = 10$	$N_2 = 10$
$\bar{X}_1 = 93$	$\bar{X}_2 = 85$
$SS_1 = 200$	$SS_2 = 160$

Compute a 95% confidence interval for the mean difference  $\mu_1 - \mu_2$ .

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