Lecture 9: Confidence Interval for One Mean STAT 310, Spring 2021

#### Introduction

- Similar to how we can model the sample proportion  $\hat{p}$  using a normal distribution, the sample mean  $\bar{x}$  can also be modeled using a normal distribution when certain conditions are met.
- ► However, we'll learn that a new distribution, called the *t*-distribution, tends to be more useful when working with the sample mean.
- ▶ In this lecture we'll first learn about this new distribution, and then use it to construct confidence intervals for the mean.

# Central Limit Theorem (CLT) for the Mean

When the sample size n is sufficiently large, the sampling distribution for  $\bar{x}$  follows an approximate normal distribution centered around the population mean  $\mu$ , and with standard error  $SE = \sigma/\sqrt{n}$ 

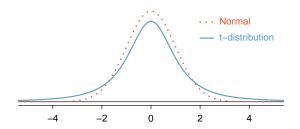
▶ Based on the CLT, we can construct a 95% confidence interval for the population mean  $\mu$  as

$$\bar{x} \pm z^* SE \implies \bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}}$$

- Nowever, one issue with this confidence interval formula is that the standard error is in terms of the population standard deviation  $\sigma$ , which is unknown.
- We can resolve this by plugging in the sample standard deviation s as the estimate of  $\sigma$ . That is, use  $SE \approx s/\sqrt{n}$ .
- ▶ This is a sensible approach when the sample size is large. But when the sample size is small, the confidence interval needs to be adjusted to account for additional uncertainty in estimating  $\sigma$  with s.
- ▶ It turns out that we can get a more accurate confidence interval, which accounts for this additional uncertainty, by using the *t*-distribution to calculate the critical value instead of the *z*-distrubiton.

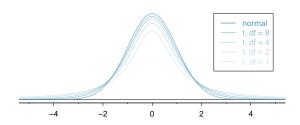
### t-disribution

- ▶ A *t*-distribution is bell-curve shaped distribution that is centered around zero.
- It looks similar to a standard normal distribution, but it has wider tails.



### t-disribution

- ▶ The shape of the t-distribution depends on the degrees of freedom, which is defined as df = n 1
- ▶ When the sample size is small the *t*-distribution has noticeably wider tails than a normal distribution.
- ▶ When the sample size is large (about 30 or more), the *t*-distribution is nearly identical to a normal distribution.



# Confidence Interval for $\mu$

A confidence interval for the population mean  $\mu$  is given by

$$\bar{x} \pm t^* \frac{s}{\sqrt{n}}$$

The critical value  $t^*$  is found using t-distribution. It depends on the confidence level and degrees of freedom, and can be computed using the R function qt().

*Example*: Calculate the critical value  $t^*$  when the confidence level is 95% and n = 15.

## Confidence Interval for $\mu$

The confidence interval formula is valid if the following conditions are satisfied:

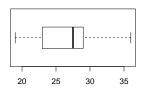
- ► The data come from a random sample. (This is called the **independence condition** in the textbook.)
- ▶ The sample size n is large ( $n \ge 30$ ). Otherwise, if the sample size is small (n < 30), the data should have an approximate normal distribution. (This is called the **normality condition** in the textbook.)
- Additionally, the data should not contain any extreme outliers.

Graphical methods (box plots, histograms) can be used to check if the data have an approximate normal distribution when the sample size n is small.

### Example

Below are some summary statistics and a box plot for the ages of a random sample of n=26 female athletes who participated in the 2012 Olympic Games in London. Calculate and interpret a 95% confidence interval for the population mean age. Also comment on whether the conditions for the confidence interval appear satisfied.

n	$\bar{x}$	S	min	max
26	26.9	4.5	19	36



## Example

Calculate and interpret a 99% confidence interval for the population mean age of female athletes who participated in the 2012 Olympic Games.