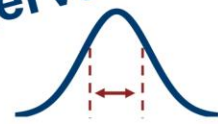


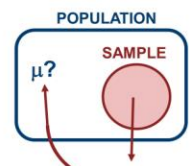
# Introduction to confidence intervals

Useful for estimating the mean



## Estimating the population mean

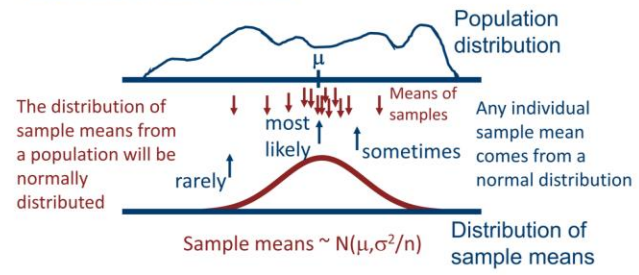
We want to know the population mean. We can't measure the whole population. We take a random sample. We calculate a sample mean.



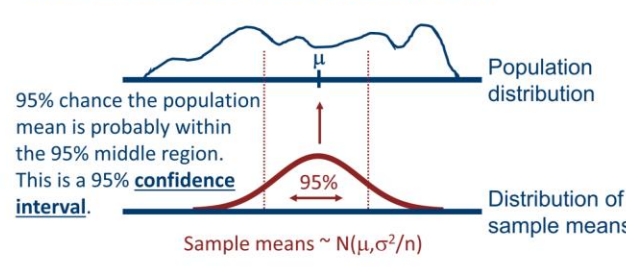
The sample mean is an estimate of the population mean, but sampling error makes it inexact.

What can we say about the likely population mean, based on the sample mean?

## The Central Limit Theorem



## Confidence intervals for the population mean

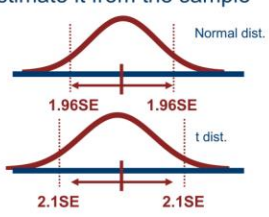


## Confidence interval reality

Unfortunately in the real-world we would never have the population variance. We must estimate it from the sample variance.

To adjust for sampling error, we use the t-distribution instead of the normal.

The t-distribution is wider to account for the sampling error.

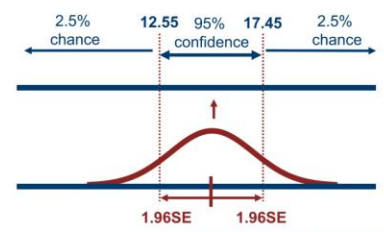


## Confidence interval example (part 1)

e.g., if we take a sample of 16 values,  $\bar{x} = 15, \sigma = 5$

The 95% confidence interval is  $\bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}}$

Which is  
 $15 \pm (1.96) \frac{5}{\sqrt{16}}$   
 $15 \pm 2.45$   
 $\{12.55, 17.45\}$

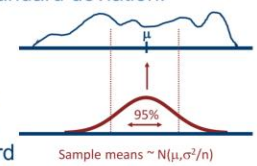


## Calculating the confidence interval for the population mean

Take a sample of n values.

Calculate the mean, variance, and standard deviation.

Get middle region of the standard normal distribution corresponding to the degree of confidence desired.



This tells the width of the confidence interval in terms of number of standard errors above and below the sample mean

## Standard error vs standard deviation

Population values  $\sim N(\mu, \sigma^2)$

Sample values  $\sim N(\bar{x}, s^2)$

Standard deviation describes the spread of the data values

Sample means  $\sim N(\mu, \sigma^2/n)$   $s_{\bar{x}}^2 = \frac{\sigma^2}{n}$  therefore  $s_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$

**Standard error**,  $s_{\bar{x}}$ , describes the spread of the sample means and, likewise, the spread of possible population means.

## The t-distribution

The t-distribution is wider to account for the sampling error.

As sample size increases, it narrows to become the normal.

There are t-tables like Z tables

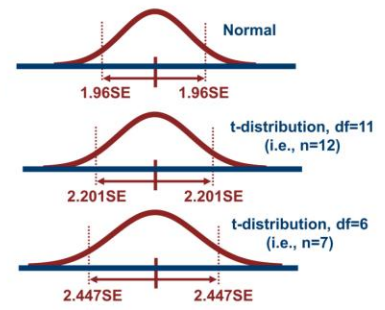


Table of z values

## Z and t tables

Z tables usually describe area to the left (versatility).

t tables usually describe  $\alpha$  area on outside (for confidence intervals) and have rows for each df (n-1)

Table of t values

## Confidence interval example (part 2)

e.g., if we take a sample of 16 values,  $\bar{x} = 15, s = 5$

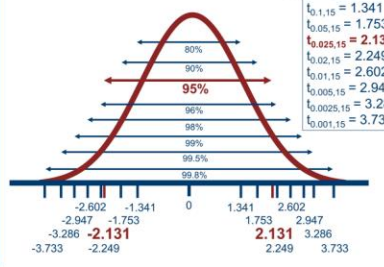
$n=16, df=15$

95% middle,  $\alpha=0.025$  on each side

Interval is **2.131** standard errors wide

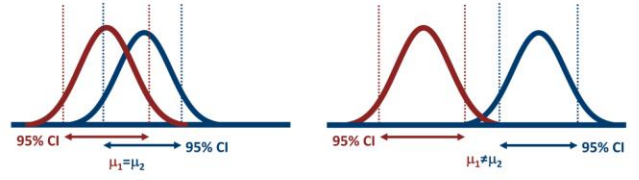
Table of t values

## Confidence interval example (part 2)



## What are confidence intervals used for?

- As a descriptive statistic for our estimate of the population mean.
- As the basis for the t-test, a test of whether two populations appear to have different means.

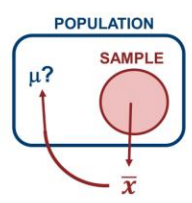


## Which method to use to calculate CI

If we know  $\mu$  and  $\sigma$   
No stats needed

If we know  $\bar{x}$  and  $\sigma$   
Use normal distribution (unrealistic)

If we know  $\bar{x}$  and  $s$   
Use t-distribution



Confidence intervals are the real-world version of significant figures

## Confidence interval (CI) comparison

If we take a sample of 16 values,  $\bar{x} = 15, s = 5$

Using the normal distribution:

The 95% CI:  $\bar{x} \pm 1.96 \frac{\sigma}{\sqrt{n}}$ ,  $15 \pm 2.45$ ,  $\{12.55, 17.45\}$

Using t-distribution:

The 95% CI:  $\bar{x} \pm 2.131 \frac{s}{\sqrt{n}}$ ,  $15 \pm 2.66$ ,  $\{12.34, 17.66\}$

We should always use the t-distribution unless the sample size is large (e.g.,  $df=25$  gives  $\sim 5\%$  error, 2.06 vs 1.96).

## Confidence interval example (part 2)

e.g., if we take a sample of 16 values,  $\bar{x} = 15, s = 5$

The 95% confidence interval is  $\bar{x} \pm 2.131 \frac{s}{\sqrt{n}}$

Which is  
 $15 \pm (2.131) \frac{5}{\sqrt{16}}$   
 $15 \pm 2.66$   
 $\{12.34, 17.66\}$

