

## Confidence Intervals - Introduction

### Key Concepts

- 1) Confidence Interval (CI) - A range of values likely to contain the population parameter
- 2) Why CI is Not a Probability - A confidence interval does not mean there is a 95% probability that the interval contains the parameter

### Example estimating mean height

- You want to estimate the mean height of 10 year old girls in the UK
- A sample of 16 gives a mean height of 90 pounds (point estimate)
- The point estimate alone lacks information about uncertainty
  - It does not tell how close it is to the true population mean

### Confidence intervals provide more information

A 95% confidence interval (CI) means that if we repeatedly took samples and calculated CIs, 95% of them would contain the true mean

### Example of a 95% CI

$$72.85 < \mu < 107.15$$

Interpretation: The mean is likely within this range 5% of the time, a CI will not contain the true mean



Why confidence intervals are not probabilities

- A 95% of CI does not mean that there is a 0.95 probability that the true mean is within the interval.

Key Issues

- Other prior information (eg previous studies showing means above 110) can affect interpretation
- Different methods can produce different CIs, that contain the population mean 95% of the time.
- The standard CI method is commonly used because it is symmetrical and contiguous around the estimate

Confidence Interval on the mean

Understanding confidence intervals for the mean

A confidence interval (CI) on the mean estimates population mean ( $\mu$ ) based on a sample mean ( $M$ ). Since the true mean is unknown, a CI gives a range where the true mean likely falls.

Example known population standard deviation ( $\sigma$ )

Assume the heights of 20 year old children are normally distributed.

Population mean ( $\mu$ ) = 90

Population standard deviation ( $\sigma$ ) = 36

Sample size ( $N$ ) = 9

Step: compute standard error (SE)

$$\sigma_M = \frac{\sigma}{\sqrt{N}} = \frac{36}{\sqrt{9}} = 12$$

The standard error ( $\sigma_M$ ) measures how much sample means vary from the population mean.



Step 2: Compute 95% Confidence Interval

For a normal distribution, 95% of values fall within 1.96 standard deviation of the mean

$$\text{Lower Limit} = 90 - (1.96 \times 12) = 66.48$$

$$\text{Upper Limit} = 90 + (1.96 \times 12) = 113.52$$

Thus, the 95% confidence interval is (66.48, 113.52)

This means that if we repeatedly sample from the population 95% of computed CIs will contain the true mean.

Example: When  $\sigma$  is Unknown (Use t distribution)

When the population standard deviation ( $\sigma$ ) is unknown we estimate it from the sample standard deviation ( $s$ ) and use the t-distribution.

Step 1: Compute sample mean and standard error

Given sample 2, 3, 5, 6, 9

$$\text{Sample Mean } (\bar{M}) = 5$$

$$\text{Sample Variance } (s^2) = 7.5$$

Estimated standard error

$$s_m = \frac{s}{\sqrt{N}} = \frac{2.5}{\sqrt{5}} = 1.118$$

Step 2: Find t value for 95% CI ( $df = N - 1 = 4$ )

From a t-table for  $df = 4$ , the t value for 95% confidence is 2.776.



Step 3: compute 95% confidence interval.

$$\text{lower limit} = 5 - (2.776 \times 1.225) = 1.60$$

$$\text{upper limit} = 5 + (2.776 \times 1.225) = 8.40$$

Thus the 95% confidence interval is (1.60, 8.40)

When to use  $t$  vs Normal distribution

- Use Normal distribution ( $z$ -values) when  $\sigma$  is known.
- Use  $t$  distribution when  $\sigma$  is unknown and estimated from sample data.
- As sample size increases ( $N \geq 100$ ), the  $t$ -distribution approaches the normal distribution.

T-Distribution key concepts

1. Difference Between the  $t$ -Distribution and Normal Distribution  
 The  $t$  distribution is similar to the Normal distribution but has heavier tails (leptokurtic), meaning it has more extreme values.
  - The shape of the  $t$ -distribution depends on the degrees of freedom (df).
  - As df increases, the  $t$ -distribution becomes more like the standard normal distribution ( $z$ -distribution).

3. Finding  $t$ -values from the  $t$ -Table

- The  $t$ -table provides critical  $t$ -values for different confidence levels and df.
- Example for 95% confidence and  $df = 8$ , the  $t$ -value is 2.306 (from table I)



- For higher confidence levels (99%) the  $t$  values are larger since a wider interval is needed to capture the population mean.

#### 4) Using a $t$ -calculator

- A  $t$ -calculator helps determine the area in the tails of a  $t$ -distribution for any given score.
- example: If  $df = 8$ , the probability that the sample mean falls within 1.96 standard error is 0.914 compared to 0.95 in the normal distribution.
- This confirms that using an estimated standard deviation increases uncertainty, requiring wider confidence intervals.