

Chapter 16

Confidence Intervals

The Basics

Intended Learning Outcomes

- Explain why a point estimate alone is not enough
- Derive the logic of a confidence interval from the sampling distribution
- Interpret the confidence level correctly
- Find critical values z^* for common confidence levels
- Compute a confidence interval for μ when σ is known
- Describe how z^* , n , and σ affect the margin of error
- Determine the sample size needed for a target margin of error
- Use a confidence interval to evaluate a claim about μ
- Transform a CI using a linear or monotonic function

The Running Question

Context: CMHC (Canada Mortgage and Housing Corporation) reports the average monthly rent for a 1-bedroom apartment in London, ON is \$1,200 per month.



The Running Question




\$1,450.00 ♡

1-Bedroom unit Available in Aylmer!!


Aylmer, London · 2 d

1-Bedroom unit Available in Aylmer!! Interested in taking a look at this cozy 1.5-bedroom, 1-...

🚗 1.5 🏠 1 🏠 Apartment 📄 1 🌟 Yes




\$1,375.00 ♡

[RentRevo - 1 Bedroom Apartment for Rent](#) 

London · 2 d

TWO MONTHS FREE & \$500 GIFTCARD!! MOVE IN TODAY! *on selected units. These spacious...

🚗 1 🏠 1 🏠 Apartment 📄 0 🌟 Yes



\$1,324.00 ♡

London 1 Bedroom Apartment for Rent: 

London · 2 d

Building Description Promotions | First month FREE rent when you sign a new 12-month lease...

🚗 1 🏠 1 🏠 Apartment 📄 0 🏠 532 sqft

[Virtual Tour](#)

A sample of $n = 47$ current Kijiji listings finds:

$$\bar{x} = \$1,467 \quad \sigma = \$250$$

Is the CMHC figure still accurate? By the end of this chapter, we will have a formal method to answer this question.

PART 1

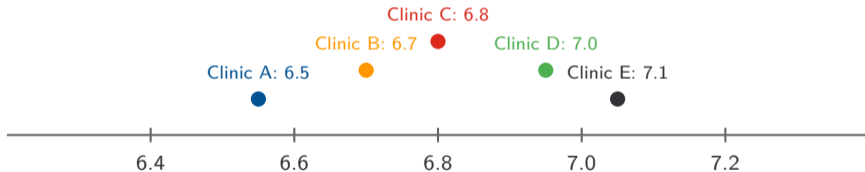
Why a Point Estimate Is Not Enough

What can a single number really tell us?

The Problem with a Single Number

A university health clinic surveys 100 students about nightly sleep and reports: $\bar{x} = 6.8$ hours.

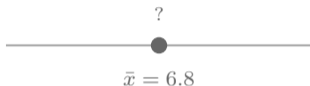
Sleeping an average of 6 or fewer hours per night is associated with adverse health outcomes. Is this a problem? We can't tell from the point estimate alone.



Every sample gives a different \bar{x} . A single number tells us nothing about *how close* it is to μ .

What We Really Want

Point estimate alone



Best single guess for μ ,
but no measure of precision.



Confidence interval



A **range of plausible values**
with a stated confidence level.

Principle: A confidence interval adds precision to the point estimate: a range of plausible values for μ together with a stated confidence level.

PART 2

From Sampling Distribution to Confidence Interval

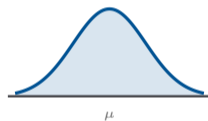
How does the sampling distribution help us build a range of plausible values?

Simple Conditions for Inference About a Mean

1. **Simple random sample (SRS)** from the population. The population is large relative to the sample.



2. The variable has an exactly **Normal distribution** with mean μ and standard deviation σ .



3. We don't know the population mean μ , but we **do know** the population standard deviation σ .

σ ✓

μ ?

Student Sleep Study

Example 16.1

A university health clinic wants to estimate mean nightly sleep among students.

Prior large-scale health surveys give a population standard deviation of $\sigma = 1.5$ hours.

The clinic surveys an SRS of $n = 100$ students and records each participant's nightly sleep duration.

Sample size	$n = 100$
Sample mean	$\bar{x} = 6.8$ hours
Population SD	$\sigma = 1.5$ hours
Goal	Interval estimate of μ

Recall: The Sampling Distribution

From our work on sampling distributions, we know:

If the population is Normal with mean μ and standard deviation σ , then the sample mean \bar{X} from an SRS of size n follows

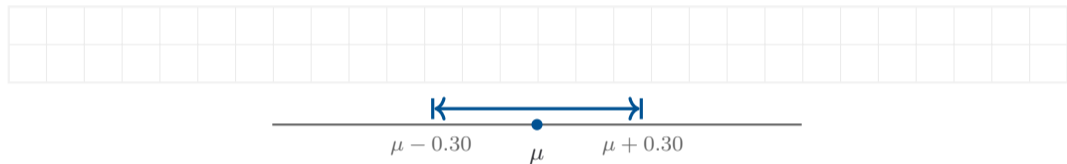
$$\bar{X} \sim N\left(\mu, \frac{\sigma}{\sqrt{n}}\right)$$

Calculate: For the sleep study, the standard error of \bar{X} is:

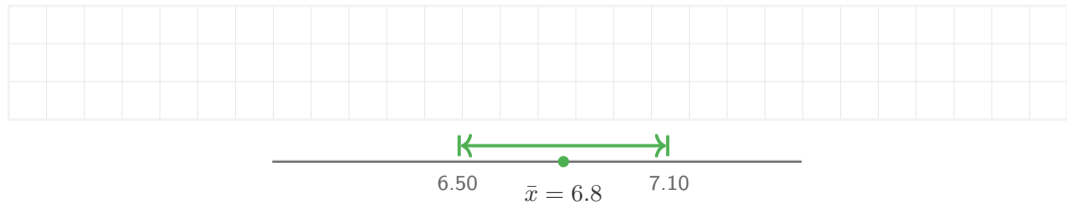
The Logic of CIs

Example 16.2

By the CLT, approximately 95% of sample means will fall within 0.30 units of the true mean μ .



But distance is symmetric, so



Why Does This Work?

The logic rests on one simple observation:

$$|\bar{x} - \mu| < 0.30 \quad \text{is the same statement as} \quad |\mu - \bar{x}| < 0.30$$

“ \bar{x} is close to μ ” and “ μ is close to \bar{x} ” are the *same* thing.

Since the sampling distribution guarantees the first statement holds in 95% of samples, the second statement also holds in 95% of samples.

So the interval $\bar{x} \pm 2(\sigma/\sqrt{n})$ captures μ in about 95% of all possible samples.

Confidence Interval

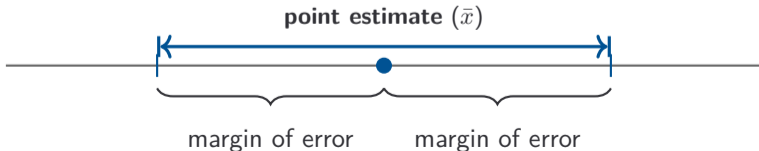
Confidence Interval

A level C confidence interval for a parameter has three components:

1. A **point estimate**: a single number calculated from the data that serves as the best guess for the unknown parameter.
2. A **margin of error**: the amount added to and subtracted from the point estimate to form the interval:

$$\text{point estimate} \pm \text{margin of error}$$

3. A **confidence level** C : the probability that the method produces an interval capturing the true parameter in repeated samples.

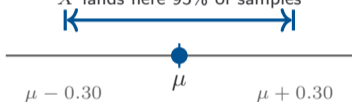


Probability vs. Confidence Language

\bar{X} is a random variable

before data is collected

\bar{X} lands here 95% of samples



Fixed: μ

Varies: \bar{X}

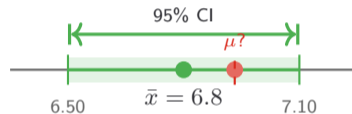
Terminology:



collect
sample

\bar{x} is a fixed number

after data is collected



Fixed: $\bar{x} = 6.8$

Unknown: μ

Terminology:

What Does “95% Confident” Mean?

Confidence Level

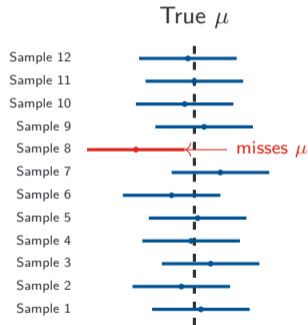
Repeat many times: $C\%$ of all resulting intervals will contain μ .

! Common misconception:

Not: “ μ has a 95% chance of being in *this* interval.”

Since μ is fixed, it's either captured or it isn't.

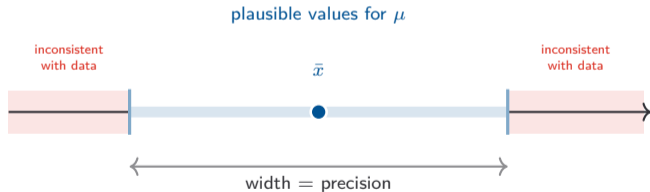
The 95% describes the *method*, not this specific interval.



Each row = one sample's 95% CI

What the Interval Tells You

- The true average nightly sleep is **plausibly near the centre**: not 3 hours, not 9 hours, but somewhere around 6.8 hours.
- Values **outside** are inconsistent with your data
- **Width** signals precision:



Acting on the Interval

Principle: Because the method is reliable, you are justified in **acting as if** the interval contains μ , not because you have proved it, but because you are using a trustworthy procedure.

Common misconception: Hedging does not mean paralysis. It stops you from **overclaiming**.

Act on the interval; do not treat it as certainty.

Paralysis

Cannot be certain,
so refuses to decide



Act on it

Plans and acts
based on the interval



Overclaiming

Claims the mean is
definitely 6.8 hours



CI in the Wild

Public Health

CI for mean daily sodium intake: [2,800, 3,400] mg

Recommended limit: 2,300 mg. Entire interval exceeds the limit.

Decision: Launch a sodium reduction campaign.

Justified regardless of where exactly in [2,800, 3,400] the truth sits.

Engineering

CI for mean bolt diameter: [9.97, 10.03] mm

Tolerance spec: 10.00 ± 0.05 mm. Entire interval within spec.

Decision: Sign off on the batch.

No need to prove the mean is exactly 10.00. The plausible range is acceptable.

CI in the Wild (cont.)

Urban Planning

CI for mean commute time: $[18, 26]$ minutes

Schedules bus frequency.

The whole interval shows the system is nowhere near the 45-minute threshold that would trigger a major infrastructure review.

Decision: No need to know whether the true mean is 21 or 23.

Clinical Medicine

CI for mean blood pressure reduction from a new drug: $[4, 12]$ mmHg

Even at the low end, 4 mmHg is clinically meaningful.

Decision: Prescribe the drug: acting on the interval, without needing to know whether the true effect is 4 or 12.

The Key Question

None of these researchers are claiming: “The true mean is X .”

They are asking:

Does the entire interval, or enough of it, support a decision?

Sometimes clear

Entire interval lies above or below a threshold

→ act on it

Sometimes not

Interval straddles the threshold

→ collect more data

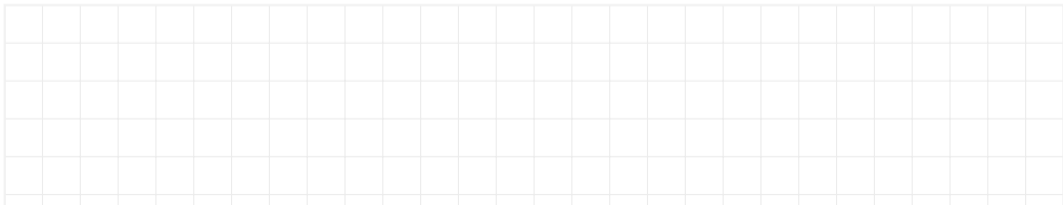
Check Your Understanding

A campus transit study reports:

We are 95% confident that the mean one-way commute time for Western students is between 18 and 26 minutes.

Which interpretation is correct?

- a) There is a 95% probability that μ lies between 18 and 26 minutes.
- b) If we repeated sampling many times, about 95% of the resulting intervals would contain μ .
- c) 95% of individual students commute between 18 and 26 minutes.
- d) The sample mean falls between 18 and 26 minutes in 95% of samples.



PART 3

Making It Precise

How do we move from “roughly 2 standard deviations” to an exact formula?

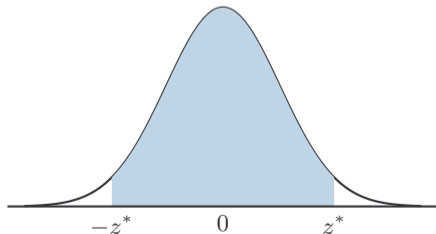
The Critical Value z^*

In the sleep example we used “2 standard deviations” for roughly 95% confidence. The exact multiplier depends on the confidence level we want.

Critical Value z^*

The **critical value** z^* is the number such that the standard Normal curve has area C between $-z^*$ and z^* :

$$P(-z^* < Z < z^*) = C$$



The Confidence Interval Formula

Context: SRS of size n from a Normal population with unknown mean μ and known standard deviation σ .

Confidence Interval for μ (σ Known)

1. Choose a confidence level C and find the corresponding critical value z^* .
2. Compute the margin of error: $\text{MOE} = z^* \cdot \frac{\sigma}{\sqrt{n}}$
3. The level C confidence interval for μ is:

$$\bar{x} \pm \text{MOE} = \bar{x} \pm z^* \cdot \frac{\sigma}{\sqrt{n}}$$

PART 4

How Confidence Intervals Behave

What makes a confidence interval wider or narrower?

Effect of Changing the Confidence Level

Example 16.7

Context: Café example with $\bar{x} = 4.2$, $\sigma = 2.0$, $n = 64$.

Task: Compute the margin of error at three confidence levels.

Confidence level	z^*	Margin of error	Interval
90%	1.64	<input type="text"/>	<input type="text"/>
95%	1.96	<input type="text"/>	<input type="text"/>
99%	2.58	<input type="text"/>	<input type="text"/>

 **Principle:** Higher confidence \Rightarrow wider interval.

PART 5

Choosing the Sample Size

How many observations do we need?

Solving for Sample Size

If you want a specific margin of error MOE, you can solve for the required sample size n .

⚙️ Sample Size for a Desired Margin of Error

1. Start with the margin of error formula: $MOE = z^* \cdot \frac{\sigma}{\sqrt{n}}$

2. Solve for n and round up:

$$n = \left\lceil \left(\frac{z^* \cdot \sigma}{MOE} \right)^2 \right\rceil$$

3. The ceiling $\lceil \]$ means **always round up**, because rounding down gives a margin of error slightly above the target.

PART 6

Using CIs to Evaluate Claims

Can a confidence interval tell us whether a claim about μ is plausible?

Evaluating a Claim with a Confidence Interval


Suppose someone claims the population mean is μ_0 . We can use a confidence interval to assess that claim.

If μ_0 is inside the CI:

The claimed value is among the plausible values for μ . The data are **consistent** with the claim.

If μ_0 is outside the CI:

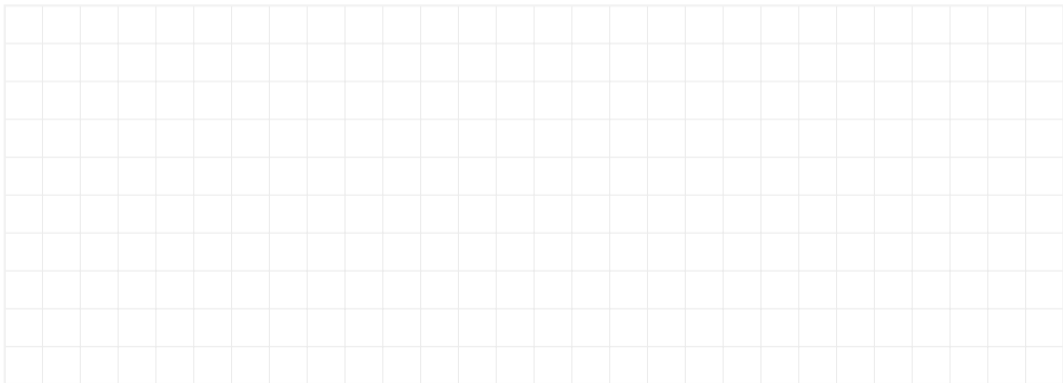
The claimed value is **not** among the plausible values. The data provide evidence **against** the claim.

 **Principle:** If the CI excludes μ_0 , the data provide evidence against the claim $\mu = \mu_0$. The CI also shows the full range of plausible values, not just a yes/no verdict.

Exercise: What If $\bar{x} = 249$?

Context: Same juice setting ($\mu_0 = 250$, $\sigma = 12$, $n = 36$), but now suppose $\bar{x} = 249$.

- Build a 95% confidence interval for μ .
- Does the interval contain $\mu_0 = 250$?
- Do the data provide evidence against the company's claim?



PART 7

Transforming Confidence Intervals

What if the quantity of interest is a function of μ ?

Transforming a Confidence Interval

Suppose you have a CI for μ_X , but the quantity of interest is $Y = g(X)$, where g is a **monotonic** (always increasing or always decreasing) function.


CI Under a Monotonic Transformation

1. Construct the level C CI for μ_X as usual. Call the endpoints L and U .
2. Apply g to **both endpoints**:

g increasing: $(g(L), g(U))$

g decreasing: $(g(U), g(L))$

3. The confidence level is **unchanged**.

 **Principle:** When g is decreasing, the endpoints **flip**: the smaller input maps to the larger output.

Linear Transformations of a CI

The most common case: $Y = aX + b$ (scaling and/or shifting).

⚙️ CI Under a Linear Transformation

1. Build a level C CI for μ_X : $[L, U]$.
2. Apply the transformation to **both endpoints**:

$$\text{CI for } \mu_Y = [aL + b, aU + b]$$

3. If $a < 0$, the endpoints **swap order**.

💡 Intuition: A linear function stretches and shifts the number line uniformly, so the confidence level is preserved exactly. The new standard deviation is $\sigma_Y = |a| \sigma_X$, which the transformation handles automatically.

Common examples: unit conversions (Celsius to Fahrenheit), rate calculations (steps to calories), currency exchange.

Fleet Fuel Efficiency

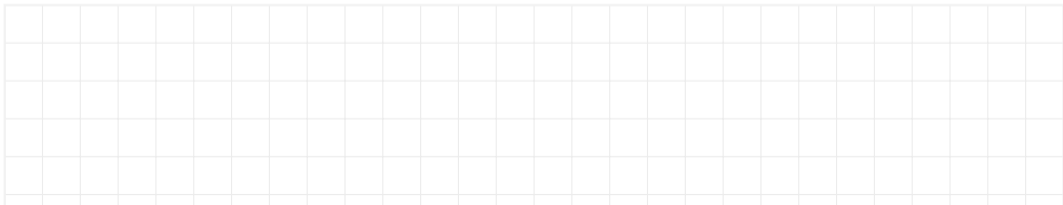
Example 16.15 — Setup

Context: A municipal fleet manager measures fuel consumption X (L/100 km) for city buses. To report to council, efficiency must be expressed in km/L:

$$Y = \frac{100}{X}$$

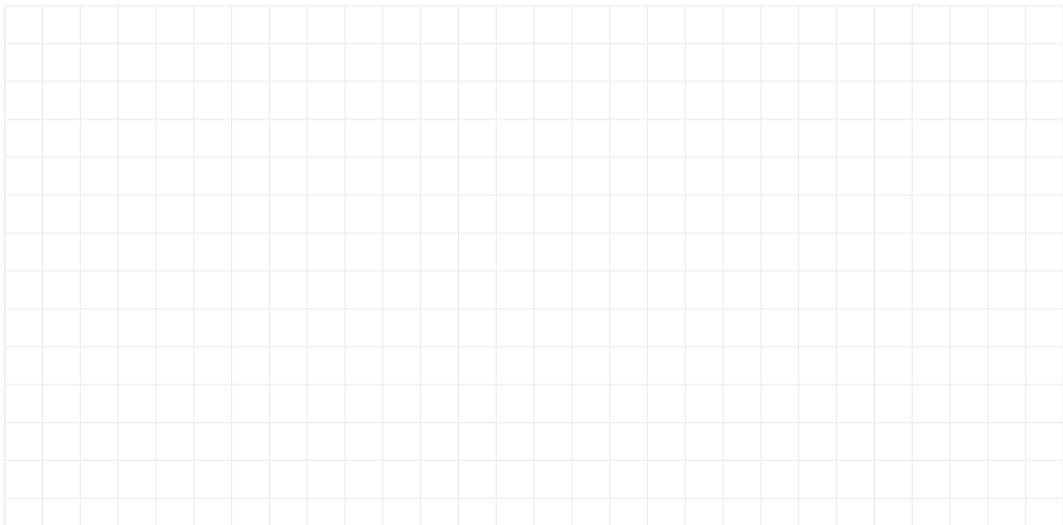
An SRS of $n = 49$ buses gives $\bar{x} = 8$ L/100 km. Industry fleet data give $\sigma = 1.4$ L/100 km.

- Construct a 95% CI for the mean fuel consumption μ_X .
- Use $Y = 100/X$ to construct a 95% CI for the fleet's mean fuel efficiency in km/L.
- The manufacturer advertises an efficiency of 12.5 km/L. Do the data support this claim?



Fleet Fuel Efficiency

Example 16.15 — Solution



CHAPTER 16

Summary

Key formulas and concepts from this chapter

Chapter 16 Summary — Formulas & Reference

■ Key formulas

- CI for μ : $\bar{x} \pm z^* \frac{\sigma}{\sqrt{n}}$
- Margin of error: $\text{MOE} = z^* \frac{\sigma}{\sqrt{n}}$
- Sample size: $n = \left\lceil \left(\frac{z^* \sigma}{\text{MOE}} \right)^2 \right\rceil$

■ Transformations

- If CI for μ is (L, U) , then CI for $g(\mu)$ is

$$(g(L), g(U))$$

when g is increasing, and

$$(g(U), g(L))$$

when g is decreasing.

■ Common z^* values

- 90%: $z^* = 1.64$
- 95%: $z^* = 1.96$
- 99%: $z^* = 2.58$

Chapter 16 Summary — Key Ideas

■ Core concepts

- A point estimate alone gives no sense of precision; a CI adds that information
- The confidence *level* is the method's long-run success rate, not the probability for one interval
- Higher confidence \Rightarrow wider interval
- Larger $n \Rightarrow$ narrower interval

■ Evaluating claims

- If $\mu_0 \notin \text{CI}$, the data provide evidence against the claim $\mu = \mu_0$
- If $\mu_0 \in \text{CI}$, we do not have evidence against $\mu = \mu_0$ (but we have not proven it)

Practice: Varsity Vertical Jump

Example 16.17

Context: A sports science lab measures vertical jump heights for varsity basketball players. A national database gives $\sigma = 8$ cm.

An SRS of $n = 25$ players on a university team gives $\bar{x} = 62$ cm.

1. Construct a 95% confidence interval for the team's mean vertical jump.
2. The national average for varsity players is 58 cm. Do the data provide evidence that this team jumps higher than the national average?
3. How many players would need to be tested to achieve a margin of error of at most 2 cm?



Practice: Greenhouse Temperature

Example 16.18

Context: A greenhouse monitors hourly air temperature ($^{\circ}\text{C}$) for tomato seedlings. Long-term sensor data give $\sigma = 2.5^{\circ}\text{C}$.

An SRS of $n = 25$ hourly readings gives $\bar{x} = 23^{\circ}\text{C}$.

1. Construct a 95% CI for the mean temperature in $^{\circ}\text{C}$.
2. The seed packet lists the ideal temperature in Fahrenheit: $F = 1.8C + 32$. Transform the CI to $^{\circ}\text{F}$.
3. The recommended mean temperature is 72°F . Do the data support the greenhouse meeting this target?

