

STA13: Elementary Statistics

Lecture 13

Book Sections 5.1-5.2

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An **estimator** is a rule or formula that tells how to calculate an estimate of a parameter based on the sample.

- ▶ The sample mean \bar{x} is an estimator of the pop. mean μ
- ▶ The sample median is another estimator of μ
- ▶ The sample variance s^2 is an estimator of σ^2

Some Vocabulary

The term **estimator** will refer to the type of statistic, rule, or formula that is used to estimate a parameter.

- ▶ "The estimator used to estimate the average age of CA voters is the sample mean based on a sample of 500 voters."

The term **estimate** will refer to the specific value of the estimator based on the sample.

- ▶ "The sample estimate of the average age of CA voters is 42 years."

The value of an estimator...

- ▶ depends on a random sample,
- ▶ varies from sample to sample,
- ▶ varies according to a sampling distribution.

A **point estimator** is a single number calculated from the sample to estimate the parameter.

Basic Idea:

Based on the sample, give the most likely value of μ .

The sample mean \bar{x} is a point estimator of μ .

The mean is the "average" or the "expected" value of an estimator.

This affects the accuracy of the estimation.

- ▶ $\text{mean} > \text{parameter} \Rightarrow \text{systematic overestimation}$
- ▶ $\text{mean} < \text{parameter} \Rightarrow \text{systematic underestimation}$

We say an estimator is **unbiased** if its expected value (or mean) is equal to the parameter being estimated.

Otherwise we say an estimator is **biased**.

An unbiased estimator does not systematically overestimate or underestimate the parameter.

\bar{x} is an unbiased estimator of the population mean μ .

- ▶ \bar{x} estimates μ
- ▶ the expected value (or mean) of \bar{x} is μ

Standard Error of an Estimator

The SE of an estimator is its **average distance from its mean**.

For an unbiased estimator, the SE tells
"how far from the truth" it is on average.

Estimators with smaller SE are preferred (smaller spread).

FACT: \bar{x} has the smallest SE of all unbiased estimates of μ .

Standard Error of an Estimator

When the parameters are unknown (most of the time), we use the sample to compute the standard error.

Sample mean \bar{x} :

- ▶ True SE: $\frac{\sigma}{\sqrt{n}}$ (use this if σ is known)
- ▶ Approximate SE: $\frac{s}{\sqrt{n}}$

An **interval estimator** is an interval calculated from the sample, meant to contain the parameter with a high (pre-specified) probability.

Basic Idea:

Based on the sample, give a plausible range of values for μ .

A **confidence interval** is a type of interval estimators.

A confidence interval takes the form

$$\text{Point Estimator} \pm \text{Margin of Error}$$

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The margin of error is chosen in such a way that it guarantees a certain degree of confidence that the interval contains the true value of the parameter.

Margin of Error

The ME depends on...

- ▶ the estimator's standard error
(ME increases if SE increases)
- ▶ the sample size
(ME decreases if n increases)
- ▶ the desired confidence
(ME increases if we want larger confidence)

Constructing the Margin of Error

CLT implies that the sample mean is approximately $N(\mu, \sigma/\sqrt{n})$ when the sample is large.

$$\blacktriangleright P(-1.96 < Z < 1.96) = 0.95 \quad Z \sim N(0, 1)$$

$$\blacktriangleright P(-1.96 < \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} < 1.96) = 0.95$$

$$\blacktriangleright P(\bar{x} - 1.96(\sigma/\sqrt{n}) < \mu < \bar{x} + 1.96(\sigma/\sqrt{n})) = 0.95$$

▶ We are 95% confident that this interval contains μ :

$$\left(\bar{x} - 1.96 \frac{\sigma}{\sqrt{n}}, \bar{x} + 1.96 \frac{\sigma}{\sqrt{n}} \right)$$

We may want a different confidence level (other than 95%)

For a $(1 - \alpha)100\%$ confidence interval

- ▶ Let $z_{\alpha/2}$ be the point such that $P(Z > z_{\alpha/2}) = \alpha/2$.
- ▶ We are $100(1 - \alpha)\%$ confident that this interval contains μ :

$$\left(\bar{x} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}}, \bar{x} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}} \right)$$

What if σ is Unknown?

If $n \geq 30$, then...

- ▶ s is a fairly good estimate of σ ,
- ▶ we can use s instead of σ if σ is unknown.

Some Useful Z-values

α	$(1 - \alpha)100\%$	$z_{\alpha/2}$
0.1	90%	1.645
0.05	95%	1.96
0.02	98%	2.33
0.01	99%	2.58

CI for the Population Mean - Large Sample

Population: Mean μ , standard deviation σ

Sample: Size $n \geq 30$, sample mean \bar{x} , sample s.d. s .

- ▶ point estimator of μ is \bar{x}
- ▶ $(1 - \alpha)100\%$ confidence interval for μ is $\bar{x} \pm z_{\alpha/2} \frac{s}{\sqrt{n}}$
(If σ is known, use it instead of s)
- ▶ margin of error is $z_{\alpha/2} \frac{s}{\sqrt{n}}$

Interpreting a Confidence Interval

We say:

We are 1 % confident that the true value of the population mean is between 2 and 3 4.

The blanks are:

1. Confidence level $100(1 - \alpha)\%$
2. Lower bound of interval
3. Upper bound of interval
4. Units in which the population mean is measured