

Lecture 12: Sampling Distributions & Standard Errors

Chapter 4.1

Goals for Today

Start Chapter 4: Arguably the most important chapter as it goes to the heart of what statistical inference is. Three important definitions today:

1. point estimate
2. sampling distribution
3. standard error

Point Estimates

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2. Say we get $\bar{x} = 5.025$. If we repeat this procedure: i.e. generate a **new** sample of size $n = 100$ and compute \bar{x}), will we get $\bar{x} = 5.025$?

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We need to characterize this random error.

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1st time We get $\bar{x} = 4.831$

2nd time We get $\bar{x} = 5.104$

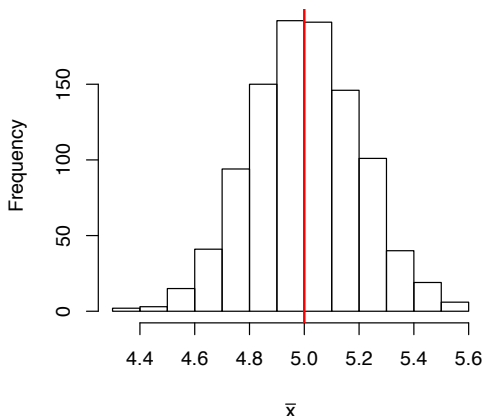
3rd time We get $\bar{x} = 4.965$

...

1000th time We get $\bar{x} = 4.957$

Sampling Distribution

This histogram is the 1000 instances of \bar{x} , where each \bar{x} is based on a sample of $n = 100$. This is the **sampling distribution** of \bar{x} :



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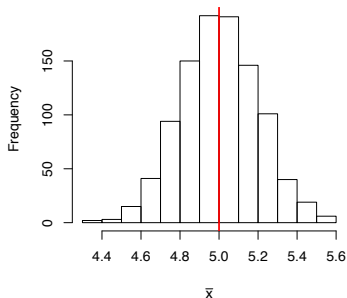
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We will only focus on sample means, including the sample proportion \hat{p} .

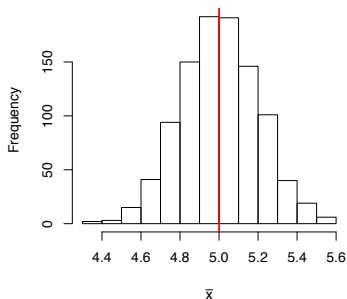
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$$\begin{aligned} & [\mu - 2SD, \mu + 2SD] = [4.6, 5.4] \\ \Rightarrow & \text{length of interval is } 4SD = 5.4 - 4.6 \\ \Rightarrow & SD = 0.2 \end{aligned}$$

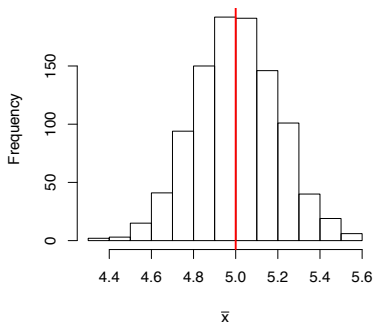
Standard Errors

Standard Error of \bar{x}

Back to Histogram

Samples were of size $n = 100$ with $\sigma = 2$. We estimated that the SD of the sampling distribution was 0.2. Using the formula:

$$SE = \frac{\sigma}{\sqrt{n}} = \frac{2}{\sqrt{100}} = \frac{2}{10} = 0.2$$



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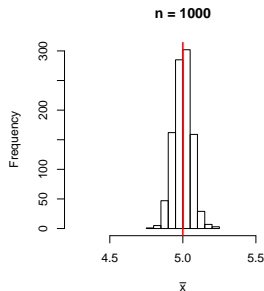
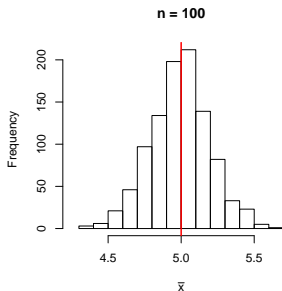
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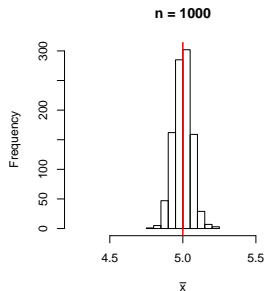
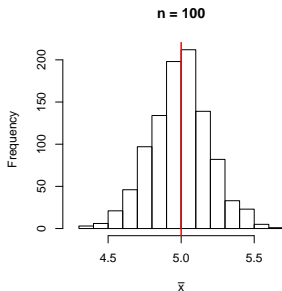
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Both are “accurate”, but the estimates on the right are “more precise.”

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Answer: No, in practice you would **not** sample repeatedly: you do this only **once** for the largest n possible.

Rather the 1000 instances of \bar{x} is a theoretical exercise to illustrate that \bar{x} 's are random and we characterize its randomness by its sampling distribution and its standard error.

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we can use the point estimate of σ . i.e. plug in s in place of σ :

$$SE = \frac{s}{\sqrt{n}}$$

Example

Say in you take a simple random sample of 100 runners in a race and you are interested in their ages:

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Assuming that the 100 runners consist of less than 10% of the population, the standard error of \bar{x} is

$$SE = \frac{s}{\sqrt{100}} = \frac{8.97}{10} = 0.897$$

Population Distribution vs Sampling Distribution

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- ▶ The **sampling distribution** characterizes the (random) behavior of point estimates.
- ▶ The standard deviation of a sampling distribution is the **standard error**: it quantifies the uncertainty/variability of point estimates.

Next Time

- ▶ Confidence Intervals
- ▶ When quoting survey results, what does: “the results of this survey are estimated to be accurate within 3.1 percentage points, 19 times out of 20” mean?
- ▶ Big One: Central Limit Theorem