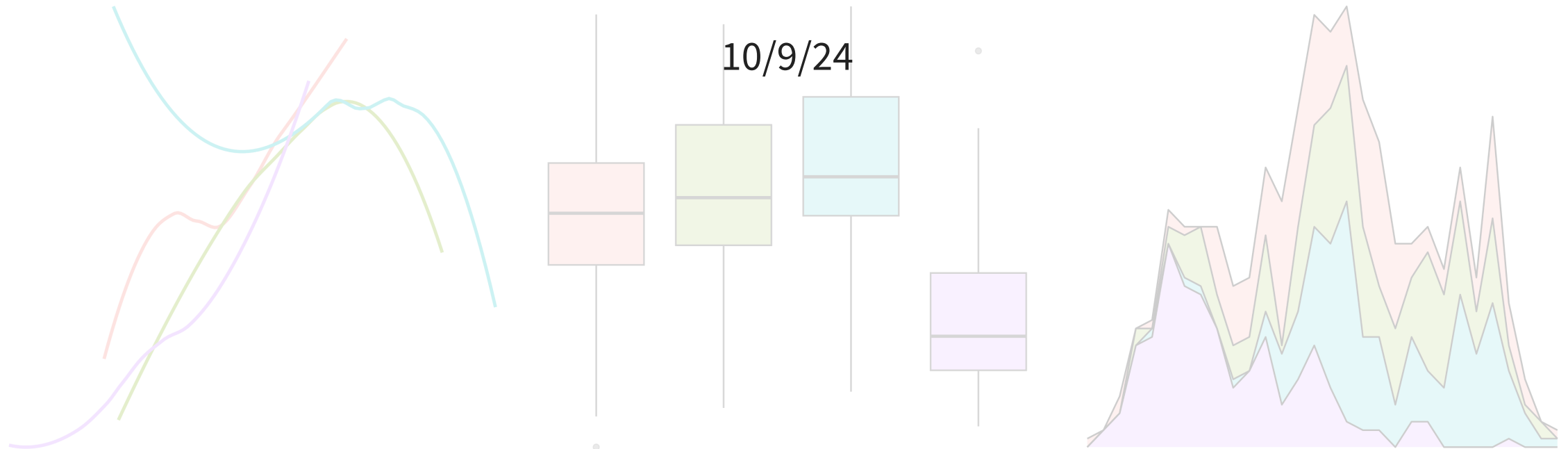
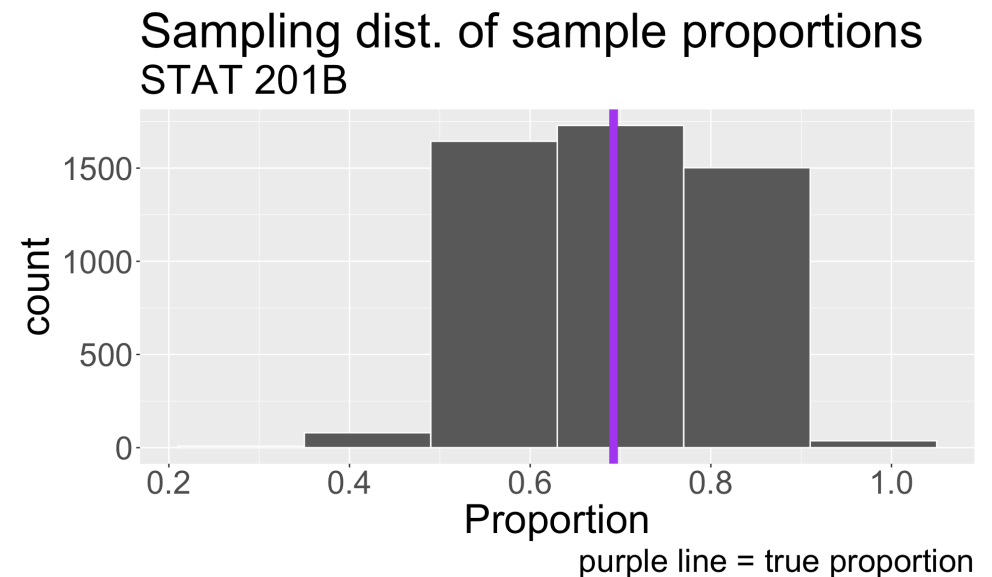
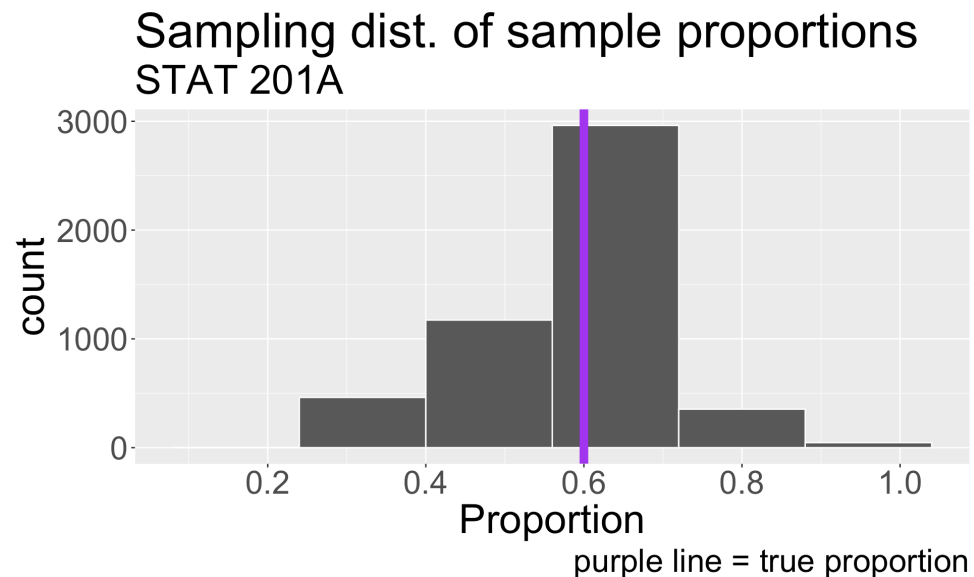


# Bootstrap Confidence Intervals



# Recap

- Sampling distribution describes how statistic behaves under repeated sampling from population
- Let's return to the data collected from our activity.
  - I will repeatedly take SRS of  $n = 10$  values from the population (call this  $\vec{x}$ ) and calculate  $\hat{p}$ . Sampling distribution of  $\hat{p}$  is as follows:



# Bootstrap recap

Taking new samples each time is costly! Bootstrap distribution is an *approximation* of the sampling distribution of the statistic!

Procedure:

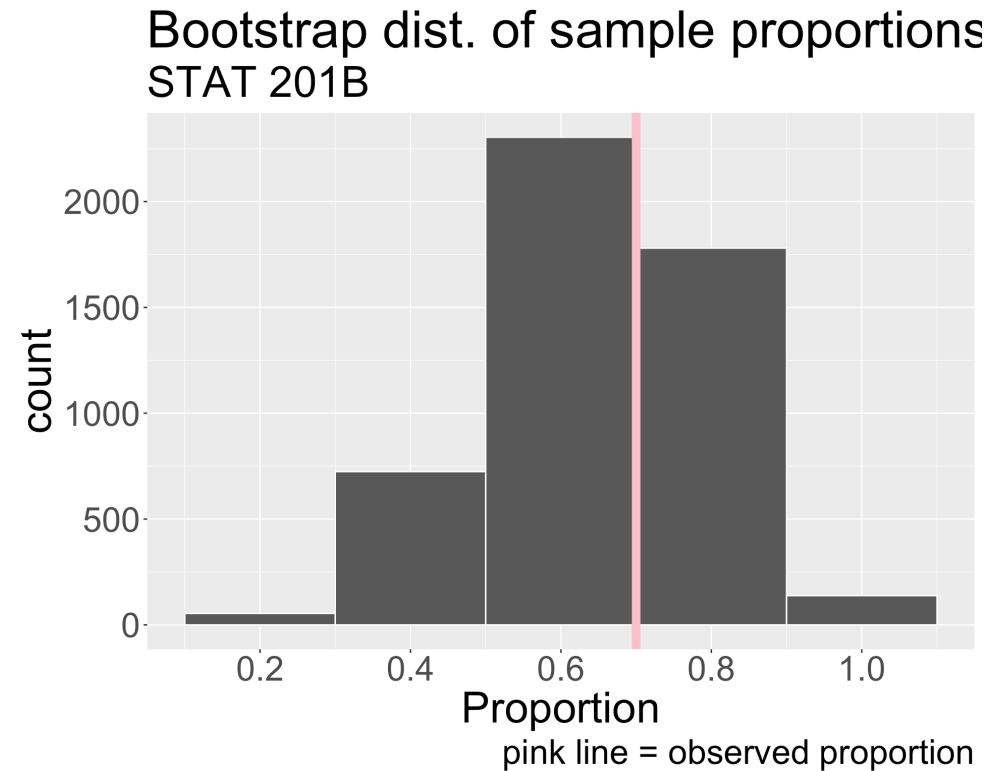
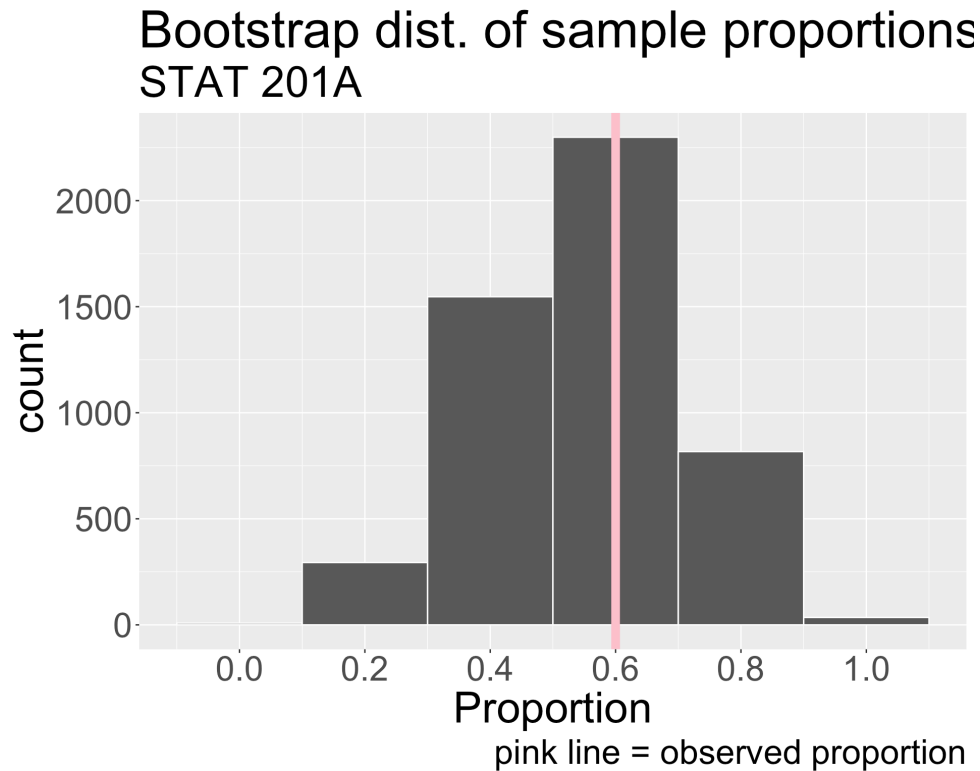
1. Assume we have a sample  $x_1, x_2, \dots, x_n$  from the population. Call this sample  $\vec{x}$ .  
Note the sample size is  $n$
2. Choose a large number  $B$ . For  $b$  in  $1, 2, \dots, B$ :
  - i. Resample: take a sample of size  $n$  with *replacement* from  $\vec{x}$ . Call this set of resampled data  $\vec{x}_b^*$
  - ii. Calculate: calculate and record the statistic of interest from  $\vec{x}_b^*$

At the end of this procedure, we will have a distribution of **resample or bootstrap statistics**

# Bootstrap distribution from activity

We have the following bootstrap distribution of sample proportions, obtained from  $B = 5000$  iterations:

# Bootstrap dist. continued



- Notice where the bootstrap distribution is centered
- What do we do with the bootstrap distribution?

# Answering estimation question

Recall our research question: What proportion of STAT 201A/STAT 201B students get at least 7 hours of sleep a night?

- Could respond using our single point estimate:  $\hat{p}_A = 0.6$  or  $\hat{p}_B = 0.7$
- But due to variability, we recognize that the point estimate will rarely (if ever) equal population parameter
- Rather than report a single number, why not report a range of values?
  - This is possible only if we have a distribution to work with!!

# Confidence intervals

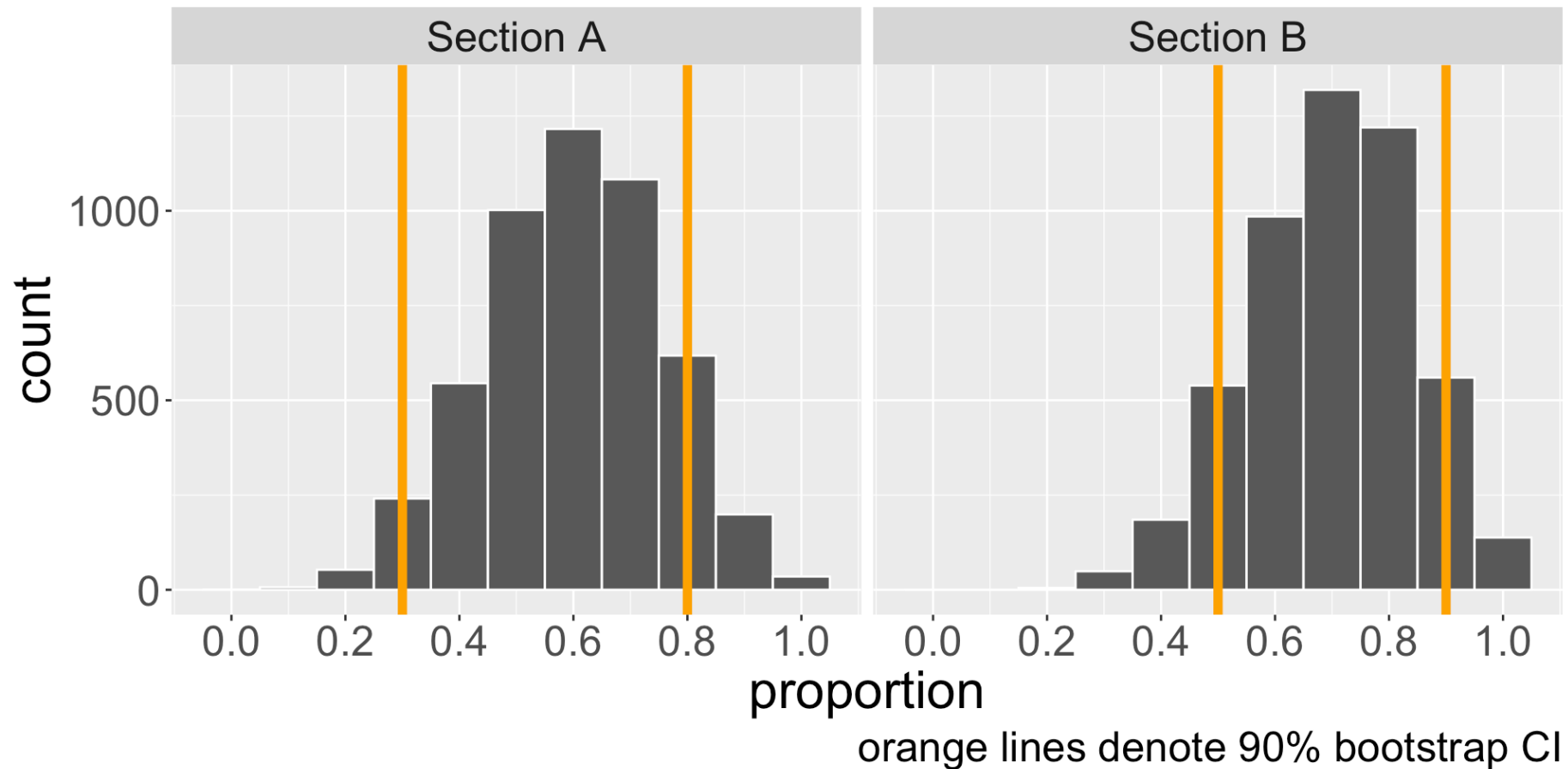
- Analogy: would you rather go fishing with a single pole or a large net?
  - A range of plausible values gives us a better chance at capturing the parameter
- A confidence interval provides such a range of values (more rigorous definition coming soon)
  - “Interval” = we specify a lower bound and an upper bound
  - Confidence intervals are not unique! Depending on the method you use, you might get different intervals

# Bootstrap percentile interval

- The  $\gamma \times 100\%$  **bootstrap percentile interval** is obtained by finding the bounds of the middle  $\gamma \times 100\%$  of the bootstrap distribution
  - e.g. If I want a 90% bootstrap percentile interval, where would the bounds be?
- Called “percentile interval” because the bounds are the  $(1 - \gamma)/2$  and  $(1 + \gamma)/2$  percentiles of the bootstrap distribution
  - e.g. if  $\gamma = 0.80$ , then the bounds would be  $(1 - 0.80)/2 = 0.10$  and  $(1 + 0.80)/2 = 0.90$  percentiles
- For our purposes, “bootstrap confidence interval” will be equivalent to “bootstrap percentile interval”



# Obtaining bootstrap confidence interval



- Section A 90% confidence interval for  $p_A$ : (0.3, 0.8)
- Section B 90% confidence interval for  $p_B$ : (0.5, 0.9)

# Interpreting a confidence interval

- Our 90% confidence interval is: (0.3, 0.8) or (0.5, 0.9). Does this mean there is a 90% chance/probability that the true proportion lies in the interval?
  - **Answer: NO**
- Remember: bootstrap distribution is based on our original sample
  - If we started with a different original sample  $\vec{x}$ , then our estimated 90% confidence interval would also be different
- **What a confidence interval (CI) represents: if we take many independent repeated samples from this population using the same method and calculate a  $\gamma \times 100\%$  CI for the parameter in the exact same way, then in theory,  $\gamma \times 100\%$  of these intervals should capture/contain the parameter**
  - $\gamma$  represents the long-run proportion of CIs that theoretically contain the true parameter
  - However, we never know if any particular interval(s) actually do!

# Interpreting a confidence interval (cont.)

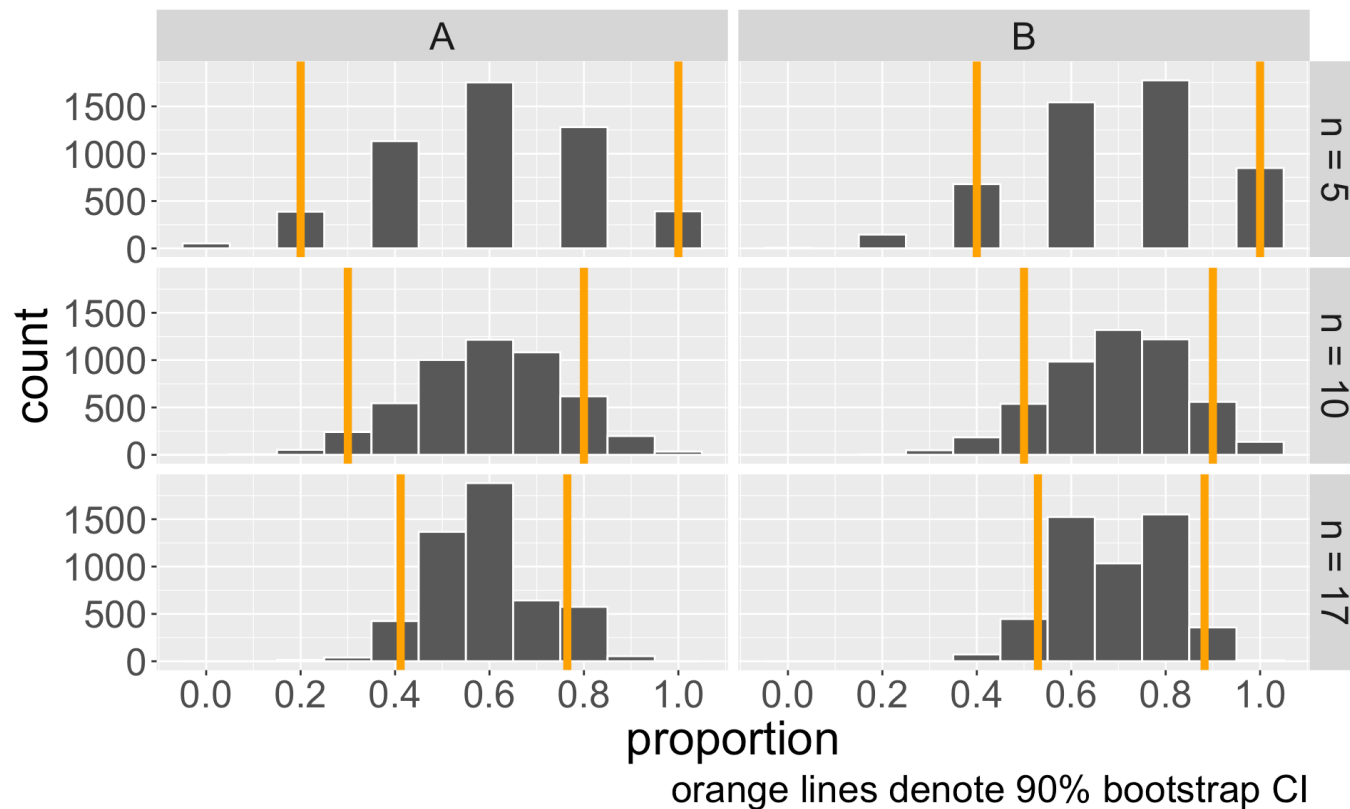
- Correct interpretation (generic): We are  $\gamma \times 100$  % confident that the population parameter is between the lower bound and upper bound.
  - Interpret our bootstrap CI in context
  - Again: why is this interpretation **incorrect**? “There is a 90% chance/probability that the true parameter value lies in the interval.”

# Remarks

- What is a virtue of a “good” confidence interval?
- How do you expect the interval to change as the original sample size  $n$  changes?  
How do you expect the interval to change as level of confidence  $\gamma$  changes?
- Once again, relies on a representative original sample!

# Comparing confidence intervals

Comparing changes in  $\gamma \times 100\%$  CI for sample sizes:  $n = 5$ ,  $n = 10$ , and  $n = 17$ :



## Section A

n	interval
n = 5	(0.2, 1)
n = 10	(0.3, 0.8)
n = 17	(0.41, 0.76)

## Section B

n	interval
n = 5	(0.4, 1)
n = 10	(0.5, 0.9)
n = 17	(0.53, 0.88)

What do you notice?

# Live code + your turn!

- Live code:
  - in-line code
- You will investigate what happens as we move  $\gamma$  between 0 to 1!