Lecture 15 Confidence interval for proportions







Review: Key Points

Chapter 18: Sampling Distribution Models

- A parameter is a number that describes the population.
- A statistic is a number that describes a sample.
- The sampling distribution of a statistic is the distribution of its value in all possible samples of the same size from the same population.
- Central Limit Theorem: The mean of a random sample is a random variable whose sampling distribution can be approximated by a Normal model. The larger the sample, the better the approximation will be.
- This works no matter what the original data's distribution is.
- The sampling distribution of a proportion can be modeled with a Normal model, $N(p,\sqrt{\frac{pq}{n}})$
- The sampling distribution of a mean can be modeled with a Normal model, $N(\mu, \frac{\sigma}{\sqrt{n}})$







Let's revisit what we learned in the last class

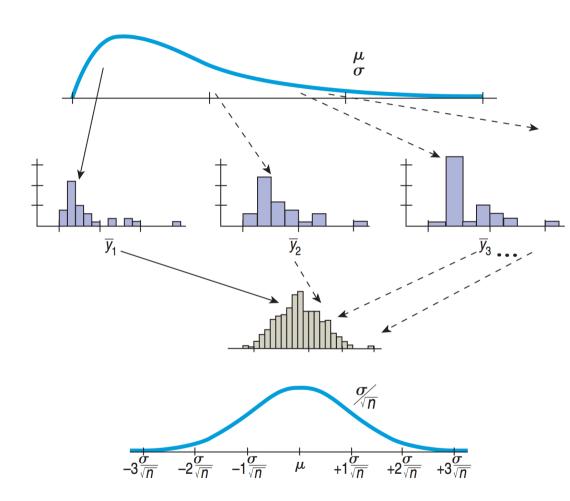
- Population model, that we cannot observe.
- We draw one real sample (solid line) of size n and show its histogram and summary statistics.
- We imagine (or simulate) drawing many other samples (dotted lines).
- We (imagine) gathering all the means into a histogram.
- The CLT tells us we can model the shape of this histogram with a Normal model, $N(\mu, \frac{\sigma}{\sqrt{n}})$







But... can we actually know the mean and SD of the sampling distribution?



- Population model, that we cannot observe.
- We draw one real sample (solid line) of size n and show its histogram and summary statistics.
- VNonaThen, siwhat) should we do? samples (dotted lines).

To be continued...

- We (imagine) gathering all the means into a histogram.
- The CLT tells us we can model the shape of this histogram with a Normal model, $N(\mu, \frac{\sigma}{\sqrt{n}})$







An example data on "F Facebook use":

- Late in 2010, Pew Research surveyed US residents about their use of social networking sites.
- Among 156 respondents aged 18-22, 48 said that they update their status at least daily.
- $\hat{p} = 48/156 = 30.8\%$
- We don't know about p. What can we say about the population, p, with \hat{p} ?
- What we know: $N(p, \sqrt{\frac{pq}{n}})$
 - The sampling distribution model of \hat{p} is centered at p.
 - The standard deviation of the sampling distribution is $\sqrt{\frac{pq}{n}}$

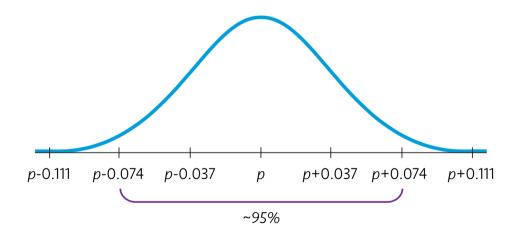






Standard error

- When we estimate the standard deviation of a sampling distribution, we call it a standard error.
- For a sample proportion, \hat{p} , the standard error is $SE(\hat{p}) = \sqrt{\frac{\hat{p}\hat{q}}{n}}$
- For the Facebook users, $SE(\hat{p}) = \sqrt{\frac{\hat{p}\hat{q}}{n}} = \sqrt{\frac{0.308 \times 0.692}{156}} = 0.037 = 3.7\%$

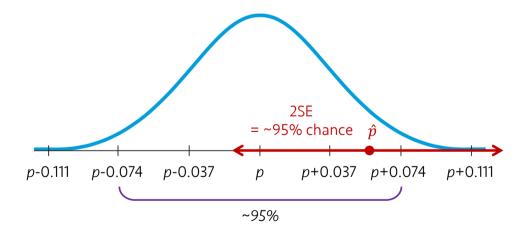








- From \hat{p} 's point of view:
 - "95% chance that p is no more than 2 SEs away from me." (from 68-95-99.7 rule)
 - "I'm 95% sure that *p* will be within my 2SE."
 - "Even if my interval catch p, I still don't know its true value, but the best I can do is an interval."

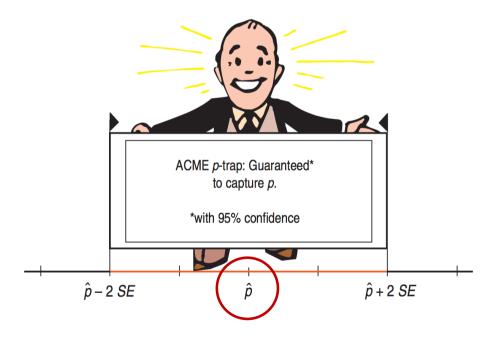








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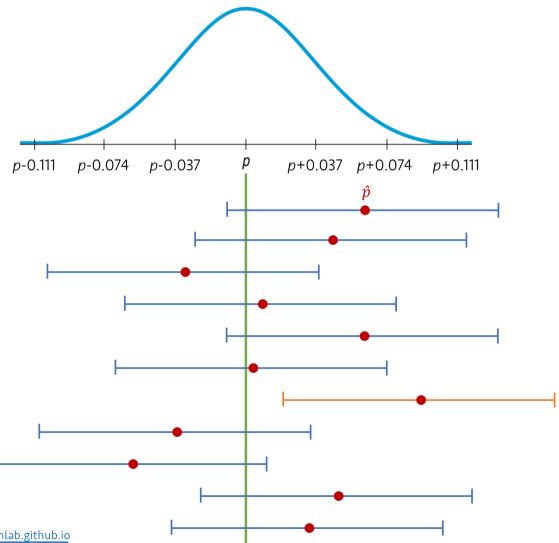


- What can we really say about p for our example data?
 - "30.8% of all Facebook users between the ages of 18 and 22 update their status daily." NO
 - "It is probably truly that 30.8% of *all* Facebook users between the ages of 18 and 22 update their status daily." NO, we don't know true *p*.
 - "We don't know exactly what proportion of Facebook users between the ages of 18 and 22 update their status daily, but we know that it's within the interval 30.8% \pm 2 × 3.7% (23.4% to 38.2%)." We're getting closer... but do we know really?
 - "We don't know exactly what proportion of Facebook users between the ages of 18 and 22 update their status daily, but the interval from 23.4% to 38.2% *probably* contains the true proportion." We're getting closer... but can we more specific?
 - "We are 95% confident that between 23.4% and 38.2% of Facebook users between the ages of 18 and 22 update their status at least daily." OKAY, finally!





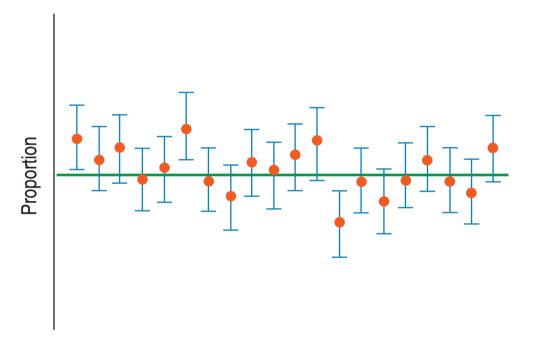


















- Each sample proportion can be used to make a confidence interval.
- The **Central Limit Theorem** assures us that (in the long run) 95% of the intervals cover the true value, and only 5% are wrong.
- The confidence intervals are random because they are based on random samples.
- Our confidence (and our uncertainty) is about the interval, not the true proportion.





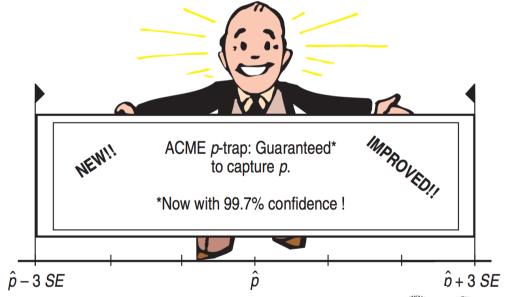


Certainty vs. Precision

- Our confidence interval has this form: $\hat{p} \pm 2SE(\hat{p})$
 - Here, $2SE(\hat{p})$ is called the margin of error (ME).
 - Any population parameter (proportion, mean, regression slope, etc.) can be estimated with some margin of error.

General form: Estimate ± ME

- ME can be
 - 95% confidence interval = 2SE
 - 99.7% confidence interval = 3SE









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General form: Estimate ± ME

- ME can be
 - 95% confidence interval = 2SE
 - 99.7% confidence interval = 3SE
 - The more confident we want to be, the larger the margin of error must be.
 - E.g., 100% confidence: the proportion of Facebook users who update daily is between 0 and 100%
 - Is this useful? NO
 - Or we can give a very narrow interval (e.g., 30.7%-30.9%) with very low confidence.
 - is this useful? Maybe not.
- Every confidence interval is a balance between certainty and precision.
 - 90%, 95%, 99% are commonly used.

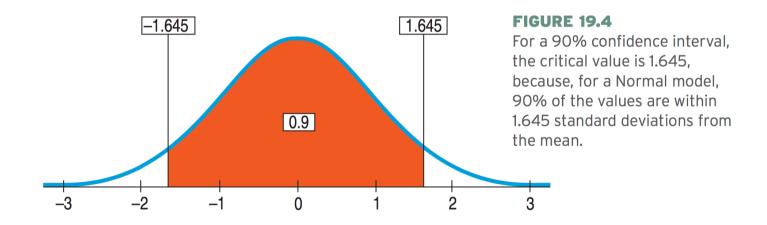






Critical values

- Critical value = the *number* of SEs (e.g., 2 in 2SEs)
- Denoted as z*
- For 95% confidence interval, the precise critical value is $z^* = 1.96$ (though we used 2 to make it simple).









Assumptions and Conditions

Assumptions and Conditions

- Independence Assumption: The individuals in the samples must be independent of each other.
- We can't know if this assumption is true or not for sure, but we can check the following conditions that provide information about the assumption.
- Conditions:
 - Randomization Condition: random assignment (experiment), random sampling (survey)
 - 10% Condition: The sample size, *n*, must be no larger than 10% of the population. If you sample more than about 10% of the population, the remaining individuals are no longer truly independent of each other.
 - Success/Failure Condition: The sample size has to be big enough so that we expect at least 10 successes and at least 10 failures (*np* and *nq* should be > 10)















Choosing your sample size

•
$$ME = z^* \sqrt{\frac{\hat{p}\hat{q}}{n}}$$

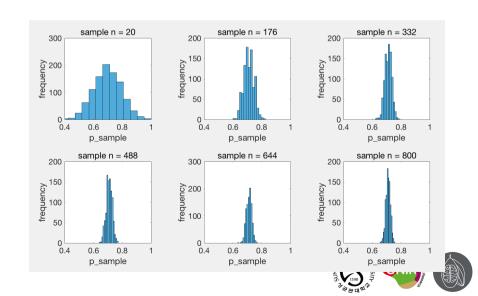
• If we want 3% Margin of error (ME) for 95% confidence ($z^* = 1.96$)

•
$$0.03 = 1.96 \times \sqrt{\frac{\hat{p}\hat{q}}{n}}$$

• If you don't know \hat{p} , just take its maximum: $\hat{p}=0.5$, then $\hat{p}\hat{q}=0.25$

•
$$\sqrt{n} = \frac{1.96\sqrt{0.5 \times 0.5}}{0.03} \approx 32.67$$

- $n \approx (32.67)^2 \approx 1067.1$
- Then, our sample size should be 1068 (rounding up).



Key Points

Chapter 19: Confidence Interval for Proportions

- Standard error: standard deviation of a sampling distribution
- For a sample proportion, \hat{p} , the standard error is $SE(\hat{p}) = \sqrt{\frac{\hat{p}\hat{q}}{n}}$
- 95% confidence interval for a proportion, $\hat{p} \pm 1.96 SE(\hat{p})$
- General form: Estimate ± Margin of Error (ME)
- Critical value, z^* = the *number* of SEs (e.g., 2 in 2SEs)
- Every confidence interval is a balance between certainty and precision.
- You can choose your sample size based on confidence interval.





