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Online Review Course of Undergraduate Probability and Statistics

Review Lecture 11

Confidence Intervals

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Course Website: www.lithoguru.com/scientist/statistics/review.html

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Sampling Distribution of the Mean

- Let X_1, X_2, \dots, X_n be iid random variables from an infinite population with mean μ and finite variance σ^2 .

$$\bar{X} = \frac{1}{n} \sum_{i=1, n} X_i \quad E(\bar{X}) = \mu \quad \sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$$

- Central Limit Theorem:

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0,1) \text{ as } n \rightarrow \infty$$

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Sampling Distribution of the Mean

- We say that \bar{X} is an **estimator** for the population mean, and that \bar{x} is a **point estimate** of the population mean
 - We want our estimator to be unbiased (or low bias)
 - We want our estimator to have a small variance (standard error)
 - An efficient estimator is one that has lower bias and/or smaller variance than another estimator
- Our estimator \bar{X} is unbiased provided our sample is random (each X_i is independent), and it is the most efficient
 - For a symmetric distribution, the median is an unbiased estimator, but is less efficient than the sample mean

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Errors in our Estimate of the Mean

- How large might $\bar{X} - \mu$ get?
 - We know by the central limit theorem that for a moderately large sample $Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim N(0,1)$

α	$Z_{\alpha/2}$
0.10	1.65
0.05	1.96
0.02	2.33
0.01	2.58

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Creating a Confidence Interval

- Let's look at the two boundary points, $\pm z_{\alpha/2}$

$$\pm z_{\alpha/2} = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} \rightarrow \mu = \bar{x} \pm z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

- The probability that the true mean will fall within this range is $1 - \alpha$
- Generally, we approximate σ with s

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Interval Estimate of the Mean

- Confidence interval (CI):

$$\bar{x} - z_{\alpha/2} \frac{s}{\sqrt{n}} < \mu < \bar{x} + z_{\alpha/2} \frac{s}{\sqrt{n}}$$

Large Sample Confidence Interval

Margin of Error

$Z_{\alpha/2}$	α
1.65	0.10
1.96	0.05
2.33	0.02
2.58	0.01

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Confidence Intervals

- Pick the value of α you want
- Do you have a large enough sample so that the sampling interval can be considered normal?
 - If yes, use the $z_{\alpha/2}$ value
 - If no, but underlying distribution is normal, use $t_{\alpha/2}$ value
- Create the confidence interval
 - We are $(1-\alpha)100\%$ confident that the true population mean is captured by our interval
 - If we ran this experiment 100 times, we expect that our confidence intervals would capture the true mean $(1-\alpha)100$ of those times
- Every statistic has a confidence interval (not just the mean)

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Another Statistic: Proportion

- Proportion examples:
 - What percentage of UT students smoke?
 - How many people prefer brand X over brand Y?
 - What fraction of the molecules have reacted?
- With two options, the population will follow a binomial distribution
 - p = population proportion (probability) of “success”
 - $q = 1 - p$ = proportion (probability) of “failure”

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Another Statistic: Proportion

- Proportion estimator
 - Sample size = n , X_i = Bernoulli RV
 - $X = \sum_{i=1,n} X_i$ = binomial RV Recall from Lecture 8

$$\hat{p} = \frac{X}{n} \quad E(\hat{p}) = \frac{E(X)}{n} = \frac{np}{n} = p$$

$$var(\hat{p}) = \frac{var(X)}{n^2} = \frac{np(1-p)}{n^2} = \frac{p(1-p)}{n} = \frac{pq}{n}$$

Standard Error $SE(\hat{p}) = \sqrt{\frac{\hat{p}\hat{q}}{n}}$

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Proportion Confidence Intervals

- The binomial distribution can be well approximated with a normal distribution whenever $np > 10$ and $nq > 10$
- When this is the case, use $z_{\alpha/2}$
 - Margin of error = $z_{\alpha/2}SE(\hat{p})$
- Example: 95% CI ($\alpha = 0.05$)
 - 95% CI = $\left(\hat{p} - 1.96\sqrt{\frac{\hat{p}\hat{q}}{n}}, \hat{p} + 1.96\sqrt{\frac{\hat{p}\hat{q}}{n}} \right)$

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Review #11: What have we learned?

- Explain estimator, point estimate, and interval estimate
- Under what conditions can we use the normal approximation for the sampling distribution of the mean?
- Know how to generate a confidence interval for any statistic
- What is the standard error for a proportion estimate?
- Under what conditions can we use the normal approximation for the sampling distribution of the proportion?

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