Chapter 8: Inference for Proportions

MATH 560-01 Statistical Data Analysis

March 8, 2021

These slides are based on material from *Introduction to the Practice of Statistics* by David S. Moore, George P. McCabe, and Bruce A. Craig, 9th edition.

Sections

8.1 Inference for a Single Proportion

8.2 Comparing Two Proportions

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- ▶ determine whether the large-sample confidence interval and large-sample significance test is appropriate based on the guidelines described in this section

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- We select a random sample of size n from the population and count the number of "successes" X in the sample.
- ▶ Then, we use the sample proportion $\hat{p} = \frac{X}{n}$ to estimate p.
- ▶ If the sample is taken with replacement, then X follows a B(n, p) distribution.
- ▶ If we take a SRS and the population size is much larger than n (at least 20 times as large), then X approximately follows a B(n, p) distribution.

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- ▶ However, when n is large, \hat{p} is approximately $N\left(p, \sqrt{\frac{p(1-p)}{n}}\right)$ (see Chapter 5) so we can use tests based on the Normal distribution.
- ▶ In this chapter, we only consider inference procedures for large sample sizes based on the Normal distribution.

Approximate large-sample confidence interval for p when n is large

ightharpoonup A level C confidence interval for p is

$$\hat{p} \pm z^* \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$$

where z^* is the value for the N(0,1) density with area C between $-z^*$ and z^* and $\hat{p} = \frac{X}{n}$.

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▶ To use this approximation, the number of successes and the number of failures observed in the sample should be at least 10.

- ▶ Example 8A: The South African mathematician John Kerrich, while a prisoner of war in World War II, tossed a coin 10,000 times and obtained 5,067 heads.
 - (a) Find a 95% confidence interval for the probability that the Kerrich's coin comes up heads.
 - (b) Are the guidelines for when to use the large-sample confidence interval for a population proportion satisfied in this setting?

► Answer:

Choosing the sample size

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- So, we need n to be at least $\left(\frac{z^*}{m}\right)^2 p^*(1-p^*)$.
- A conservative choice for p^* is 0.5 since $p^*(1-p^*)$ is maximized when $p^* = 0.5$.
- So choosing n to be at least $\frac{1}{4} \left(\frac{z^*}{m} \right)^2$ will work regardless of the true p.

▶ Example 8B: Suppose we plan to compute a 99% confidence interval for the probability that a coin comes up heads and want the margin of error to be no more than 0.01. How many times do we need to toss the coin to guarantee this margin of error?

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 $\frac{\text{Two-sided test for a population proportion}}{\text{when } n \text{ is large}}$

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Two-sided test for a population proportion when n is large

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- 4. Reject H_0 if the P-value $\leq \alpha$

The book's guidelines recommend that this large-sample test should only be used when the expected number of successes np_0 and the expected number of failures $n(1-p_0)$ are both greater than 10.

One-sided (left-sided) test for a population proportion when n is large

- 1. Test $H_0: p = p_0$ vs. $H_a: p < p_0$
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- **Example 8C**: The South African mathematician John Kerrich, while a prisoner of war in World War II, tossed a coin 10,000 times and obtained 5,067 heads.
 - (a) Is there significant evidence at the 5% level that the probability that Kerrich's coin comes up heads is not 0.5?
 - (b) Are the guidelines for when to use the large-sample significance test for a population proportion satisfied in this setting?

► Answer:

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- ▶ Here is notation used for this section:

	Population	Sample	Count of	Sample
Population	proportion	size	successes	proportion
1	p_1	n_1	X_1	\hat{p}_1
2	p_2	n_2	X_2	\hat{p}_2

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$$\begin{array}{l} \mu_{\hat{p}_1 - \hat{p}_2} = \mu_{\hat{p}_1} - \mu_{\hat{p}_2} = p_1 - p_2 \\ \sigma_{\hat{p}_1 - \hat{p}_2}^2 = \sigma_{\hat{p}_1}^2 + \sigma_{\hat{p}_2}^2 = \frac{p_1(1 - p_1)}{n_1} + \frac{p_2(1 - p_2)}{n_2} \end{array}$$

Approximate confidence interval for $p_1 - p_2$ when n_1 and n_2 are large

▶ A level C confidence interval for $p_1 - p_2$ is

$$D \pm z^* \sqrt{\frac{\hat{p}_1(1-\hat{p}_1)}{n_1} + \frac{\hat{p}_2(1-\hat{p}_2)}{n_2}}$$

where z^* is the value for the N(0,1) density with area C between $-z^*$ and z^* and $D = \hat{p}_1 - \hat{p}_2$.

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➤ To use this approximation, the number of successes and the number of failures observed in each sample should be at least 10.

Example 8D: A study examined the proportion of high school students who cheated on tests at least twice in 2002 and 2004. A reported 9054 out of 24142 students said they cheated at least twice in 2004. A reported 5794 out of 12121 said they cheated at least twice in 2002. Compute a 90\% confidence interval for $p_{2004} - p_{2002}$ where p_{2004} is the proportion of students from the 2004 population who cheated at least twice and p_{2002} is the proportion of students from the 2002 population who cheated at least twice.

► Answer:

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Significance Test for Difference in Proportions

- ► The standard error calculation is slightly different in the setting of a significance test with a null hypothesis $H_0: p_1 = p_2.$
- ▶ If H_0 is true, then both population have the same proportion (say p) and we use a pooled estimate $\hat{p} = \frac{X_1 + X_2}{n_1 + n_2}$ in the calculation of the standard error; that is, when \bar{H}_0 is true, $\sigma_D = \sqrt{p(1-p)\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}$ so $SE_{D_p} = \sqrt{\hat{p}(1-\hat{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}.$

Two-sided test for equality of population proportions when n_1 and n_2 are large

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2. Test statistic:
$$z = \frac{\hat{p}_1 - \hat{p}_2}{\sqrt{\hat{p}(1-\hat{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$$
 where $\hat{p}_1 = \frac{X_1}{n_1}, \ \hat{p}_2 = \frac{X_2}{n_2}, \ \text{and} \ \hat{p} = \frac{X_1 + X_2}{n_1 + n_2}.$

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- 4. Reject H_0 if the P-value $\leq \alpha$

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One-sided (left-sided) test for equality of population proportions when n_1 and n_2 are large

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- 2. Test statistic: $z = \frac{\hat{p}_1 \hat{p}_2}{\sqrt{\hat{p}(1-\hat{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}}$ where $\hat{p}_1 = \frac{X_1}{n_1}, \, \hat{p}_2 = \frac{X_2}{n_2}, \, \text{and } \hat{p} = \frac{X_1 + X_2}{n_1 + n_2}.$
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Example 8E: An experiment is performed to compare the effectiveness of a new drug to a current drug typically used to treat headaches. A randomized controlled double-blind design is used in which 100 subjects who have headaches receive the new drug and 100 subjects who have headaches receive the standard drug. Among the subjects who receive the new drug, 75 experience relief within 2 hours. Among the subjects who receive the standard drug, 60 experience relief within 2 hours. Is this sufficient evidence to conclude that the new drug is more effective in quickly providing some relief to subjects with headaches? Perform an appropriate significance test at level $\alpha = .01$.

► Answer: