# Statistics 305/605: Introduction to Biostatistical Methods for Health Sciences

Chapter 14: Inference for Proportions

Jinko Graham

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# Inference for Proportions (Chapter 14)

- ▶ Instead of quantitative measurements, we classify each sampled individual into one of two categories.
  - success, failure (canonical)
  - breast cancer, no breast cancer
  - Etc.
- Mant to make inference about the proportion p of successes in a population, or about the difference between the proportions  $p_1$  and  $p_2$  of successes in two populations.

# Example: Women's Health Initiative (WHI)

- A randomized controlled trial, called the Women's Health Initiative, randomized 16,608 post-menopausal women aged 50-79 years to receive either hormone replacement therapy in the form of estrogen plus progestin (EP;  $n_1 = 8506$ ), or a placebo ( $n_2 = 8102$ ).
- ▶ After five years, 166 of those in the EP group had developed invasive breast cancer, compared to 122 in the placebo group.
- ► The populations to compare are postmenopausal women aged 50-79 years, who are taking EP (population 1) or placebo (population 2).
- ▶ The sample proportions are  $\hat{p}_1 = 166/8506 = 0.0195$  and  $\hat{p}_2 = 122/8102 = 0.0151$
- ▶ It looks like the EP group has a higher risk of breast cancer, but could this difference be due to chance?

## Outline of Approach

- Similar approach to inference of population means (quantitative outcomes), with some minor differences.
- ▶ Inference is based on the sampling distribution of  $\hat{p}_1 \hat{p}_2$
- ► Two-stage reasoning:
  - 1. Transform the statistic  $\hat{p}_1 \hat{p}_2$  into an initial pivotal quantity,  $Z_1$ , whose denominator depends on the unknown parameters  $p_1$  and  $p_2$ .
  - 2. Get a final pivotal quantity, Z, by replacing the unknown  $p_i$ 's in the denominator of  $Z_1$  with estimates.
- ► Cls and hypothesis tests follow from the approximate sampling distribution of the final pivotal quantity, Z.
- Note: Not covering Inference for a single proportion (text, sections 14.2 − 14.5).

## Sampling Distribution of $\hat{p}_1 - \hat{p}_2$

- Assume that we have independent simple random samples (SRSs) of size  $n_1$  and  $n_2$ , from the two parent populations.
- ▶ Then the distribution of  $\hat{p}_1 \hat{p}_2$  has
  - ▶ mean  $p_1 p_2$  and
  - ► SD

$$\sqrt{p_1(1-p_1)/n_1+p_2(1-p_2)/n_2}$$
.

- ▶ If the sample sizes are "large", the CLT tells us that the shape of this distribution is approximately normal.
  - ► (The CLT applies because it turns out that a proportion is an average . . . of 0's and 1's.)

#### Initial Transformation

- For large samples,  $\hat{p}_1 \hat{p}_2$  is **approximately** normally distributed with
  - ▶ mean  $p_1 p_2$ , and
  - ► SD

$$\sqrt{p_1(1-p_1)/n_1+p_2(1-p_2)/n_2}$$

So the distribution of

$$Z_1 = rac{(\hat{p}_1 - \hat{p}_2) - (p_1 - p_2)}{\sqrt{p_1(1 - p_1)/n_1 + p_2(1 - p_2)/n_2}}$$

is approximately N(0,1).

#### $Z_1$ with Estimated SDs

▶ Inserting the estimates  $\hat{p}_1$  and  $\hat{p}_2$  for the unknown parameters  $p_1$  and  $p_2$  into  $Z_1$  above gives

$$Z = \frac{(\hat{p}_1 - \hat{p}_2) - (p_1 - p_2)}{\sqrt{\hat{p}_1(1 - \hat{p}_1)/n_1 + \hat{p}_2(1 - \hat{p}_2)/n_2}}$$

- ▶ What is the distribution of *Z*?
  - ▶ Under certain conditions (see below) it is reasonable to assume that *Z* is approximately normal.

## Rule of Thumb for Normal Approximation

- ► The normal approximation to the distribution of Z is considered reliable when the sample sizes n<sub>1</sub> and n<sub>2</sub> are "large"
- ▶ The definition of large depends on the underlying  $p_j$ 's. The text (page 324) suggests:
  - ▶  $n_1p_1 \ge 5$  and  $n_1(1-p_1) \ge 5$  and
  - ▶  $n_2p_2 \ge 5$  and  $n_2(1-p_2) \ge 5$
- ▶ The parameters  $p_1$  and  $p_2$  are not known so we insert the estimates  $\hat{p}_1 = (\text{number of successes in sample } 1)/n_1$  and  $\hat{p}_2 = (\text{number of successes in sample } 2)/n_2$ .
- ► After inserting estimates, one can simplify the requirements to the following rule:
  - ► The normal approximation is reliable when there are at least 5 successes and 5 failures in both sample 1 and sample 2.

## Checking rule of thumb for WHI Data

- At least 5 successes and failures in both samples.
  - ► True: 166 cancer, 8340 cancer-free in the EP group; 122 cancer, 7980 cancer-free in the placebo group.

#### Confidence Intervals

▶ The level-*C* CI for  $p_1 - p_2$  is of the form:

estimate  $\pm$  margin of error

- ▶ The estimate is  $\hat{p}_1 \hat{p}_2$
- ▶ The margin of error is  $z^* \times SE$  where
  - $z^*$  is the upper (1-C)/2 critical value of the standard normal distribution.
  - ▶ SE is the estimated SD of  $\hat{p}_1 \hat{p}_2$  in the denominator of Z; namely,  $SE = \sqrt{\hat{p}_1(1-\hat{p}_1)/n_1 + \hat{p}_2(1-\hat{p}_2)/n_2}$

## Example (WHI)

- ▶ Recall: 16,608 women aged 50-79 years randomized to receive either estrogen plus progestin (EP;  $n_1 = 8506$ ), or a placebo ( $n_2 = 8102$ ). After five years, 166 in the EP group developed invasive breast cancer, compared to 122 in placebo group.
- ► For EP,  $\hat{p}_1 = 166/8506$  and, for placebo,  $\hat{p}_2 = 122/8102$ .
- ▶ 95% CI is estimate  $\pm$  margin of error, where
  - estimate of  $p_1 p_2$  is  $\hat{p}_1 \hat{p}_2 = 0.0044$
  - margin of error is a critical value times standard error of difference.
- ▶ The critical value is 1.96 (see R demo).
- ► The standard error is  $\sqrt{\hat{p}_1(1-\hat{p}_1)/n_1 + \hat{p}_2(1-\hat{p}_2)/n_2} = 0.002$
- ▶ The margin of error is therefore 1.96 \* .002 = 0.00392.
- ▶ Putting it all together, the CI is  $0.0044 \pm 0.00392$  or approximately (0.0005, 0.008).

#### Test Statistic

- ▶ The null hypothesis is  $H_0: p_1 p_2 = 0$ .
- Numerator of the test statistic is therefore the estimated difference  $(\hat{p}_1 \hat{p}_2)$  minus 0.
- Denominator of the test statistic is

$$SE = \sqrt{\hat{p}_1(1-\hat{p}_1)/n_1 + \hat{p}_2(1-\hat{p}_2)/n_2}$$

- ▶ BUT, under  $H_0$ , we have  $p_1 = p_2$ . Call this common value p.
- Assuming a common proportion p in the two populations, we pool the 2 samples to obtain an estimate  $\hat{p}$ ; i.e.,

$$\hat{p} = (\text{number of cancers in both samples})/(n_1 + n_2).$$

- ► The formula for the SE of  $\hat{p}_1 \hat{p}_2$  simplifies to  $\sqrt{\hat{p}(1-\hat{p}) \times (1/n_1 + 1/n_2)}$
- ▶ So the statistic for testing  $H_0: p_1 p_2 = 0$  is

$$Z = \frac{(\hat{p}_1 - \hat{p}_2)}{\sqrt{\hat{p}(1-\hat{p})(1/n_1 + 1/n_2)}}$$

## Example (WHI, continued)

- $\vdash$   $H_0: p_1-p_2=0.$
- ► Test statistic numerator is  $\hat{p}_1 \hat{p}_2 = 0.0044$ .
- ► Test statistic denominator is the SE based on the pooled estimate of  $p = p_1 = p_2$ .
  - Pooled estimate of the common population proportion is  $\hat{p} = (166 + 122)/(8506 + 8102) = 0.0173$ .
  - ▶ So the SE is

$$\sqrt{\hat{p}(1-\hat{p})(1/n_1+1/n_2)} 
= \sqrt{0.0173(1-0.0173)(1/8506+1/8102)} 
= 0.002.$$

▶ The test statistic value is z = 0.0044/0.002 = 2.2.

#### *p*-value

- ▶ The *p*-value is the chance of a value of the test statistic that is as or more extreme than what we did observe in our data, when the null hypothesis is true.
- ► Same logic as we saw earlier for inference of population means (Chapter 11).
- ▶ Let Z be a standard normal random variable and z be the observed value of the test statistic.
- ▶ For  $H_a: p_1 p_2 \neq 0$ , the *p*-value is  $p = 2P(Z \geq |z|)$ .

## Example (WHI, continued)

- Suppose we wish to test  $H_0$ :  $p_1 p_2 = 0$  vs.  $H_a$ :  $p_1 p_2 \neq 0$  at level  $\alpha = 0.05$ .
- For an observed value of the test statistic z = 2.2, computer software calculates a pvalue of about 0.03 (see R demo)
- ▶ We therefore reject  $H_0$  at the 5% level: There is statistical evidence that women taking EP have a higher risk of invasive breast cancer than those taking the placebo.
  - Note: We say that the EP group has higher risk than the placebo group because  $\hat{p}_1$  (for EP) is greater than  $\hat{p}_2$  (for placebo), as evidenced by z = 2.2 being greater than zero.

#### Summary

- ▶ Inference for the difference  $p_1 p_2$  between two population proportions is based on a pivotal quantity.
- Confidence intervals are of the form estimate ± margin of error, where
  - estimate is difference between sample means, and
  - margin of error is a critical value times the standard error of the difference in sample proportions
- ▶ To test the null hypothesis  $H_0: p_1 p_2 = 0$  against an alternative  $H_a$  we calculate a test statistic and the p-value
  - ► The p-value is the chance of seeing a value of the test statistic as or more extreme than the value that was observed in our data, under the null hypothesis.
  - ▶ Compare the p-value to a significance level  $\alpha$  to obtain a statistical hypothesis test
- ► The statistical inference is considered reliable when there are at least 5 successes and 5 failures in each sample.