

# Hypothesis Testing: Categorical Data

## Ex. 10.4

- A hypothesis: an important factor for breast cancer is age at first birth.
- An international study was set up to test the hypothesis.
  - Breast cancer cases were identified among women in selected hospitals in the United States, Greece, Yugoslavia, Brazil, and Japan.
  - Controls were chosen from women of comparable age who were in the hospital at the same time as the cases, but who did not have breast cancer.
  - All women were asked about their age at first birth.
  - The set of women with at least one birth was arbitrarily divided into two categories:
    - Women whose age at first birth  $\leq 29$
    - Women whose age at first birth  $\geq 30$
- Results among women with at least one birth
  - 683 out of 3220 (21.2%) women with breast cancer had an age at first birth  $\geq 30$
  - 1498 out of 10,245 (14.6%) women without breast cancer had an age at first birth  $\geq 30$
- How can we assess whether this difference is significant?

# Two-Sample Test for Binomial Proportions

- $p_1$  = the probability that age at first birth is  $\geq 30$  in case women.
- $p_2$  = the probability that age at first birth is  $\geq 30$  in control women.
- Whether or not the underlying probability of having an age at first birth of  $\geq 30$  is different in the two groups.
- $H_0: p_1 = p_2 = p$  versus  $H_1: p_1 \neq p_2$

# Normal-Theory Method

- Base the significance test on the difference between the sample proportions  $\hat{p}_1 - \hat{p}_2$
- Assume samples are large enough

$\hat{p}_1 - \hat{p}_2$  is normally distributed

$$\frac{pq}{n_1} + \frac{pq}{n_2} = pq \left( \frac{1}{n_1} + \frac{1}{n_2} \right) \quad z = (\hat{p}_1 - \hat{p}_2) / \sqrt{pq(1/n_1 + 1/n_2)} \approx N(0,1)$$

$$\hat{p} = \frac{n_1 \hat{p}_1 + n_2 \hat{p}_2}{n_1 + n_2} = \frac{x_1 + x_2}{n_1 + n_2}$$

To better accommodate the normal approximation to the binomial

$$|\hat{p}_1 - \hat{p}_2| - \left( \frac{1}{2n_1} + \frac{1}{2n_2} \right)$$

**Two-Sample Test for Binomial Proportions (Normal-Theory Test)** To test the hypothesis  $H_0: p_1 = p_2$  versus  $H_1: p_1 \neq p_2$ , where the proportions are obtained from two independent samples, use the following procedure:

- (1) Compute the test statistic

$$z = \frac{|\hat{p}_1 - \hat{p}_2| - \left( \frac{1}{2n_1} + \frac{1}{2n_2} \right)}{\sqrt{\hat{p}\hat{q}\left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

where  $\hat{p} = \frac{n_1\hat{p}_1 + n_2\hat{p}_2}{n_1 + n_2} = \frac{x_1 + x_2}{n_1 + n_2}$ ,  $\hat{q} = 1 - \hat{p}$

and  $x_1, x_2$  are the number of events in the first and second samples, respectively.

- (2) For a two-sided level  $\alpha$  test,

if  $z > z_{1-\alpha/2}$

then reject  $H_0$ ;

if  $z \leq z_{1-\alpha/2}$

then accept  $H_0$ .

- (3) The approximate  $p$ -value for this test is given by

$$p = 2[1 - \Phi(z)]$$

- (4) Use this test only when the normal approximation to the binomial distribution is valid for each of the two samples—that is, when  $n_1\hat{p}\hat{q} \geq 5$  and  $n_2\hat{p}\hat{q} \geq 5$ .

$$z = \frac{|\hat{p}_1 - \hat{p}_2| - \left( \frac{1}{2n_1} + \frac{1}{2n_2} \right)}{\sqrt{\hat{p}\hat{q}\left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

$$\text{where } \hat{p} = \frac{n_1\hat{p}_1 + n_2\hat{p}_2}{n_1 + n_2} = \frac{x_1 + x_2}{n_1 + n_2}, \hat{q} = 1 - \hat{p}$$

Sample proportion of case women whose age at first birth was  $\geq 30$  is

$$\hat{p}_1 = 683/3220 = .212$$

For control women

$$\hat{p}_2 = 1498/10,245 = .146$$

$$\hat{p} = (683 + 1498)/(3220 + 10,245) = .162$$

$$\hat{q} = 1 - .162 = .838$$

$$n_1\hat{p}\hat{q} = 3220(.162)(.838) = 437 \geq 5$$

$$n_2\hat{p}\hat{q} = 10,245(.162)(.838) = 1391 \geq 5$$

The test statistic is given by

$$\begin{aligned} z &= \left\{ |.212 - .146| - \left[ \frac{1}{2(3220)} + \frac{1}{2(10,245)} \right] \right\} / \sqrt{.162(.838)\left( \frac{1}{3220} + \frac{1}{10,245} \right)} \\ &= .0657/.00744 \\ &= 8.8 \end{aligned}$$

The  $p$ -value =  $2 \times [1 - \Phi(8.8)] < .001$ , and the results are highly significant.

# Contingency-Table Method

- The data in the previous example can be represented as a 2×2 contingency table.

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Status	Age at first birth		Total
	≥ 30	≤ 29	
Case	683	2537	3220
Control	1498	8747	10,245
Total	2181	11,284	13,465

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Source: Reprinted with permission of *WHO Bulletin*, 43, 209–221, 1970.

- Row margins
- Column margins
- Grand total

# Significance Testing Using Contingency-Table Approach

- Observed contingency table
- Expected table

General contingency table for the international-study data in Example 10.4 if (1) of  $n_1$  women in the case group,  $x_1$  are exposed and (2) of  $n_2$  women in the control group,  $x_2$  are exposed (that is, having an age at first birth  $\geq 30$ )

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Case-control status	Age at first birth		Total
	$\geq 30$	$\leq 29$	
Case	$x_1$	$n_1 - x_1$	$n_1$
Control	$x_2$	$n_2 - x_2$	$n_2$
Total	$x_1 + x_2$	$n_1 + n_2 - (x_1 + x_2)$	$n_1 + n_2$

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# Computation of Expected Values for Contingency Tables

- Under null hypothesis, the expected number of units in the (1, 1) cell is

$$n_1 \hat{p} = n_1(x_1 + x_2) / (n_1 + n_2)$$

- For the (2, 1) cell, it is

$$n_2 \hat{p} = n_2(x_1 + x_2) / (n_1 + n_2)$$

**Computation of Expected Values for  $2 \times 2$  Contingency Tables** The expected number of units in the  $(i, j)$  cell, which is usually denoted by  $E_{ij}$ , is the product of the  $i$ th row margin multiplied by the  $j$ th column margin, divided by the grand total.

$$E_{11} = \text{expected number of units in the (1, 1) cell} \\ = 3220(2181)/13,465 = 521.6$$

$$E_{12} = \text{expected number of units in the (1, 2) cell} \\ = 3220(11,284)/13,465 = 2698.4$$

$$E_{21} = \text{expected number of units in the (2, 1) cell} \\ = 10,245(2181)/13,465 = 1659.4$$

$$E_{22} = \text{expected number of units in the (2, 2) cell} \\ = 10,245(11,284)/13,465 = 8585.6$$

#### Expected table for the breast-cancer data in Example 10.4

Case-control status	Age at first birth		Total
	$\geq 30$	$\leq 29$	
Case	521.6	2698.4	3220
Control	1659.4	8585.6	10,245
Total	2181	11,284	13,465

# Yates-Corrected Chi-Square Test for 2×2 Contingency Table

The best test is based on statistic  $(O - E)^2 / E$ , where  $O$  and  $E$  are the observed and expected number of units, respectively, in a particular cell.

**Yates-Corrected Chi-Square Test for a 2 × 2 Contingency Table** Suppose we wish to test the hypothesis  $H_0: p_1 = p_2$  versus  $H_1: p_1 \neq p_2$  using a contingency-table approach, where  $O_{ij}$  represents the observed number of units in the  $(i, j)$  cell and  $E_{ij}$  represents the expected number of units in the  $(i, j)$  cell.

(1) Compute the test statistic

$$X^2 = (|O_{11} - E_{11}| - .5)^2 / E_{11} + (|O_{12} - E_{12}| - .5)^2 / E_{12} \\ + (|O_{21} - E_{21}| - .5)^2 / E_{21} + (|O_{22} - E_{22}| - .5)^2 / E_{22}$$

which under  $H_0$  approximately follows a  $\chi^2_1$  distribution.

- (2) For a level  $\alpha$  test, reject  $H_0$  if  $X^2 > \chi^2_{1,1-\alpha}$  and accept  $H_0$  if  $X^2 \leq \chi^2_{1,1-\alpha}$ .
- (3) The approximate  $p$ -value is given by the area to the right of  $X^2$  under a  $\chi^2_1$  distribution.
- (4) Use this test only if none of the four expected values is less than 5.

$$X^2 = \left( |O_{11} - E_{11}| - .5 \right)^2 / E_{11} + \left( |O_{12} - E_{12}| - .5 \right)^2 / E_{12} \\ + \left( |O_{21} - E_{21}| - .5 \right)^2 / E_{21} + \left( |O_{22} - E_{22}| - .5 \right)^2 / E_{22}$$

Assess the breast cancer data in Example 10.4 using contingency-table approach

$$X^2 = \frac{(|683 - 521.6| - .5)^2}{521.6} + \frac{(|2537 - 2698.4| - .5)^2}{2698.4} \\ + \frac{(|1498 - 1659.4| - .5)^2}{1659.4} + \frac{(|8747 - 8585.6| - .5)^2}{8585.6} \\ = 77.89 \sim \chi_1^2 \text{ under } H_0$$

Because  $\chi_{1,.999}^2 = 10.83 < 77.89 = X^2$

$$p < 1 - .999 = .001$$

**Short Computational Form for the Yates-Corrected Chi-Square Test for  $2 \times 2$  Contingency Tables** Suppose we have the  $2 \times 2$  contingency table in Table 10.7. The  $X^2$  test statistic in Equation 10.5 can be written

$$X^2 = n \left( |ad - bc| - \frac{n}{2} \right)^2 / [(a+b)(c+d)(a+c)(b+d)]$$

Thus the test statistic  $X^2$  depends only on (1) the grand total  $n$ , (2) the row and column margins  $a + b$ ,  $c + d$ ,  $a + c$ ,  $b + d$ , and (3) the magnitude of the quantity  $ad - bc$ . To compute  $X^2$ ,

(1) Compute

$$\left( |ad - bc| - \frac{n}{2} \right)^2$$

Start with the first column margin, and proceed counterclockwise.

- (2) Divide by each of the two column margins.
- (3) Multiply by the grand total.
- (4) Divide by each of the two row margins.

**General contingency table**

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$a$	$b$	$a + b$
$c$	$d$	$c + d$
$a + c$	$b + d$	$n = a + b + c + d$

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# Two-Sample Test for Binomial Proportions for Matched-Pair Data (McNemar's Test)

## Ex 10.21

- Comparing two different chemotherapy treatments for breast cancer, A and B.
  - The two groups should be as comparable as possible on other prognostic factors.
- A matched study
  - The patients are assigned to pairs matched on age and clinical conditions
  - A random member of each matched pair gets treatment A and the other gets treatment B.
  - The patients are followed for 5 years, with survival as the outcome variable.

**A  $2 \times 2$  contingency table comparing treatments A and B for breast cancer based on 1242 patients**

Treatment	Outcome		Total
	Survive for 5 years	Die within 5 years	
A	526	95	621
B	515	106	621
Total	1041	201	1242

- Yates-corrected chi-square statistic is 0.59, which is not significant.
- Using this test assumes that the samples are independent.

**A  $2 \times 2$  contingency table with the matched pair as the sampling unit based on 621 matched pairs**

Outcome of treatment A patient	Outcome of treatment B patient		Total
	Survive for 5 years	Die within 5 years	
Survive for 5 years	510	16	526
Die within 5 years	5	90	95
Total	515	106	621

- Probability that the treatment B member of the pair survived given that the treatment A member of the pair survived =  $510/526 = .970$
- Probability that the treatment B member of the pair survived given that the treatment A member of the pair died =  $5/95 = .053$
- Concordant pair
  - A matched pair in which the outcome is the same for each member of the pair.
- Discordant pair
  - A matched pair in which the outcomes differ for the members of the pair.
- Type A discordant pair
  - Treatment A member of the pair has the event and B does not.
- Type B discordant pair
  - Treatment B member of the pair has the event and A does not.



- Let  $p$  = probability that a discordant pair is of type A.
- $H_0: p = 1/2$  versus  $H_1: p \neq 1/2$ .

#### McNemar's Test for Correlated Proportions—Normal-Theory Test

- (1) Form a  $2 \times 2$  table of matched pairs, where the outcomes for the treatment A members of the matched pairs are listed along the rows and the outcomes for the treatment B members are listed along the columns.
- (2) Count the total number of discordant pairs ( $n_D$ ) and the number of type A discordant pairs ( $n_A$ ).
- (3) Compute the test statistic

$$X^2 = \left( \left| n_A - \frac{n_D}{2} \right| - \frac{1}{2} \right)^2 / \left( \frac{n_D}{4} \right)$$

An equivalent version of the test statistic is also given by

$$X^2 = \left( |n_A - n_B| - 1 \right)^2 / (n_A + n_B)$$

- (4) For a two-sided level  $\alpha$  test,

if  $X^2 > \chi_{1,1-\alpha}^2$

then reject  $H_0$ ;

if  $X^2 \leq \chi_{1,1-\alpha}^2$

then accept  $H_0$ .

- (5) The exact  $p$ -value is given by  $p\text{-value} = \Pr(\chi_1^2 \geq X^2)$ .

- (6) Use this test only if  $n_D \geq 20$ .