

# STAT 417 Lecture Note 23

In previous lecture, we showed how to use  $\chi^2$  goodness fit test to test models. Today, we will talk about how to ~~use~~ test relationships among variables. We first look at a general form of

$\chi^2$  - test:

Consider the following  $k \times r$  cells. ( $k, r \geq 2$ ).

	1	2	...	r
1				
2		$O_{22}$ $E_{22}$		
...				
k				

$n$  elements are put into these cells. each element goes to one cell.

Let  $O_{ij}$  = observed frequency of the cell  $C_{ij}$ .

$E_{ij}$  = expected frequency of cell  $C_{ij}$ .

$i=1, 2, \dots, k.$

$j=1, 2, \dots, r.$

Then the  $\chi^2$  - test statistic is

$$\chi^2 = \sum_{i=1}^k \sum_{j=1}^r \frac{(O_{ij} - E_{ij})^2}{E_{ij}} \approx \chi^2((k-1)(r-1))$$

Theorem.  $\chi^2 \sim \chi^2((k-1)(r-1))$ . Under various null hypothesis.

Next, we show how to use  $\chi^2$ -test to check relationships among variables.

Eg 1. (Independence checking). The most important relationship between variables is independence. This summer we have  $n=10$  graduate students, from whom we get samples  $(X_1, Y_1), \dots, (X_{10}, Y_{10})$ .

where  $X_i = \begin{cases} 1, & \text{if the } i\text{th student has taken STAT 417.} \\ 2, & \text{if not.} \end{cases}$

$Y_i = \begin{cases} 1, & \text{doctor} \\ 2, & \text{lawyer} \\ 3, & \text{scientist} \\ 4, & \text{unemployed.} \end{cases}$

The 10 samples are below.

$(1, 1), (2, 4), (1, 2), (1, 3), (1, 2), (2, 4), (2, 4), (1, 3), (2, 4), (1, 3)$ .

Summarizing the samples in a  $2 \times 4$  cell.

	$Y=1$	$Y=2$	$Y=3$	$Y=4$	
$X=1$	1 $O_{11}$	2 $O_{12}$	3 $O_{13}$	0 $O_{14}$	6
$X=2$	0 $O_{21}$	0 $O_{22}$	0 $O_{23}$	4 $O_{24}$	4
	1	2	3	4	

Question: Does "taking STAT 417" affect your job hunting?

To test this, we just have to check whether  $X$  and  $Y$  are independent. The observed frequency in each cell is easy to collect.

Next we should find the expected frequency under independence assumption.

that is, the  $E_{ij}$ .

If  $X$  and  $Y$  are independent, then

$$p_{11} = P(X=1, Y=1) = P(X=1)P(Y=1) = (0.6)(0.1) = 0.06$$

$$p_{12} = P(X=1, Y=2) = P(X=1)P(Y=2) = (0.6)(0.2) = 0.12$$

$$p_{13} = P(X=1, Y=3) = P(X=1)P(Y=3) = (0.6)(0.3) = 0.18$$

$$p_{14} = \cancel{P(X=1, Y=4)} = P(X=1)P(Y=4) = (0.6)(0.4) = 0.24$$

$$p_{21} = P(X=2, Y=1) = P(X=2)P(Y=1) = (0.4)(0.1) = 0.04$$

$$p_{22} = P(X=2, Y=2) = P(X=2)P(Y=2) = (0.4)(0.2) = 0.08$$

$$p_{23} = P(X=2, Y=3) = P(X=2)P(Y=3) = (0.4)(0.3) = 0.12$$

$$p_{24} = P(X=2, Y=4) = P(X=2)P(Y=4) = (0.4)(0.4) = 0.16$$

$$\text{So } E_{11} = \cancel{(10)} n p_{11} = (10)(0.06) = 0.6$$

$$E_{12} = n p_{12} = (10)(0.12) = 1.2$$

$$E_{13} = n p_{13} = (10)(0.18) = 1.8$$

$$E_{14} = n p_{14} = (10)(0.24) = 2.4$$

$$E_{21} = n p_{21} = (10)(0.04) = 0.4$$

$$E_{22} = n p_{22} = (10)(0.08) = 0.8$$

$$E_{23} = n p_{23} = (10)(0.12) = 1.2$$

$$E_{24} = n p_{24} = (10)(0.16) = 1.6$$

The  $\chi^2$ -test is

$$\chi^2 = \frac{(1-0.6)^2}{0.6} + \frac{(2-1.2)^2}{1.2} + \frac{(3-1.8)^2}{1.8} + \frac{(0-2.4)^2}{2.4}$$
$$+ \frac{(0-0.4)^2}{0.4} + \frac{(0-0.8)^2}{0.8} + \frac{(0-1.2)^2}{1.2} + \frac{(4-1.6)^2}{1.6}$$

$$= 10 > \chi_{0.95}^2(3) = 7.81.$$

So, STAT 417 has significant impact on your future ~~for~~ job hunting.

EX1. Suppose ~~Y is~~  $Y \in \{0, 1\}$  is binary variable.

$X \in \{1, 2, 3, 4\}$  is categorical variable.

$n=100$  samples were collected, and is summarized in the following

$2 \times 4$  table.

12	10	16	14	52
13	15	9	11	48
25	25	25	25	

Test if  $X$  and  $Y$  are independent.

$$\text{Sol: } E_{11} = E_{12} = E_{13} = E_{14} = (0.25)(0.52)(100) = 13.$$

$$E_{21} = E_{22} = E_{23} = E_{24} = (0.25)(0.48)(100) = 12.$$

$$\begin{aligned} \text{So } \chi^2 &= \frac{(12-13)^2}{13} + \frac{(10-13)^2}{13} + \frac{(16-13)^2}{13} + \frac{(14-13)^2}{13} \\ &+ \frac{(13-12)^2}{12} + \frac{(15-12)^2}{12} + \frac{(9-12)^2}{12} + \frac{(11-12)^2}{12} = 3.21 < \chi_{0.95}^2(3) = 7.81. \end{aligned}$$

So there is ~~no~~ evidence that  $X$  and  $Y$  are independent.