

# STAT 135 Lecture 14

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## Remark 0.1 (Test of independence)

“Nice application” of goodness of fit

## Example 0.2 (Two categorical variables)

Survey about taste in music and sports

Sport	Pop	Hiphop	Country	Metal	Jazz
Basketball	40	50			
Football	20				
Soccer					
Baseball					

Is there an association?

Or is there no association? (independence)

$H_0$ : Variables are independent

$H_1$ : Not independent

## Remark 0.3

Table with I rows and J columns, think of this as multinomial with IJ possible outcomes

Probability of being in cell  $ij = \pi_{ij}$

Let  $P_i$  denote the probability of being in role  $i$  and  $q_j$  be the probability of being in column  $j$

$q_j = \sum \pi_{ij}$

$H_0 : \pi_{ij} = p_i q_j$

$H_1$  : Not independent

## Method 0.4

Data  $n_{ij}$  = count in cell  $ij$

Under  $H_0$  we can think of

$i - 1$  row probabilities  $p_1, p_2, \dots, p_I$

$j - 1$  column probabilities  $q_1, q_2, \dots, q_J$

Think of the unknown p's and q's as parameters

$\pi_{ij}(p_i, q_j) = p_i q_j$

### Method 0.5

Proceed “as usual”

1. Estimate the  $p_i$  and  $q_j$  by MLE

$$\hat{p}_i = \frac{n_{i.}}{n} \quad \hat{q}_j = \frac{n_{.j}}{n}$$

2. Under  $H_0$  : Expected count in cell  $ij = n\hat{p}_i\hat{q}_j$

3. Chi squared statistic

$$\sum_{i=1}^I \sum_{j=1}^J \frac{O_{ij} - E_{ij}}{E_{ij}} = \sum_{i=1}^I \sum_{j=1}^J \frac{\left(n_{ij} - \frac{n_{i.}n_{.j}}{n}\right)^2}{\frac{n_{i.}n_{.j}}{n}}$$

$$T \sim \chi_d^2 \rightarrow T \sim \chi_{n-1-r}^2 \sim \chi_{IJ-1-(I-1)-(J-1)}^2 = \chi_{(I-1)(J-1)}^2$$

	1	...	J
1	$n_{1j}$		
$\vdots$			
I			

J columns

} I rows

Let  $n_{i.} = \sum_{\forall j} n_{ij}$        $n_{.j} = \sum_{\forall i} n_{ij}$

### Example 0.6 (Chi Squared Test of Homogeneity)

4 7-sided dice (not fair)

1	2	3	4	5	6	7
$p_1$	$p_2$	$p_3$	$p_4$	$p_5$	$p_6$	$p_7$
Side	Die 1	Die 2	Die 3	Die 4		
1	117					
2	25					
3	$\vdots$					
$\vdots$	$\vdots$					
7	100	250	70	90		

Question: Are all the dice “the same?”

$I = \#$  sides,  $J = \#$  dice

$H_0 : \pi_{i1} = \pi_{i2} = \cdots = \pi_{ij}$  for each  $i, i = 1, \dots, I$

**Solution.**  $J$  multinomial distributions with  $I$  possible outcomes

Under  $H_0$  denote the common  $\pi_{ij}$  by  $\pi_i$

$$\pi_1 = P(X = 1) = \frac{n_{1.}}{n}$$

$$\pi_2 :$$

$$\pi_I = \frac{n_{I.}}{n}$$

$n_{i.}$  = count for side  $i$ ,  $n_{.j}$  = count for die  $j$ ,  $n_{ij}$  = count for side  $i$  on die  $j$

Pick a  $j$

$$T_j = \sum_{i=1}^n \frac{\left(n_{ij} - \frac{n_{i.}n_{.j}}{n}\right)^2}{\frac{n_{i.}n_{.j}}{n}} \quad T_j \sim \chi_{I-1}^2$$

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Sum up over the columns (dice)

$$T = \sum_{j=1}^J \sum_{i=1}^I \frac{(n_{ij} - \frac{n_{i.}n_{.j}}{n})^2}{\frac{n_{i.}n_{.j}}{n}}$$
$$T \sim \chi^2_{J(I-1)-(I-1)} = \chi^2_{(I-1)(J-1)}$$

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**Remark 0.7** (Homogeneity or Independence)

- Questions are very similar
- Is outcome independent of die label?
- Only real difference is the sampling model
- Analysis turns out the same either way

**Method 0.8** (Non-parametric two sample tests)

Don't want to make any assumptions and don't have large samples

Suppose we have

$$X_1, \dots, X_n \stackrel{iid}{\sim} F$$

$$Y_1, \dots, Y_m \stackrel{iid}{\sim} G$$

Consider:

$$H_0 : F = G$$

1. Group all  $m + n$  observations together
2. Sort them by increasing size
3. Just look at ranks
4. Look at sum of ranks of control
5. Reject  $H_0$  if sum of ranks too extreme
6. Look at all  $\binom{m+n}{n}$  ways to assign ranks, all equally likely under  $H_0$
7. Get p-value based on where observed falls