

## Exercises: Contingency tables

Lecturer &amp; author: Georgios Karagiannis

georgios.karagiannis@durham.ac.uk

**Exercise 1.** Consider a  $I \times J \times K$  contingency table, with classification variables  $X, Y, Z$ . Prove that if

1.  $X$  and  $Y$  are conditionally independent on  $Z$ ; and
2.  $X$  and  $Z$  are conditionally independent on  $Y$

then:

$Y$  and  $Z$  are jointly independent from  $X$

**Hint:** Write down the probability forms involved, and try to derive the result by using simple probability calculus.

**Solution 2.** I have

$$\pi_{ijk} = \frac{\pi_{i+k}\pi_{+jk}}{\pi_{++k}} \quad (1)$$

$$\pi_{ijk} = \frac{\pi_{ij+}\pi_{+jk}}{\pi_{+j+}} \quad (2)$$

By dividing those two, we get

$$\pi_{i+k}\pi_{+j+} = \pi_{ij+}\pi_{++k}$$

By summing with respect to  $j$  we get

$$\pi_{i+k} = \pi_{i++}\pi_{++k} \quad (3)$$

By substituting (3) in (1), we get

$$\pi_{ijk} = \pi_{i++}\pi_{+jk}$$

Hence  $Y$  and  $Z$  are jointly independent from  $X$

**Exercise 3.** Consider a  $I \times J \times K$  contingency table, with classification variables  $X, Y, Z$ . Prove that if  $Y$  and  $Z$  are jointly independent from  $X$ ,  
then

1.  $X$  and  $Y$  are conditionally independent on  $Z$ ; and
2.  $X$  and  $Z$  are conditionally independent on  $Y$

**Hint:** Write down the probability forms involved, and try to derive the result by using simple probability calculus.

**Solution.** Because  $Y$  and  $Z$  are jointly independent from  $X$ , I have that

$$\pi_{ijk} = \pi_{i++}\pi_{+jk} \quad (4)$$

By summing (4) with respect to  $j$  we get

$$\pi_{i+k} = \pi_{i++}\pi_{++k} \quad (5)$$

$$\implies \pi_{i++} = \frac{\pi_{i+k}}{\pi_{++k}} \quad (6)$$

By substituting (6) in (4), we get

$$\pi_{ijk} = \frac{\pi_{i+k}\pi_{+jk}}{\pi_{++k}} \quad (7)$$

Hence,  $X$  and  $Y$  are conditionally independent on  $Z$

**Solution 4.** By summing (4) with respect to  $k$  we get

$$\pi_{ij+} = \pi_{i++}\pi_{+j+} \quad (8)$$

$$\implies \pi_{i++} = \frac{\pi_{ij+}}{\pi_{++k}} \quad (9)$$

By substituting (9) in (4), we get

$$\pi_{ijk} = \frac{\pi_{ij+}\pi_{+jk}}{\pi_{++k}} \quad (10)$$

Hence,  $X$  and  $Z$  are conditionally independent on  $Y$

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The next exercise is from Homework 1

**Exercise 5.** The 1988 General Social Survey compiled by the National Opinion Research Center asked: “Do you support or oppose the following measures to deal with AIDS? (1) Have the government pay all of the health care costs of AIDS patients; (2) Develop a government information program to promote safe sex practices, such as the use of condoms. Table 1 summarizes opinions about health care costs (H) and the information program (I), classified also by the respondent’s gender (G).

Gender (G)	Information Opinion (I)	Health Opinion (H)	
		Support	Oppose
Male	Support	76	160
	Oppose	6	25
Female	Support	114	181
	Oppose	11	48

Table 1: Source: 1988 General Social Survey, National Opinion Research Center.

1. Compute the marginal GH-table
2. For the GH-table, compute the MLE of the marginal odds ratio, the confidence intervals. Interpret the result. (sig. level 5%)
3. Perform a hypothesis test, in order to test if the Information Opinion and the Health Opinion are independent at each level of the Gender. (sig. level 5%)
4. Compute the partial (conditional) IH odds ratio at each level of the Gender. Interpret the result.

**Solution.** <sup>1</sup>

1. The marginal table is

	Health Opinion (H)		
Gender (G)	Support	Oppose	Total
Male	82	185	267
Female	125	229	354
Total	207	414	621

2. The MLE of the marginal odds ratio is

$$\hat{\theta}^{HO} = \frac{80 \times 229}{185 \times 125} = 0.8120216$$

Overall, it is more likely for a Female to support health care, that what it is for a Male.

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<sup>1</sup>R-script is available to double check.

The 95% confidence interval for  $\log(\theta^{HO})$  is

$$\begin{aligned} & \{\log(\hat{\theta}) \pm z_{1-\frac{0.05}{2}} \sqrt{\sum_{i,j} \frac{1}{n_{i,j}}}\} \\ & \{\log(0.8120216) \pm 1.959964 \times \sqrt{0.1731108}\} \\ & \{-0.5475192, 0.1310626\} \end{aligned}$$

The 95% confidence interval for  $\theta^{HO}$  is

$$\theta^{HO} \in (e^{-0.5475192}, e^{0.1310626}) = (0.57, 1.14)$$

At sig. level 5%, we cannot reject the hypothesis that the Gender and the Health opinion are independent.

3. Based on this re-ordered table, I perform the Mantel-Haenszel Chi-Squared Test

$$\begin{aligned} & \begin{cases} H_0 : & I, H \text{ are independent accross the partial tables at each level of } G \\ H_1 : & I, H \text{ are not independent accross the partial tables at each level of } G \end{cases} \\ & \implies \begin{cases} H_0 : & \theta_{(1)}^{I,H} = \theta_{(2)}^{I,H} = 1 \\ H_1 : & \theta_{(1)}^{I,H} \neq \theta_{(2)}^{I,H} \end{cases} \end{aligned}$$

The statistic is

$$T_{MH} = \frac{[\sum_k (n_{11k} - \mu_{11k})]^2}{\sum_k \sigma_{11k}^2} \xrightarrow{D} \chi_{df}^2$$

where

$$\mu_{11k} = \frac{n_{1+k}n_{+1k}}{n_{++k}}$$

and

$$\sigma_{11k}^2 = \frac{n_{1+k}n_{2+k}n_{+1k}n_{+2k}}{n_{++k}^2(n_{++k} - 1)}$$

We reject the Null hypothesis for large values. The Rejection area at  $\alpha$  sig. level is

$$R = \{T_{MH}^{obs} \geq \chi_{df, 1-\alpha}^2\}$$

Calculations:

$n_{i,+,k}$	1	2	$n_{+,j,k}$	1	2
1	236	295	1	82	125
2	31	59	2	185	229
$n_{+,+,k}$			267	354	
$\hat{\mu}_{1,1,k}$			72.4794	104.1667	
$\sigma_{1,1,k}^2$			5.852685	11.262590	

The observed statistic is

$$T_{MH}^{obs} = \frac{178.3275}{17.11528} = 10.4192$$

The degrees of freedom are

$$df = 1$$

and the critical value at  $\alpha = 5\%$  sig. level is

$$\chi_{1,0.95}^2 = 3.841459$$

Because the Rejection area at  $\alpha = 5\%$  sig. level is

$$R = \{T_{MH}^{obs} \geq \chi_{df,1-\alpha}^2\}$$

I reject the null hypothesis that the Health Opinion and the Information Opinion are independent at each level of Gender at sig. level 5%.

4. For Males Male, the MLE of the odds ratio is

$$\hat{\theta}_{(1)}^{IH} = \frac{n_{111}n_{221}}{n_{121}n_{211}} = 1.979167$$

showing that it is more likely for a supporter of the Information program to support the opinion for the Government to pay health care costs.

For Females, the MLE of the odds ratio is

$$\hat{\theta}_{(2)}^{IH} = \frac{n_{112}n_{222}}{n_{122}n_{212}} = 2.748368$$

showing that it is more likely for a supporter of the Information program to support the opinion for the Government to pay health care costs.

In fact the dependency between Information Opinion and the Health Opinion is stronger in Female.

**Exercise 6.** Let  $X$  and  $Y$  be discrete random variables with possible values  $x_1, \dots, x_m$  and  $y_1, \dots, y_n$  respectively and such that  $p_{ij} > 0$  for all  $i$  and  $j$ . Here,  $p_{ij}$  denotes

$$\Pr(X = x_i \cap Y = y_j).$$

Let  $p_{i|j}$  denote  $p_{ij}/p_j$  where  $p_j$  denotes  $\sum_i p_{ij}$  so that  $p_j = \Pr(Y = y_j)$  and  $p_{i|j} = \Pr(X = x_i | Y = y_j)$ . Also let  $p_i$  denote  $\sum_j p_{ij}$ .

Then the following statements are equivalent:

1.  $X$  and  $Y$  are independent;
2.  $p_{i|j}$  does not depend on  $j$  for all  $i$ ;
3. there exist  $g_1, \dots, g_m$  and  $h_1, \dots, h_n$  such that  $p_{ij} = g_i h_j$  for all  $i$  and  $j$ ;
4. the “odds”

$$\frac{p_{i|j}}{p_{i'|j}}$$

for values of  $X$  given  $Y = y_j$  do not depend on  $j$  for all  $i$  and  $i'$ ;

5. the “odds ratios”

$$\left( \frac{p_{i|j}}{p_{i'|j}} \right) / \left( \frac{p_{i|j'}}{p_{i'|j'}} \right) = 1$$

for all  $i, i', j$  and  $j'$ .

**Solution.**

- (2) trivially implies (4).
- From (3),  $p_j = \sum_i (g_i h_j) = h_j \sum_i g_i$  and so  $p_{i|j} = g_i / \sum_i g_i$  which does not depend on  $j$ .

Thus (3) implies (2).

- From (4)

$$\frac{p_{i|j}}{p_{i'|j}} = k_{ii'}$$

for some matrix  $k$  and therefore  $p_{ij} = k_{i1} p_{1j}$  and taking  $g_i = k_{i1}$  and  $h_j = p_{1j}$ , we have (3).

Thus (4) implies (3).

- From (1),  $p_{ij} = p_i p_j$  and this is (3) if we take  $g_i = p_i$  and  $h_j = p_j$ .

Thus (1) implies (3).

- From (2), there exist  $k_i$  such that  $p_{i|j} = k_i$  for all  $i$  and  $j$ . This implies that  $p_{ij} = k_i p_j$  which implies that  $p_i = k_i \sum_j p_j = k_i$  and so  $p_{ij} = p_i p_j$  for all  $i$  and  $j$ .

Thus (2) implies (1).

- (4) and (5) are trivially equivalent.

The first three bullet points show that (2), (3) and (4) are equivalent. The next two bullets show that they are also equivalent to (1) and the final bullet point that they are equivalent to (5).

The next exercise is from Problem Class 1

**Exercise 7.** <sup>2</sup> The 674 subjects classified in Table 2 were the defendants in indictments involving cases with multiple murders in Florida between 1976 and 1987. The variables in Table 2 are

**Y:** death penalty verdict, with categories ( $j = 1$ : Yes, 2: No)

**X:** race of Defendant, with categories ( $i = 1$ : White, 2: Black)

**Z:** race of Victim, with categories ( $k = 1$ : White, 2: Black)

Victim's Race (Z)	Defendant's Race (X)	Death Penalty (Y)	
		Yes	No
White	White	53	414
	Black	11	37
Black	White	0	16
	Black	4	139

Table 2: Death Penalty Verdict by Defendant's Race and Victims' Race

**HINT:** Regarding the computation of Odds ratio, if the matrix has ZERO cells we use a correction by adding +0.5 at each cell. Precisely,

$$\theta = \frac{(n_{11} + 0.5)(n_{22} + 0.5)}{(n_{21} + 0.5)(n_{12} + 0.5)}$$

if any of  $n_{11}$ ,  $n_{12}$ ,  $n_{21}$ ,  $n_{22}$  is zero. This is just a treatment, so that we do not get in infinite numbers

**E.g.**, in this exercise, as the (1, 1, 2)th cell is  $n_{112} = 0$ , the above remedy is applied for the computation of  $\theta_{(2)}^{XY}$ ,  $\theta_{(1)}^{ZY}$ , etc...

<sup>2</sup>R-script is available to double check. [https://htmlpreview.github.io/?https://github.com/georgios-stats/Topics\\_in\\_Statistics\\_Michaelmas\\_2020/blob/master/Contingency\\_Tables/q7\\_R.nb.html](https://htmlpreview.github.io/?https://github.com/georgios-stats/Topics_in_Statistics_Michaelmas_2020/blob/master/Contingency_Tables/q7_R.nb.html)

1. Compute the XY-marginal table, and the marginal counts
2. Based on the XY-marginal table,
  - (a) Compute the conditional proportion of the White defendants received a death penalty, and that of the Black defendants received a death penalty. What do you observe?
  - (b) Compute the MLE of the odd ratio and the asymptotic confidence interval at 5% sig. level. What is the association between Death Penalty and the Defendant's Race?
  - (c) Perform a Goodness of fit test to test if the Death Penalty and the Defendant's Race are independent at 5% sig. level.
3. Test the Hypothesis that the Death Penalty and the Defendant's Race are independent across the Victim's Race levels at sig. level 5%.
4. Based on the XY-partial table,
  - (a) Compute the MLE of the conditional XY odds ratios at each level of Victim's Race (Z),
  - (b) Test the hypothesis that the Death Penalty and Defendant's Race are independent when the Victim is White, and against the alternative hypothesis that it is more likely for the Death penalty to be imposed for a Black defendant when the Victim is White. (Sig. level 5% )
  - (c) Test the hypothesis that the Death Penalty and Defendant's Race are independent when the Victim is Black, and against the general alternative hypothesis that they are dependent when the Victim in Black. (Sig. level 5% )
5. Test the hypothesis that the Death Penalty and the Victim's Race are independent across the Defendant's Race levels (Sig. level 5%).
6. Compute the ZY conditional odds ratios for each level of the Defendant's Race, and discuss what they imply.
7.
  - (a) Compute the marginal YZ-table.
  - (b) Compute the marginal YZ-odds ratios.
  - (c) Test the hypothesis that the Death Penalty and the Victims Race are independent against the alternative that it is more likely for a Death Penalty to be imposed when the Victim is White, than what it is when the Victim is Black. (Sig. level 5%)
8.
  - (a) Compute the marginal XZ-table.



- (b) Compute the marginal XZ-odds ratio.
- (c) Test the hypothesis that the Defendant's Race and the Victims rate are independent, against the alternative that it is more likely for a Victim to be White when the defendant is White, than what it is when the Defendant is Black. (Sig. level 5%)

9. Any comments?

**Solution.**

1. The XY-marginal table is

	Death Penalty (Y)		
Defendant's Race (X)	Yes	No	Total
White	53	430	483
Black	15	176	191
Total	68	606	674

2.

- (a) The conditional proportion of the White defendants received a death penalty is

$$p_{j=1|i=1} = \frac{53}{483} = 0.109$$

Namely, the 11% of the White defendants received death penalty

The conditional proportion of the Black defendants received a death penalty is

$$p_{j=1|i=2} = \frac{15}{191} = 0.079$$

Namely, the 7.9% of the White defendants received death penalty

- (b) The MLE of the marginal odds ratio is

$$\hat{\theta} = \frac{n_{11}n_{22}}{n_{12}n_{21}} = \frac{53 \times 176}{15 \times 430} = 1.446202$$

The asymptotic 95% CI for  $\log(\theta)$  is

$$\begin{aligned} & \{\log(\hat{\theta}) \pm z_{1-\frac{0.05}{2}} \sqrt{\sum_{i,j} \frac{1}{n_{i,j}}}\} \\ & \{\log(1.446202) \pm 1.959964 \times \sqrt{0.09354202}\} \\ & \{-0.2305073, 0.9683883\} \end{aligned}$$

So the 95% CI for the marginal odds ratio  $\theta$  is

$$(0.7941306, 2.6336964)$$

The observed XY-marginal odds ratio implies that it is more likely a death penalty to be received given that the defender is White.

(c) The pair of hypothesis test is

$H_0$  :X and Y are independent

$H_1$  :X and Y are not independent

The MLE of the counts under the null hypothesis are

$$\hat{\mu}_{i,j} = \frac{n_{i,+}n_{+,j}}{n_{+,+}}, \text{ for } i = 1, 2, j = 1, 2$$

I calculate

X	Y	$n_{i,j}$	$\hat{\mu}_{i,j}$	$\frac{(n_{i,j} - \hat{\mu}_{i,j})^2}{\hat{\mu}_{i,j}}$	$n_{i,j} \log \frac{n_{i,j}}{\hat{\mu}_{i,j}}$
White	Yes	53	48.72997	0.3741671	8.903753
White	No	430	434.27	0.04198575	-8.497935
Black	Yes	15	19.27003	0.9461923	-7.515025
Black	No	176	171.73	0.10617339	8.645364
Total				1.468519	1.536156

The chi-square Pearson's chi-square statistic

$$X^2 = \sum_{i,j} \frac{(n_{i,j} - \hat{\mu}_{i,j})^2}{\hat{\mu}_{i,j}}$$

The Rejection area at  $\alpha = 5\%$  sig. level is

$$R = \{X^{obs} \geq \chi_{df,1-\alpha}^2\} = \{1.468519 \geq 3.841459\}$$

The likelihood ratio statistics is

$$G^2 = 2 \sum_{i,j} n_{i,j} \log \frac{n_{i,j}}{\hat{\mu}_{i,j}}$$

The Rejection area at  $\alpha = 5\%$  sig. level is

$$R = \{X^{obs} \geq \chi_{df,1-\alpha}^2\} = \{1.536156 \geq 3.841459\}$$

It cannot reject the null hypothesis that the Death Penalty and the Dependence race are independent at 5% sig. level.

3. We perform the Mantel-Haenszel Chi-Squared Test

$$\begin{cases} H_0 : & X, Y \text{ are independent across the partial tables at each level of } Z \\ H_1 : & X, Y \text{ are not independent across the partial tables at each level of } Z \end{cases}$$

$$\implies \begin{cases} H_0 : & \theta_{(1)}^{xy} = \theta_{(2)}^{xy} = 1 \\ H_1 : & \theta_{(1)}^{xy} \neq \theta_{(2)}^{xy} \end{cases}$$

The statistic is

$$T_{MH} = \frac{[\sum_k (n_{11k} - \mu_{11k})]^2}{\sum_k \sigma_{11k}^2} \xrightarrow{D} \chi_{df}^2$$

where

$$\mu_{11k} = \frac{n_{1+k}n_{+1k}}{n_{++k}}$$

and

$$\sigma_{11k}^2 = \frac{n_{1+k}n_{2+k}n_{+1k}n_{+2k}}{n_{++k}^2(n_{++k} - 1)}$$

We reject the Null hypothesis for large values. The Rejection area at  $\alpha$  sig. level is

$$R = \{T_{MH}^{obs} \geq \chi_{df, 1-\alpha}^2\}$$

Calculations:

$n_{i,+,k}$	1	2	$n_{+,j,k}$	1	2
1	467	16	1	64	4
2	48	155	2	451	155
$n_{+,+,k}$			515	159	
$\hat{\mu}_{1,1,k}$			58.03495	0.4025157	
$\sigma_{1,1,k}^2$			4.746102	0.3551374	

The observed statistic is

$$T_{MH}^{obs} = \frac{29.56605}{5.101239} = 5.7959$$

The degrees of freedom are

$$df = 1$$

and the critical value at  $a = 5\%$  sig. level is

$$\chi_{1,0.95}^2 = 3.841459$$

Because the Rejection area at  $a = 5\%$  sig. level is

$$R = \{T_{MH}^{obs} \geq \chi_{df,1-a}^2\}$$

I reject the null hypothesis that the the Death Penalty and the Defendant's Race are independent across the Victim's Race levels at sig. level 5%

4.

(a) When the Victim is White the MLE of the partial odds ratio is

$$\hat{\theta}_{(1)}^{XY} = \frac{n_{111}n_{221}}{n_{121}n_{211}} = 0.4306105$$

showing negative dependence between X and Y given Z=1.

When the Victim is Black the MLE of the marginal odds ratio is

$$\hat{\theta}_{(2)}^{XY} = \frac{(n_{112} + 0.5)(n_{222} + 0.5)}{(n_{122} + 0.5)(n_{212} + 0.5)} = 0.9393939$$

implying a slight negative dependence between X and Y given Z=2.

(b) The pair of hypothesis tests is

$$\begin{cases} H_0 : \theta_{(1)}^{xy} = 1 \\ H_1 : \theta_{(1)}^{xy} < 1 \end{cases}$$

The statistic is

$$Z = \frac{\log(\hat{\theta}_{(1)}^{xy}) - \log(1)}{\sqrt{\sum_{i,j} \frac{1}{n_{i,j,1}}}} \xrightarrow{H_0} N(0, 1)$$

and because I reject for small values the rejection area at  $a$  sig. level is

$$R = \{Z^{obs} \leq -z_a\}$$

The observed statistic is

$$Z^{obs} = \frac{-0.8425513 - 0}{0.3731213} = -2.258117$$

and the critical value is  $z_{0.05} = -1.644854$ . Therefore, at sig. level 5%, I reject the Null hypothesis against the alternative that it is more likely for the Death penalty to be

imposed for a Black defendant than a White one when the Victim is White.

(c) The pair of hypothesis tests is

$$\begin{cases} H_0 : \theta_{(2)}^{xy} = 1 \\ H_1 : \theta_{(2)}^{xy} \neq 1 \end{cases}$$

The statistic is

$$Z = \frac{\log(\hat{\theta}_{(2)}^{xy}) - \log(1)}{\sqrt{\sum_{i,j} \frac{1}{n_{i,j,1}}}} \xrightarrow{H_0} N(0, 1)$$

and because I reject for small values the rejection area at  $\alpha$  sig. level is

$$R = \{|Z^{obs}| \geq z_{1-\frac{\alpha}{2}}\}$$

The observed statistic is

$$Z^{obs} = \frac{-0.0625204 - 0}{1.513274} = -0.04131464$$

and the critical value is  $z_{0.975} = 1.959964$ . Therefore, at sig. level 5%, I cannot reject the null hypothesis that the Death Penalty and Defendant's Race are independent when the Victim is Black.

5. We perform the Mantel-Haenszel Chi-Squared Test

$$\begin{cases} H_0 : Z, Y \text{ are independent across the partial tables at each level of } X \\ H_1 : Z, Y \text{ are not independent across the partial tables at each level of } X \end{cases}$$

$$\Rightarrow \begin{cases} H_0 : \theta_{(1)}^{ZY} = \theta_{(2)}^{ZY} = 1 \\ H_1 : \theta_{(1)}^{ZY} \neq \theta_{(2)}^{ZY} \end{cases}$$

The statistic is

$$T_{MH} = \frac{\sum_i (n_{i11} - \mu_{i11})^2}{\sum_k \sigma_{i11}^2} \xrightarrow{D} \chi_{df}^2$$

where

$$\mu_{i11} = \frac{n_{i+1}n_{i1+}}{n_{i++}}$$

and

$$\sigma_{i11}^2 = \frac{n_{i+1}n_{i+2}n_{i1+}n_{i2+}}{n_{i++}^2(n_{i++} - 1)}$$

and we reject the Null hypothesis for large values. The Rejection area at  $\alpha$  sig. level is

$$R = \{T_{MH}^{obs} \geq \chi_{df, 1-\alpha}^2\}$$

Calculations:

$n_{i,+k}$	1	2
1	467	16
2	48	143

$n_{i,j,+}$	1	2
1	53	430
2	15	176

$n_{i,+,+}$	483	191
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$\hat{\mu}_{i,1,+}$	51.24	3.769
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$\sigma_{i,1,1}^2$	1.514398	2.614333
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The observed statistic is

$$T_{MH}^{obs} = \frac{80.74928}{4.12873} = 19.5579$$

The degrees of freedom are

$$df = 1$$

and the critical value at  $a = 5\%$  sig. level is

$$\chi_{1,0.95}^2 = 3.841459$$

Because the Rejection area at  $a = 5\%$  sig. level is

$$R = \{T_{MH}^{obs} \geq \chi_{df,1-a}^2\}$$

I reject the null hypothesis that the Death Penalty and the Victim's Race are independent across the Defendant's Race levels at sig. level 5%.

6. When the Defendant is White the MLE of the marginal odds ratio is

$$\hat{\theta}_{(1)}^{ZY} = \frac{(n_{111} + 0.5)(n_{122} + 0.5)}{(n_{112} + 0.5)(n_{121} + 0.5)} = 4.259349$$

showing positive dependence between Z and Y given X=1.

When the Defendant is Black the MLE of the marginal odds ratio is

$$\hat{\theta}_{(2)}^{ZY} = \frac{n_{211}n_{222}}{n_{212}n_{221}} = 10.33108$$

showing positive dependence between Z and Y given X=2.

- regardless of defendant's race, the death penalty was considerably more likely when the victims were white than when the victims were black.

7.

(a) The YZ marginal table is

	Death Penalty (Y)	
Victim's Race (Z)	Yes	No
White	64	451
Black	4	155

(b) The observed marginal odds ratio  $\theta^{YZ}$  is

$$\hat{\theta}^{YZ} = 5.498891$$

The odds ratio shows that the death penalty was considerably more likely when the victims were white than when the victims were black.

(c) The pair of hypothesis tests is

$$\begin{cases} H_0 : \theta^{YZ} = 1 \\ H_1 : \theta^{YZ} > 1 \end{cases}$$

The statistic is

$$Z = \frac{\log(\hat{\theta}^{YZ}) - \log(1)}{\sqrt{\sum_{i,j} \frac{1}{n_{i,j,1}}}} \xrightarrow{H_0} N(0, 1)$$

and because I reject for large values the rejection area at  $\alpha$  sig. level is

$$R = \{Z^{obs} \geq z_{1-\alpha}\}$$

The observed statistic is

$$Z^{obs} = \frac{1.704546 - 0}{0.5237308} = 3.25$$

and the critical value is  $z_{0.95} = 1.644854$ . Therefore, at sig. level 5%, I reject the null hypothesis against the alternative that it is more likely for a Death Penalty to be imposed when the Victim is White, than what it is when the Victim is Black.

8.

(a) The XZ marginal table is

	Victim's Race (Z)	
Defendant's Race (X)	White	Black
White	467	16
Black	48	143

(b) The observed marginal odds ration  $\theta^{XZ}$  is

$$\hat{\theta}^{XZ} = 86.95443$$

The odds that a white defendant had white victims are estimated to be 87.0 times the odds that a black defendant had white victims.

(c) The pair of hypothesis tests is

$$\begin{cases} H_0 : \theta^{XZ} = 1 \\ H_1 : \theta^{XZ} > 1 \end{cases}$$

The statistic is

$$Z = \frac{\log(\hat{\theta}^{XZ}) - \log(1)}{\sqrt{\sum_{i,j} \frac{1}{n_{i,j,1}}}} \xrightarrow{D} N(0, 1)$$

and because I reject for large values the rejection area at  $\alpha$  sig. level is

$$R = \{Z^{obs} \geq z_{1-\alpha}\}$$

The observed statistic is

$$Z^{obs} = \frac{4.465384 - 0}{0.304085} = 14.68466$$

and the critical value is  $z_{0.95} = 1.644854$ . Therefore, at sig. level 5%, I reject the null hypothesis against the alternative that it is more likely for a Victim to be white when the defendant is White.

9. This is called 'Simpson's Paradox'; namely, when the X, Y marginal association has a different direction from X,Y conditional associations at each each level of Z.

Here we observed that:

- According to (Q. 2): overall, the death penalty was appeared to be imposed more often for White Defendants than for Black Defendants. However, we cannot reject the hypothesis that the Defendant's Race and the Death Penalty are marginally independent at sig. level 5%.
- According to (Q. 4): when the Victim was White, the death penalty was imposed more often for Black Defendants than for White Defendants, and this association is sig. at sig. level 5%.
- According to (Q. 4): when the Victim was Black, the death penalty was imposed more often for Black Defendants than for White Defendants. However, at sig. level 5%, the Death Penalty and Defendant's Race are independent when the Victim is Black.



- According to (Q. 7), overall, at sig. level 5%, it is more likely for a Victim to be White when the defendant is White.
  - According to (Q. 6), regardless of defendant's race, the death penalty was considerably more likely when the Victims were White than when the Victims were Black.
  - The explanation is the following. The data imply that whites are tending to kill whites, and killing whites is more likely to result in the death penalty. This suggests that the marginal association should show a greater tendency for white defendants to receive the death penalty than do the conditional associations.
- [Agresti (2007) An Introduction to Categorical Data Analysis, Wiley]

**Exercise 8.** Assume a  $2 \times 2$  table whose marginal counts  $n_{i+}$ ,  $n_{+j}$  are fixed and predetermined when the experiment was performed.

1. Derive the sampling distribution of the experiment. In particular derive, the sampling distribution of the counts of (1, 1)th cell  $N_{11}$ .

The Hypergeometric experiment<sup>a</sup>

- Let, r.v.  $X$  be the number of successes in  $n$  draws without replacement from a finite population of size  $N$ , that contains exactly  $K$  objects with the feature defining success and  $N - K$  objects with a feature defining failure.
- Then  $X$  is distributed according to the Hypergeometric distribution , as

$$X \sim \text{Hg}(N, K, n)$$

with probability

$$\Pr(X = x|N, K, n) = \frac{\binom{K}{x} \binom{N-K}{n-x}}{\binom{N}{n}} 1(X \in \mathcal{X})$$

where  $\mathcal{X} = \{x \in \mathbb{N} | x \in (\max(0, n + K - N), \min(K, n))\}$

<sup>a</sup>Notation is restricted in the scope of this box.

2. Let

$$\theta = \frac{\pi_{11}\pi_{22}}{\pi_{12}\pi_{21}}$$

be odds ratio.

(a) Derive the exact p-value associated to the hypothesis test

$$H_0 : \theta = 1 \text{ vs. } H_1 : \theta > 1$$

(b) Derive the exact p-value associated to the hypothesis test

$$H_0 : \theta = 1 \text{ vs. } H_1 : \theta < 1$$

3. (Famous Tea-Milk example) Dr Bristol was claiming that she could tell if the milk was poured before the tea or if the tea was poured before the milk in the cup. Fisher did not believe her. Fisher set 8 cups of tea. In 4 cups he poured first the tea and then the milk, while in the rest 4 cups he poured first the milk and then the tea. Dr Bristol drank all of them and then she said which 4 has the milk poured first.

This experiment guaranteed that both the column and row marginal counts  $n_{i+}$ ,  $n_{+j}$  were fixed. The 2 way observed count table is presented below

		poured first (guess)		total
		Milk	Tea	
poured first (reality)	$n_{i,j}$	3	1	4
		1	3	4
total		4	4	8

Perform a suitable (exact) statistical hypothesis test (at sig. level 5%) addressing the aforementioned dispute. To derive this test, use the exact sampling distribution derived in the 1st part of the exercise.

### Solution.

1. The random counts of the (1,1)th cell are distributed as

$$N_{11} \sim \text{Hg}(n_{++}, n_{1+}, n_{+1})$$

with

$$\Pr(N_{11} = t | n_{++}, n_{1+}, n_{+1}) = \frac{\binom{n_{1+}}{t} \binom{n_{++} - n_{1+}}{n_{+1} - t}}{\binom{n_{++}}{n_{+1}}} 1(X \in \mathcal{X})$$

where  $\mathcal{X} = \{x \in \mathbb{N} | x \in (\max(0, n_{1+} + n_{+1} - n_{++}), \min(n_{1+}, n_{+1}))\}$ .

This is because the marginal counts  $n_{i+}$ ,  $n_{+j}$  are fixed. Assume we wish to find the probability that  $\Pr(N_{11} = t)$ . This experiment is like from a finite population  $n_{++}$  I get  $n_{+1}$  and from those  $n_{+1}$ , I get  $\{N_{11} = t\}$  from the one characteristic and  $\{N_{21} = n_{+1} - t\}$  from the other.

Hence, there are in total  $\binom{n_{++}}{n_{+1}}$  different combinations in total; while there are  $\binom{n_{1+}}{t} \binom{n_{++}-n_{1+}}{n_{+1}-t}$  different combinations for  $\{N_{11} = t\}$  and  $\{N_{21} = n_{+1} - t\}$ . So the Probability is

$$\Pr(N_{11} = t | n_{++}, n_{1+}, n_{+1}) = \frac{\binom{n_{1+}}{t} \binom{n_{++}-n_{1+}}{n_{+1}-t}}{\binom{n_{++}}{n_{+1}}} 1(X \in \mathcal{X})$$

The distribution that describes the aforementioned experiment, and probability mass function, is the Hyper-geometric distribution

$$N_{11} \sim \text{Hg}(n_{++}, n_{1+}, n_{+1})$$

2.

(a) For the pair of hypothesis

$$H_0 : \theta = 1 \text{ vs. } H_1 : \theta > 1$$

the pvalue is

$$\text{p-value} = \Pr(N_{11} \geq n_{11}) = \sum_{t=n_{11}}^n \Pr_{\text{Hg}}(N_{11} = t)$$

(b) For the pair of hypothesis

$$H_0 : \theta = 1 \text{ vs. } H_1 : \theta < 1$$

the pvalue is

$$\text{p-value} = \Pr(N_{11} \leq n_{11}) = \sum_{t=0}^{n_{11}} \Pr_{\text{Hg}}(N_{11} = t)$$

3. It is the following statistical hypothesis test.

$$H_0 : \theta = 1 \text{ (independence); } H_1 : \theta > 1 \text{ (positive dependence)}$$

As I have the exact sampling distribution, I will use this and not the asymptotic one.

I have  $N_{11} \sim \text{Hg}(n_{++} = 8, n_{1+} = 4, n_{+1} = 4)$  where  $N_{11} \in \{0, 4\}$ . Also I have  $n_{11}^{obs} = 3$ . So the pvalue is

$$\begin{aligned} \text{p-value} &= \Pr(N_{11} \geq 3) = \sum_{t=3}^8 \Pr_{\text{Hg}}(N_{11} = t) \\ &= \Pr_{\text{Hg}}(N_{11} = 3) + \Pr_{\text{Hg}}(N_{11} = 4) \\ &\quad + \cancel{\Pr_{\text{Hg}}(N_{11} = 5)}^0 + \cancel{\Pr_{\text{Hg}}(N_{11} = 6)}^0 + \cancel{\Pr_{\text{Hg}}(N_{11} = 7)}^0 + \cancel{\Pr_{\text{Hg}}(N_{11} = 8)}^0 \\ &= 0.229 + 0.014 + 0 + 0 + 0 + 0 = 0.243 > 0.05 \end{aligned}$$

So Dr. Bristol cannot tell if the Milk or the Tea was poured first in the cup, at sig. level 5%

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