

**STA 250/MTH 342 Intro to Mathematical Statistics**  
**Homework 11**

**Solution 1:** Exercise 10.1.8 (page 633)

When  $H_0$  holds true, we denote  $X$  the height of a man selected randomly from the city, so  $Z = X - 68 \sim \mathcal{N}(0, 1)$ . We have

$$\begin{aligned} p_1^0 &= \Pr(X < 66) = \Pr(Z < -2) = 0.02275, \\ p_2^0 &= \Pr(66 \leq X < 67.5) = \Pr(-2 \leq Z < -0.5) = 0.2858, \\ p_3^0 &= \Pr(67.5 \leq X < 68.5) = \Pr(-0.5 \leq Z < 0.5) = 0.3829, \\ p_4^0 &= \Pr(68.5 \leq X < 70) = \Pr(0.5 \leq Z < 2) = 0.2858, \\ p_5^0 &= \Pr(X \geq 70) = \Pr(Z \geq 2) = 0.02275. \end{aligned}$$

Since the sample size is  $n = 500$ , we compare the frequency and its expectation under  $H_0$  in the following table.

	$N_i$	$np_i^0$
$X < 66$	18	11.35
$66 \leq X < 67.5$	177	142.9
$67.5 \leq X < 68.5$	198	191.5
$68.5 \leq X < 70$	102	142.9
$X \geq 70$	5	11.35

We do the  $\chi^2$ -test, and the below testing statistic has approximately a  $\chi_4^2$  distribution.

$$Q = \sum_{i=1}^5 \frac{(N_i - np_i^0)^2}{np_i^0} = 27.50.$$

Therefore the  $p$ -value is  $1 - F_{\chi_4^2}(27.5) = 1.575 \times 10^{-5}$ , and one rejects  $H_0$  so long as the level  $\alpha$  is greater than  $1.575 \times 10^{-5}$ .

**Solution 2:** Exercise 10.3.4 (page 645)

We carry out the test of independence. The contingency table is given below

	Wears a moustache	Does not wear a moustache	Total
Between 18 and 30	12	28	40
Over 30	8	52	60
Total	20	80	100

The MEL's  $\hat{m}_{ij} = \frac{X_i + X_{+j}}{n}$  are listed below.

$\hat{m}_{ij}$	Wear a moustache	Does not wear a moustache
Between 18 and 30	8	32
Over 30	12	48

So the observed  $\chi^2$  testing statistic is  $Q = 25/6$ . The degree of freedom is 1, and the  $p$ -value is  $1 - F_{\chi^2_1}(25/6) = 0.04123$ . Therefore when the level  $\alpha$  is greater than 0.04123 we reject  $H_0$ .

**Solution 3:** Exercise 10.4.5 (page 652)

The correct table to be analyzed is as follows:

Supplier	Defectives	Nondefectives
1	1	14
2	7	8
3	7	8

The value of  $Q$  found from this table is 7.2. If  $Q$  has the  $\chi^2$  distribution with  $(3 - 1)(2 - 1) = 2$  degrees of freedom, then  $\Pr(Q \geq 7.2) = 0.027 < 0.05$ .

**Solution 4:**

For Problem 3 in HW10, we have the rejection region  $\mathcal{R}(K) = \{\Lambda > K\}$ , where

$$\Lambda = \frac{\max_{\Theta} L}{\max_{\Theta_0} L} = \prod_{i=1}^m \left( \frac{X_i \sum n_j}{n_i \sum X_j} \right)^{X_i} \left( \frac{(n_i - X_i) \sum n_j}{n_i \sum (n_j - X_j)} \right)^{n_i - X_i}. \quad (1)$$

We know that under “smoothness” conditions, when the sample size is large, the LR statistic  $2 \log \Lambda \sim_{\text{approx}} \chi_h^2$ . The degrees of freedom is  $h = p - q$ , where  $p$  is the number of free parameters estimated in computing the global MLE, and  $q$  is the number of free parameters estimated in computing the restricted MLE. In this case,

$$\begin{aligned} \Theta_0 &= \{(p_1, \dots, p_m) | 0 \leq p_1 = \dots = p_m \leq 1\}, \\ \Theta &= \{(p_1, \dots, p_m) | 0 \leq p_1, \dots, p_m \leq 1\}. \end{aligned}$$

So we have  $h = m - 1$ , thus  $2 \log \Lambda \sim_{\text{approx}} \chi_{m-1}^2$ . The corresponding  $p$ -value of this generalized likelihood ratio test is  $1 - F_{\chi_{m-1}^2}(2 \log \Lambda)$ , here  $\Lambda$  is calculated based on (1) from the observed data.

For Problem 4 in HW10, we have the rejection region  $\mathcal{R}(K) = \{\Lambda > K\}$ , where

$$\Lambda = \frac{\max_{\theta \in \Theta} L}{\max_{\theta \in \Theta_0} L} = \frac{L(\bar{X}, \bar{Y}, \bar{W}, \hat{\sigma}_X, \hat{\sigma}_Y, \hat{\sigma}_W)}{L(\bar{X}, \bar{Y}, \bar{W}, \hat{\sigma}, \hat{\sigma}, \hat{\sigma})}. \quad (2)$$

where  $\hat{\sigma}$  is defined in

$$\hat{\sigma}^2 = \frac{\sum_{i=1}^{n_1} (X_i - \bar{X})^2 + \sum_{i=1}^{n_2} (Y_i - \bar{Y})^2 + \sum_{i=1}^{n_3} (W_i - \bar{W})^2}{n_1 + n_2 + n_3}.$$

$\hat{\sigma}_X$  is defined in

$$\hat{\sigma}_X^2 = \frac{1}{n_1} \sum_{i=1}^{n_1} (X_i - \bar{X})^2.$$

and  $\hat{\sigma}_Y$  and  $\hat{\sigma}_W$  are defined similarly as  $\hat{\sigma}_X$ . In this case,

$$\Theta_0 = \{\theta = (\mu_X, \mu_Y, \mu_W, \sigma_X, \sigma_Y, \sigma_W) | 0 < \sigma_X = \sigma_Y = \sigma_W < \infty\},$$

$$\Theta = \{\theta = (\mu_X, \mu_Y, \mu_W, \sigma_X, \sigma_Y, \sigma_W) | 0 < \sigma_X, \sigma_Y, \sigma_W < \infty\}.$$

So we have  $h = 3 - 1 = 2$ , thus  $2 \log \Lambda \sim_{approx} \chi_2^2$ . The corresponding  $p$ -value of this generalized likelihood ratio test is  $1 - F_{\chi_2^2}(2 \log \Lambda)$ , here  $\Lambda$  is calculated based on (2) from the observed data.

### Solution 5:

(a). When  $\mu_X = \mu_Y$ ,  $T$  has a  $t$ -distribution with freedom  $n + m - 2$ .

(b). Denote  $\sigma^2$  the variance of both  $X_i$ 's and  $Y_j$ 's. Let

$$Z = \frac{\bar{X} - \bar{Y} - \mu_X + \mu_Y}{\sigma \cdot \sqrt{\frac{1}{m} + \frac{1}{n}}}, \quad W = \frac{(n-1)s_X^2 + (m-1)s_Y^2}{\sigma^2}.$$

We have learned that  $Z \sim \mathcal{N}(0, 1)$ ,  $W \sim \chi_{n+m-2}^2$  and that  $Z$  and  $W$  are independent, therefore

$$\frac{\bar{X} - \bar{Y} - \mu_X + \mu_Y}{s_{pooled} \sqrt{\frac{1}{m} + \frac{1}{n}}} = \frac{Z}{\sqrt{W/(m+n-2)}} \sim t_{n+m-2}.$$

(c). Since with confidence  $1 - \alpha$ ,

$$F_{t_{n+m-2}}^{-1} \left( \frac{\alpha}{2} \right) \leq \frac{\bar{X} - \bar{Y} - \mu_X + \mu_Y}{s_{pooled} \sqrt{\frac{1}{m} + \frac{1}{n}}} \leq F_{t_{n+m-2}}^{-1} \left( 1 - \frac{\alpha}{2} \right),$$

we solve the inequalities to give the confidence interval,

$$\bar{X} - \bar{Y} - F_{t_{n+m-2}}^{-1} \left( 1 - \frac{\alpha}{2} \right) s_{pooled} \sqrt{\frac{1}{m} + \frac{1}{n}} \leq \mu_X - \mu_Y \leq \bar{X} - \bar{Y} - F_{t_{n+m-2}}^{-1} \left( \frac{\alpha}{2} \right) s_{pooled} \sqrt{\frac{1}{m} + \frac{1}{n}}.$$

(d). For testing the hypothesis  $H_0 : \mu_X = \mu_Y$  against  $H_1 : \mu_X \neq \mu_Y$ , with level  $\alpha$  the rejection region could be chosen as

$$\mathcal{R}(\alpha) = \left\{ |T| > F_{t_{n+m-2}}^{-1} \left( 1 - \frac{\alpha}{2} \right) \right\}.$$

One has

$$\begin{aligned}
T \in \mathcal{R}(\alpha) &\iff T > F_{t_{m+n-2}}^{-1} \left(1 - \frac{\alpha}{2}\right) \text{ or } T < -F_{t_{m+n-2}}^{-1} \left(1 - \frac{\alpha}{2}\right) = F_{t_{m+n-2}}^{-1} \left(\frac{\alpha}{2}\right) \\
&\iff 0 < \bar{X} - \bar{Y} - F_{t_{n+m-2}}^{-1} \left(1 - \frac{\alpha}{2}\right) s_{pooled} \sqrt{\frac{1}{m} + \frac{1}{n}}, \text{ or} \\
&\quad \bar{X} - \bar{Y} - F_{t_{n+m-2}}^{-1} \left(\frac{\alpha}{2}\right) s_{pooled} \sqrt{\frac{1}{m} + \frac{1}{n}} < 0 \\
&\iff 0 \text{ is not in the CI found in (c).}
\end{aligned}$$

### Solution 6:

$H_0$ : death rate (death/day) is expected to be equal for each month  $H_A$ : death rate (death/day) is not equal for each month expected death/day =  $\frac{\text{total number of death}}{365} = \frac{1668+1407+\dots+1526}{365} = 46.3452$

Month	Day Number	Exp Death	$x_i - m_i$
Jan	31	1436.70137	231.2986
Feb	28	1297.6658	109.34247
Mar	31	1436.70137	-66.7014
Apr	30	1390.35616	-81.3562
May	31	1436.70137	-95.7014
Jun	30	1390.35616	-52.35616
Jul	31	1436.70137	-30.70137
Aug	31	1436.70137	9.2986
Sep	30	1390.35616	-58.35616
Oct	31	1436.70137	-73.70137
Nov	30	1390.35616	19.648356
Dec	31	1436.70137	89.29863

$Q = \sum_{i=1}^m \frac{(x_i - m_i)^2}{m_i} = 75.4273$  where d.f = 12-1=11 and  $Q \sim \chi^2(11)$ . Note that here  $p = 12$  (12 different incidence rates) and  $q = 1$  (a common incidence rate), and so  $p - q = 11$ . Therefore,

$P(Q \geq 75.4273) = 1.122 * 10^{-11}$ , since the p-value is extremely small, we will reject the null

hypothesis. We can see a seasoned pattern in the death rate by looking at  $x_i - m_i$ , which shows that more deaths occur in the death winter.

### Solution 7:

$H_0$ : probability of breaking at each point  $p_i$  is equal:  $p_1 = \dots = p_5 = p$ .

$x_i = P(\text{bar breaks in } i \text{ places}) = \binom{5}{i} p^i (1-p)^{5-i}, 0 \leq i \leq 5$ .  $N_i$ =number of bar breaks at i.

$$L(p) = \frac{280!}{N_0!N_1!N_2!N_3!N_4!N_5!} X_0^{N_0} X_1^{N_1} X_2^{N_2} X_3^{N_3} X_4^{N_4} X_5^{N_5}$$

$$\begin{aligned}
\log(L(p)) &= \log 280! - \sum_{i=0}^5 \log N_i! + \sum_{i=0}^5 N_i \log X_i = \\
&\log 280! - \sum_{i=0}^5 \log N_i! + \sum_{i=0}^5 N_i (\log \binom{5}{i} + i \log p + (5-i) \log(1-p)) \\
\frac{d}{dp} \log L(p) &= \sum_{i=0}^5 N_i \left[ \frac{i}{p} - \frac{5-i}{1-p} \right] \\
\frac{d^2}{dp^2} \log L(p) &= \sum_{i=0}^5 N_i \left( -\frac{i}{p^2} - \frac{5-i}{(1-p)^2} \right) < 0 \\
\text{Then given } \frac{d}{dp} \log L(p) &= \sum_{i=0}^5 N_i \left[ \frac{i}{p} - \frac{5-i}{1-p} \right] = 0, \text{ the MLE is:}
\end{aligned}$$

$$\hat{p} = \frac{\sum_{i=0}^5 i N_i}{5 * 280} = 0.142$$

Then the table of the MLE of the expected counts is:

Breaks/Bar	Expected Frequency
0	130.2
1	107.7
2	35.7
3	5.9
4	0.49
5	0.02

We need to pool 3,4,5 to have at least 5 expected counts in each category and the table of the observed count with pooling is:

Breaks/Bar	Observed	Expected
0	157	130.2
1	69	107.7
2	35	35.7
3,4 or 5	19	6.4

$$Q = \frac{(157-130.1)^2}{130.1} + \frac{(69-107.8)^2}{107.8} + \frac{(35-35.7)^2}{35.7} + \frac{(19-6.4)^2}{6.4} = 44.3$$

$Q \sim \chi^2$  distribution with 3-1=2 d.f. Note that after the pooling there are a total of 4 categories and so the number of parameters in the unrestricted case is  $p = 4 - 1 = 3$ , while under the null, the number of free parameters is  $q = 1$ .

$P(Q \geq 44.3) = 1 - F(44.3)$  where  $F$  is the cdf of  $\chi^2$  with 2 d.f.

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