

An Outline of Solution to STAT 545 Midterm Exam - 10/15

1. (30 points)

(a) (15 points)

- (5 points)

$$OR = 2 * 8 / (3 * 3) \text{ (3 points)}$$

Interpretation: The odds of Y 's falling in category 1 at level 1 of X is $(2*8/(3*3))$ times that at level 2 of X ; (2 points)

- (5 points)

$$\text{Diff in prop} = 2/5 - 3/11 \text{ (3 points)}$$

Interpretation: The proportion of being in category 1 of Y at level 1 of X is $(2/5-3/11)$ more than that at level 2 of X , or the difference in the proportion of being in category 1 of Y between levels 1 and 2 of X is $(2/5-3/11)$; (2 points)

- (5 points)

$$RR = (2/5)/(3/11) \text{ (3 points)}$$

Interpretation: The proportion of being in category 1 of Y at level 1 of X is $((2/5)/(3/11))$ times that at level 2 of X ; (2 points)

(b) (15 points)

- (5 points)

$$\log OR = \log(2 * 8 / (3 * 3)), \text{ (1 point)}$$

$$SE(\log OR) = \sqrt{1/2 + 1/3 + 1/3 + 1/8}, \text{ (1 point)}$$

$$\log OR \pm 1.96 * SE(\log OR), \text{ (2 points)}$$

$$95\%CL : (\exp(LL), \exp(RL)). \text{ (1 point)}$$

- (5 points)

$$\text{Diff in prop} = 2/5 - 3/11, \text{ (1 point)}$$

$$SE(\text{Diff in prop}) = \sqrt{2/5(1 - 2/5)/5 + 3/11(1 - 3/11)/11}, \text{ (2 points)}$$

$$\text{Diff in prop} \pm 1.96 * SE(\text{Diff in prop}). \text{ (2 points)}$$

- (5 points)

$\log RR = \log(2/5/(3/11))$, (1 point)

$SE(\log RR) = \sqrt{1/2 - 1/5 + 1/3 - 1/11}$, (1 point)

$LL = \log(RR) - 1.96 * SE(\log(RR))$, $RL = \log(RR) + 1.96 * SE(\log(RR))$,
(2 points)

95%CL : (exp LL, exp RL). (1 point)

2. (30 points)

(a) (14 points) Pearson χ^2 statistic:

$$X^2 = \sum_i \sum_j (n_{ij} - \hat{m}_{ij})^2 / \hat{m}_{ij},$$

where $\hat{m}_{ij} = n_{i+}n_{+j}/n$ with $n_{i+} = \sum_k n_{ik}$ and $n_{+j} = \sum_k n_{kj}$, $i = 1, 2$, $j = 1, 2$.

(5 points) $DF = 1$. (2 points)

Likelihood ratio χ^2 statistics:

$$G^2 = 2 \sum_i \sum_j n_{ij} \log(n_{ij}/\hat{m}_{ij}),$$

where \hat{m}_{ij} are defined the same as above. (5 points) $Df = 1$. (2 points)

(b) (16 points) Should not use the test statistics from (a). (2 points)

Reason: Pearson and likelihood ratio χ^2 tests are based on large samples, while the data given are from a small sample. The Pearson and likelihood ratio tests may not be valid. (4 points)

An alternative: Fisher's exact test. Hypergeometric distribution for n_{11} conditional on both row and column totals:

$$\frac{\binom{n_{1+}}{n_{11}} \binom{n_{2+}}{n_{+1} - n_{11}}}{\binom{n}{n_{+1}}},$$

where $n_{11} \in [m_-, m_+]$ with $m_- = \max(0, n_{1+} + n_{+1} - n)$ and $m_+ = \min(n_{1+}, n_{+1})$.

(6 points)

P-value:

$$\frac{\binom{4}{1} \binom{4}{3}}{\binom{8}{4}} + \frac{\binom{4}{0} \binom{4}{4}}{\binom{8}{4}}$$

(4 points)

3. (20 points)

(a) (10 points)

- (8 points) $L(\theta) = \theta^{2n_{11}} \theta^{n_{12}+n_{21}} (1-\theta)^{n_{12}+n_{21}} (1-\theta)^{2n_{22}}$. Solve $L'(\theta) = 0$ for $\theta \implies \hat{\theta} = (p_{1+} + p_{+1})/2$.
- (2 points) Show $\hat{\theta}$ actually maximizes the likelihood, e.g., by showing the second derivative at $\hat{\theta} < 0$.

(b) (10 points)

- (6 points) $X^2 = \sum_i \sum_j (n_{ij} - \hat{m}_{ij})^2 / \hat{m}_{ij}$ with $\hat{m}_{11} = n\hat{\theta}^2$, $\hat{m}_{12} = \hat{m}_{21} = n\hat{\theta}(1-\hat{\theta})$, and $\hat{m}_{22} = n(1-\hat{\theta})^2$.
- (4 points) $DF = 2$.

4. (20 points)

(a) (10 points) Let $\sum_{y=1}^{\infty} r(\theta) \exp(\theta y) / y! = 1$. (2 points)

We have $r(\theta) = \left\{ \sum_{y=1}^{\infty} \exp(\theta y) / y! \right\}^{-1} = \left\{ \sum_{y=1}^{\infty} (\exp(\theta))^y / y! \right\}^{-1}$ (3 points) = $\{\exp(\exp(\theta)) - 1\}^{-1}$. (5 points)

(b) (10 points) $f_Y(y; \theta)$ is an exponential-family distribution with $\theta = \theta$, $b(\theta) = \log(\exp(\exp(\theta)) - 1)$, $\phi = a(\phi) = 1$, $c(y, \phi) = \log y!$. (4 points)

The kernel of the log-likelihood function for a sample of i.i.d. observations Y_1, \dots, Y_n is $\theta \sum_{i=1}^n Y_i - n \log(\exp(\exp(\theta)) - 1)$. (2 points)

The likelihood equation therefore is $d \log(\exp(\exp(\theta)) - 1) / d\theta = \bar{Y}$, where $\bar{Y} = \sum_{i=1}^n Y_i / n$. (2 points)

After simplification, we get

$$\frac{\exp(\theta)}{1 - \exp(-\exp(\theta))} = \bar{Y}. \text{ (2 points)}$$