

STAT2602 Mid-term Class Test

NOT TO BE TAKEN AWAY

Fall, 2024

[Total: 50 marks]

Time: 10:30am - 12:10pm

1. Suppose that a random variable X has a p.d.f given by

$$f(x) = \begin{cases} 3e^{-3x}, & x > 0; \\ 0, & \text{otherwise.} \end{cases}$$

- (i) Calculate the moment generating function of X . **[5 marks]**
- (ii) Calculate $E(X)$ and $E(X^2)$. **[2 marks]**
- (iii) Calculate $E(e^{X/2})$, $E(e^X)$, and $E(e^{4X})$, if exists. **[3 marks]**

[Total: 10 marks]

2. The number of male and female customers who visited a company on n successive days were recorded as $(X_1, Y_1), (X_2, Y_2), \dots, (X_n, Y_n)$, such that there were X_i male and Y_i female visitors on the i -th day. Assume that for each i , X_i and Y_i follow Poisson distributions with means λ and $\beta\lambda$ respectively, for $\lambda, \beta > 0$, and that the $2n$ observations are independent.

It is known that the Poisson distribution with mean $\theta > 0$ has the p.d.f (or p.m.f.) function

$$f(x|\theta) = \frac{e^{-\theta}\theta^x}{x!}, \quad x = 0, 1, 2, \dots$$

- (i) Show that the likelihood function of (λ, β) based on the observations $(X_1, Y_1), (X_2, Y_2), \dots, (X_n, Y_n)$ is

$$L(\lambda, \beta) = \frac{e^{-n(1+\beta)\lambda} \beta^{\sum_{i=1}^n Y_i} \lambda^{\sum_{i=1}^n X_i + \sum_{i=1}^n Y_i}}{\prod_{i=1}^n (X_i! Y_i!)}.$$

[5 marks]

- (ii) If it is known that $\beta = 1$, show that the statistic

$$T = \sum_{i=1}^n (X_i + Y_i)$$

is sufficient for λ .

[5 marks]

- (iii) If both λ and β are unknown, show that the statistic

$$T = \left(\sum_{i=1}^n X_i, \sum_{i=1}^n Y_i \right)$$

is sufficient for (λ, β) . [5 marks]

- (iv) From now on suppose that $\beta = \lambda$, but the value of λ is unknown.

- (a) Show that the statistic

$$T = \sum_{i=1}^n (X_i + 2Y_i)$$

is sufficient for λ . [5 marks]

- (b) Show that the Fisher information based on the $2n$ observations is

$$I(\lambda) = n \left(4 + \frac{1}{\lambda} \right).$$

[5 marks]

- (c) Show that the maximum likelihood estimator of λ is

$$\hat{\lambda} = \sqrt{\frac{1}{16} + \frac{T}{2n}} - \frac{1}{4}, \quad \text{where} \quad T = \sum_{i=1}^n (X_i + 2Y_i)$$

[5 marks]

- (d) If n is large, the maximum likelihood estimator $\hat{\lambda}$ has an approximately normal distribution. State the mean and variance of this normal distribution. [5 marks]

- (e) It is observed that the total numbers of male and female visitors on $n = 20$ successive days are

$$\sum_{i=1}^{20} X_i = 40, \quad \text{and} \quad \sum_{i=1}^{20} Y_i = 10$$

respectively. Calculate the maximum likelihood estimate of λ and its standard error. [5 marks]

[Total: 40 marks]

A LIST OF STATISTICAL FORMULAE

1. $M_X(t) = E(e^{tX})$. $E(X^r) = \left(\frac{d^r}{dt^r} M_X(t) \right) \Big|_{t=0}$.

2. For $X \sim \text{Uniform}(a, b)$,

$$f(x) = \frac{x}{b-a}$$

3. For $X \sim N(\mu, \sigma^2)$,

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right), \quad M_X(t) = \exp\left(\mu t + \frac{1}{2}\sigma^2 t^2\right).$$

4. $\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$. $S^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2$.

5. $\text{Bias}(\hat{\theta}) = E(\hat{\theta}) - \theta$. $E[(\hat{\theta} - \theta)^2] = \text{Var}(\hat{\theta}) + [\text{Bias}(\hat{\theta})]^2$.

6. $I(\theta) = E\left[\left(\frac{\partial \ln f(X; \theta)}{\partial \theta}\right)^2\right] = E\left[-\frac{\partial^2 \ln f(X; \theta)}{\partial \theta^2}\right]$. $\text{Var}(\hat{\theta}) \geq \frac{1}{nI(\theta)}$.

7. $I_n(\theta) = E\left[\left(\frac{\partial \ell(\theta)}{\partial \theta}\right)^2\right] = E\left[-\frac{\partial^2 \ell(\theta)}{\partial \theta^2}\right] = nI(\theta)$, where $\ell(\theta) = \ln L(\theta)$,

$L(\theta) = \mathbf{f}(X_1, X_2, \dots, X_n; \theta)$ for $\theta \in \Omega$, and $\mathbf{f}(x_1, x_2, \dots, x_n; \theta)$ is the joint p.d.f. for the random sample X_1, X_2, \dots, X_n .

8. Factorization: $\mathbf{f}(x_1, \dots, x_n; \theta) = g(T(x_1, \dots, x_n); \theta)h(x_1, \dots, x_n)$.

9. Exponential family: $f(x; \theta) = h(x)c(\theta) \exp\left(\sum_{i=1}^s p_i(\theta)t_i(x)\right)$.

10. CLT: $\frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \approx N(0, 1)$ for large n .

11. MLE: $\frac{\hat{\theta} - \theta}{\sqrt{1/I_n(\theta)}} \approx N(0, 1)$ for large n .

***** END OF PAPER *****