

# Solutions to Midterm

**Bonus:** (4 marks) Suppose  $X_1, \dots, X_n$  are iid and follow the Normal distribution  $N(\mu, 1)$  with unknown  $\mu$ . Show the distributions of  $\bar{X} = \sum_{i=1}^n X_i/n$  and  $\sum_{i=1}^n (X_i - \bar{X})^2$ .

The dist. of  $\bar{X}$  is normal (1 mark)  $N(\mu, \frac{1}{n})$  (1 mark)

The dist. of  $\sum_{i=1}^n (X_i - \bar{X})^2$  is chi square (1 mark)  $\chi^2(n-1)$  (1 mark)

**Question 1:** (5 marks) Consider  $X_1, \dots, X_n$  are iid Gamma distribution  $G(\alpha/2, 1/2)$ :

$$f(x|\alpha, \lambda) = \frac{1}{\Gamma(\alpha/2)} \frac{1}{2^{\alpha/2}} x^{\alpha/2-1} e^{-x/2}, \quad 0 \leq x < \infty.$$

a. (2 marks) Find the moment estimate of  $\alpha$ .

$$E(X) = \frac{\alpha/2}{1/2} = \alpha \quad (1 \text{ mark})$$

$$\hat{\alpha} = \bar{X} \quad (1 \text{ mark})$$

b. (3 marks) Find the sample distribution of the moment estimate by central limit theorem.

$$\text{Var}(X) = 2\alpha \quad (1 \text{ mark})$$

$$\sqrt{n} \cdot \frac{\hat{\alpha} - \alpha}{\sqrt{2\alpha}} \rightarrow N(0, 1) \quad (2 \text{ marks})$$

Question 2: (5 marks) Consider  $X_1, \dots, X_n$  are iid Normal distribution  $N(0, \sigma^2)$ .

a. (1 mark) Compute the MLE estimate of  $\sigma^2$ .

$$L(\sigma^2) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{X_i^2}{2\sigma^2}\right\}$$

$$\ell(\sigma^2) = \log(L(\sigma^2))$$

$$\frac{\partial \ell(\sigma^2)}{\partial \sigma^2} = 0 \Rightarrow \hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n X_i^2$$

(0.5 mark)

(Detail needed)

b. (2 marks) What's the distribution of the MLE estimate by the Exact method?

chi square distribution (1 mark)

$$\frac{n \hat{\sigma}^2}{\sigma^2} \sim \chi^2(n)$$

(1 mark)

c. (2 marks) What's the distribution of the MLE estimate by the Large sample theory?

Fisher information for  $\sigma^2$ :  $I(\sigma^2) = \frac{1}{2\sigma^4}$  (1 mark)

(Detail needed)

$$\sqrt{n I(\hat{\sigma}^2)} (\hat{\sigma}^2 - \sigma^2) \rightarrow N(0, 1) \quad (1 \text{ mark})$$

Question 3: (5 marks) Consider the model

$$X_1, \dots, X_n | \theta \sim \text{iid Normal}, N(\theta, 1),$$

$$\Theta \sim N(2, 2)$$

a. (3 marks) Find the posterior pdf of  $\Theta$ . Justify your answer.

Normal distribution (1 mark)

Posterior dist.  $N\left(\bar{x} \cdot \frac{2}{2 + 1/n} + 2 \cdot \frac{1/n}{2 + 1/n}, \frac{2/n}{2 + 1/n}\right)$ .

$\downarrow$  (1 mark)                       $\downarrow$  (1 mark)  
 (Detail needed)

b. (2 marks) Suppose a sample size  $n = 2$  results in the observations  $X_1 = 0$  and  $X_2 = 2$ . Given these data and square loss, find the Bayes' estimate of  $\theta$ .

Posterior mean (1 mark)

$$\hat{\theta} = \frac{0+2}{2} \frac{2}{2 + 1/2} + 2 \frac{1/2}{2 + 1/2}$$

$$= 1 \times \frac{4}{5} + 2 \times \frac{1}{5}$$

$$= \frac{6}{5} = 1.2 \quad (1 \text{ mark})$$

Question 4: (5 marks) Consider  $X_1, \dots, X_n$  are iid Poisson distribution  $Pois(\lambda)$  with pdf

$$P(X = x|\lambda) = \lambda^x e^{-\lambda} / x!.$$

a. (3 marks) Find the Fisher information. Justify your answer.

$$\ell(\lambda) = \log L(\lambda) = x \log \lambda - \lambda - \log x! \quad (1 \text{ mark})$$

$$\ell''(\lambda) = -\frac{x}{\lambda^2} \quad (1 \text{ mark})$$

$$\text{Fisher info } I(\lambda) = -E(\ell''(\lambda)) = \frac{E(x)}{\lambda^2} = \frac{\lambda}{\lambda^2} = \frac{1}{\lambda} \quad (1 \text{ mark})$$

(Detail needed)

b. (2 marks) Find the efficient estimate of  $\lambda$ .

$$E(\bar{X}) = \lambda \quad 0.5 \text{ mark}$$

$$\text{Var}(\bar{X}) = \frac{\text{Var}(X)}{n} = \frac{\lambda}{n} = \frac{1}{nI(\lambda)} \quad 0.5 \text{ mark}$$

$$\Rightarrow \bar{X} \text{ is efficient} \quad 1 \text{ mark}$$

**Question 5:** (5 marks) Consider  $X_1, \dots, X_n$  are iid Exponential distribution  $\text{Exp}(\lambda)$  with pdf

$$f(x|\lambda) = \lambda e^{-\lambda x} \text{ if } x \geq 0, \text{ otherwise } 0.$$

a. (2 marks) Determine the MLE of  $\lambda$ .

$$\ell(\lambda) = n \log \lambda - \lambda \sum_{i=1}^n x_i \quad (0.5 \text{ mark})$$

$$\ell'(\lambda) = \frac{n}{\lambda} - \sum_{i=1}^n x_i = 0 \quad (0.5 \text{ mark})$$

$$\Rightarrow \hat{\lambda} = \frac{n}{\sum_{i=1}^n x_i} = \frac{1}{\bar{x}} \quad (1 \text{ mark})$$

(Detail needed)

b. (3 marks) Is the MLE a consistent estimate of  $\lambda$ ? Justify your answer.

MLE is a consistent estimate of  $\lambda$  by Large Sample Theory. (3 marks).  
 OR:  $\bar{x}$  is a consistent estimate of  $E(X) = 1/\lambda$

by the law of large number. (1 mark)

So  $\bar{x} \rightarrow 1/\lambda$  in prob. We prove  $\frac{1}{\bar{x}} \rightarrow \frac{1}{1/\lambda}$  in prob.

$\forall \epsilon > 0$

$$\begin{aligned} P\left(\left|\frac{1}{\bar{x}} - \frac{1}{1/\lambda}\right| > \epsilon\right) &= P\left(\left|\frac{\bar{x} - 1/\lambda}{\bar{x} \cdot 1/\lambda}\right| > \epsilon\right) \\ &= P(|\bar{x} - 1/\lambda| > \epsilon |\bar{x}|/\lambda) = P(|\bar{x} - 1/\lambda| > \epsilon |\bar{x}|/\lambda, |\bar{x} - 1/\lambda| < \frac{1}{2\lambda}) \\ &\quad + P(|\bar{x} - 1/\lambda| > \epsilon |\bar{x}|/\lambda, |\bar{x} - 1/\lambda| \geq \frac{1}{2\lambda}) \end{aligned}$$

$$\leq P(|\bar{x} - 1/\lambda| > \epsilon \cdot \frac{1}{2\lambda^2}) + P(|\bar{x} - 1/\lambda| \geq \frac{1}{2\lambda})$$

$\rightarrow 0$  by  $\bar{x} \rightarrow 1/\lambda$  in prob. (1 mark)  
 (Detail needed)

MLE  $\hat{\lambda}$  is a consistent estimate of  $\lambda$ . (1 mark)

Question 6: (5 marks) Consider  $X_1, \dots, X_n$  are iid Uniform distribution  $U(-\theta, 0)$  with pdf

$$f(x|\theta) = 1/\theta \text{ if } -\theta \leq x \leq 0, \text{ otherwise } 0, \text{ where } 0 < \theta < \infty.$$

a. (2 marks) Determine the sufficient statistic for  $\theta$ .

$$L(\theta) = \prod_{i=1}^n f(x_i|\theta) = \prod_{i=1}^n \frac{1}{\theta} \mathbb{1}(-\theta, x_i) \quad (0.5 \text{ mark})$$

where  $\mathbb{1}(a, b) = 1$  if  $a \leq b$ , otherwise 0.

$$\Rightarrow L(\theta) = \frac{1}{\theta^n} \mathbb{1}(-\theta, \min\{x_1, x_2, \dots, x_n\}) \quad (0.5 \text{ mark})$$

(Detail needed)

$\Rightarrow \min\{x_1, \dots, x_n\}$  is the sufficient statistic for  $\theta$ . (1 mark)

b. (3 marks) Find an unbiased estimate of  $\theta$  based on the sufficient statistic.

Define  $X_{(1)} = \min\{x_1, \dots, x_n\}$

$$\begin{aligned} P(X_{(1)} \leq x) &= 1 - P(X_{(1)} > x) = 1 - P(x_1 > x, \dots, x_n > x) \\ &= 1 - \prod_{i=1}^n P(x_i > x) = 1 - \left(\frac{-x}{\theta}\right)^n \end{aligned}$$

$$\Rightarrow \text{pdf of } X_{(1)} \text{ is } \frac{n(-x)^{n-1}}{\theta^n} \quad (1 \text{ mark})$$

$$\begin{aligned} E(X_{(1)}) &= \int_{-\theta}^0 x \cdot \frac{n(-x)^{n-1}}{\theta^n} dx = \int_{-\theta}^0 (-x) \frac{n(-x)^{n-1}}{\theta^n} d(-x) \\ &\stackrel{t=-x}{=} \int_0^{\theta} t \frac{nt}{\theta^n} dt = - \int_0^{\theta} \frac{nt}{\theta^n} dt = - \frac{n}{\theta^n} \frac{t^{n+1}}{n+1} \Big|_0^{\theta} = - \frac{n\theta}{n+1} \end{aligned}$$

(1 mark)  
(Detail needed)

$\Rightarrow$  The unbiased estimate of  $\theta$  based on the sufficient stat.

$$\text{is } - \frac{(n+1)}{n} \min\{x_1, \dots, x_n\} \quad (1 \text{ mark})$$