# STATS100B – Introduction to Mathematical Statistics Homework 7

Feng Shiwei UID:305256428

#### Question a

Solution:

$$L = \frac{1}{(2\pi)^{\frac{n}{2}}} |\mathbf{\Sigma}|^{-\frac{1}{2}} e^{\frac{1}{2}(\mathbf{Y} - \mu \mathbf{1})' \mathbf{\Sigma}^{-1}(\mathbf{Y} - \mu \mathbf{1})}$$

$$lnL = -\frac{n}{2} ln(2\pi \mathbf{\Sigma}^{2}) - \frac{1}{2} ln|\mathbf{V}| - \frac{1}{2\sigma^{2}} (\mathbf{Y} - \mu) V^{-1} (\mathbf{Y} - \mu)$$

$$\frac{\partial \ln L}{\partial \mu} = -\frac{1}{2\sigma^{2}} \left[ -\mathbf{Y}' \mathbf{V}^{-1} \mathbf{1} - \mathbf{1}' \mathbf{V}^{-1} \mathbf{Y} + 2\mu \mathbf{1}' \mathbf{V}^{-1} \mathbf{1} \right] = 0$$

$$\therefore \hat{\mu} = \frac{\mathbf{1}' \mathbf{V}^{-1} \mathbf{Y}}{\mathbf{1}' \mathbf{V}^{-1} \mathbf{1}}$$

$$\frac{\partial \ln L}{\partial \sigma^{2}} = -\frac{n}{2\sigma^{2}} + \frac{1}{\sigma^{4}} (\mathbf{Y} - \mu \mathbf{1})' \mathbf{V}^{-1} (\mathbf{Y} - \mu \mathbf{1}) = 0$$

$$\therefore \hat{\sigma^{2}} = \frac{(\mathbf{Y} - \hat{\mu} \mathbf{1})' \mathbf{V}^{-1} (\mathbf{Y} - \hat{\mu} \mathbf{1})}{n}$$

## Question b

Solution:

$$E(\hat{\mu}) = E\left(\frac{1'V^{-1}\boldsymbol{Y}}{1'V^{-1}1}\right) = \frac{1'V^{-1}E(\boldsymbol{Y})}{1'V^{-1}1} = \frac{1'V^{-1}\mu 1}{1'V^{-1}1} = \mu$$

## Question c

Solution:

$$\begin{split} \boldsymbol{I}(\boldsymbol{\theta}) &= -E \begin{pmatrix} \frac{\partial^2 \ln L}{\partial \mu^2} & \frac{\partial^2 \ln L}{\partial \mu \partial \sigma^2} \\ \\ \frac{\partial^2 \ln L}{\partial \sigma^2 \partial \mu} & \frac{\partial \ln L}{\partial \sigma^{2(2)}} \end{pmatrix} \\ & \frac{\partial^2 \ln L}{\partial \mu^2} = -\frac{1'V^{-1}1}{\sigma^2} \\ \frac{\partial^2 \ln L}{\partial \mu \partial \sigma^2} &= \frac{1}{\sigma^4} \left[ \mu 1'V^{-1}1 - 1'V^{-1}\boldsymbol{Y} \right] \end{split}$$

$$\frac{\partial^{2} \ln L}{\partial \sigma^{2} \partial \mu} = \frac{1}{\sigma^{4}} \left[ \mu \mathbf{1}' V^{-1} \mathbf{1} - \mathbf{1}' V^{-1} \mathbf{Y} \right]$$
$$\frac{\partial \ln L}{\partial \sigma^{2}(2)} = \frac{n}{2\sigma^{4}} - \frac{1}{\sigma^{6}} \left( Y - \mu \mathbf{1} \right)' V^{-1} \left( \mathbf{Y} - \mu \mathbf{1} \right)$$
$$\therefore \mathbf{I}(\boldsymbol{\theta}) \Rightarrow ($$

#### Question d

Solution:

$$f(y_i) = \frac{1}{\sqrt{2\pi}i\sigma} e^{-\frac{(y-i\theta)^2}{2(i\sigma)^2}}$$

$$L = \prod_{i=1}^n f(y_i) = (2\pi)^{-\frac{n}{2}} \prod_{i=1}^n i\sigma^{-n} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n \left(\frac{y_i-i\theta}{i}\right)^2}$$

$$\ln L = -\frac{n}{2} \ln 2\pi + \sum_{i=1}^n \ln i - n \ln \sigma - \frac{1}{2\sigma^2} \sum_{i=1}^n \left(\theta - \frac{y_i}{i}\right)^2$$

$$\frac{\partial \ln L}{\partial \theta} = -\frac{1}{\sigma^2} \sum_{i=1}^n \left(\theta - \frac{y_i}{i}\right) = 0$$

$$\therefore \hat{\theta} = \frac{1}{n} \sum_{i=1}^n \frac{y_i}{i}$$

$$var\left(\hat{\theta}\right) = \frac{1}{n^2} \sum \frac{var(y_i)}{i^2} = \frac{1}{n^2} \sum_{i=1}^n \frac{i^2\sigma^2}{i^2} = \frac{\sigma^2}{n}$$

$$\because \frac{\partial^2 \ln L}{\partial \theta^2} = -\frac{n}{\sigma^2}$$

$$var\left(\hat{\theta}\right) \ge -\frac{1}{E\left(\frac{\partial^2 \ln L}{\partial \theta^2}\right)} = \frac{\sigma^2}{n}$$

 $\therefore \hat{\theta}$  is an efficient estimator.

$$\begin{split} \sum_{i=1}^{4} \left( X_i - \bar{X} \right)^2 &= \sum_{i=1}^{4} X_i^2 - 4\bar{X}^2 \\ &= \sum_{i=1}^{4} X_i^2 - 4 \left[ \frac{1}{4} (X_1 + X_2 + X_3 + X_4) \right]^2 \\ &= \sum_{i=1}^{4} X_i^2 - \frac{1}{2} \left( \sum_{i=1}^{4} X_i^2 + 2 \sum_{1 \le i < j \le 4} X_i X_j \right) \\ &= \frac{3}{4} \sum_{i=1}^{4} X_i^2 - \frac{1}{2} \sum_{1 \le i < j \le 4} X_i X_j \end{split}$$

$$RHS = \left(\frac{1}{2}X_{1}^{2} - X_{1}X_{2} + \frac{1}{2}X_{2}^{2}\right) + \left(\frac{2}{3}X_{3}^{2} + \frac{1}{6}\left(X_{1} + X_{2}\right)^{2} - \frac{2}{3}X_{3}\left(X_{1} + X_{2}\right)\right) + \frac{3}{4}\left(X_{4} - \frac{1}{3}\left(X_{1} + X_{2} + X_{3}\right)\right)^{2} \\ = \cdots \cdot \cdot \left(simple \ but \ tedious \ simpli \ fications\right)$$

$$= \frac{3}{4}\sum_{i=1}^{4}X_{i}^{2} - \frac{1}{2}\sum_{1 \leq i < j \leq 4}X_{i}X_{j}$$

$$\therefore \sum_{i=1}^{4}\left(X_{i} - \bar{X}\right)^{2} = \frac{\left(X_{1} - X_{2}\right)^{2}}{2} + \frac{\left[X_{3} - \frac{\left(X_{1} + X_{2}\right)}{2}\right]^{2}}{\frac{3}{2}}\left[\frac{X_{4} - \frac{\left(X_{1} + X_{2} + X_{3}\right)}{3}\right]^{2}}{\frac{4}{3}}$$

$$\therefore X_{1} - X_{2} \sim N(0, \sqrt{2})$$

$$\therefore \frac{\left(X_{1} - X_{2}\right)^{2}}{2} \sim \chi_{1}^{2}$$

$$\therefore \frac{\left(X_{3} - \frac{X_{1} + X_{2}}{2}\right)^{2}}{\frac{3}{2}} \sim \chi_{1}^{2}$$

$$\therefore \frac{\left(X_{3} - \frac{X_{1} + X_{2}}{2}\right)^{2}}{\frac{4}{3}} \sim N(0, \sqrt{\frac{4}{3}})$$

$$\therefore \frac{\left(X_{4} - \frac{X_{1} + X_{2} + X_{3}}{3}\right)^{2}}{\frac{4}{3}} \sim \chi_{1}^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{3} - X_{4}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$X = \left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$\left(X_{1} - X_{2} - X_{1} - X_{2}\right)^{2}$$

$$\left(X_{1} - X_{2} - X_{2} - X_{2}\right)^{2}$$

$$\left(X_{1} - X_{2} - X_{2} - X$$

So the three terms in the RHS are independent each with a  $\chi_1^2$  distribution.

## Question e

Suppose the radius equals R.  $R_i = R + \epsilon_i$ ,  $i = 1, 2, \dots n$ .

$$E(\bar{R}) = \frac{1}{n} \sum_{i=1}^{n} E(R + \epsilon_1) = \frac{1}{n} \sum_{i=1}^{n} R + E(\epsilon_1) = R$$
$$var(\bar{R}) = var(\bar{R}) + (E\bar{R})^2 = \frac{1}{n^2} \sum_{i=1}^{n} var(R + \epsilon_1) + R^2 = \frac{\sigma^2}{n} + R^2$$

#### Question g

$$E(T-\theta)^2 = var(T) + B^2$$

$$var(T) = var(\alpha_1 \bar{X} + \alpha_2 cS)$$

$$= \alpha_1^2 var(\bar{X}) + \alpha_2^2 var(cS)$$

$$= \alpha_1^2 \frac{\theta^2}{n} + \alpha_2^2 (c^2 - 1)\theta^2$$

$$B = E(T) - \theta = (\alpha_1 + \alpha_2 - 1)\theta$$

$$\therefore E(T-\theta)^2 = \left[\frac{\alpha_1^2}{n} + (c^2 - 1)\alpha_2^2 + (\alpha_1 + \alpha_2 - 1)^2\right]\theta^2$$

$$\left\{\frac{\partial E(T-\theta)^2}{\partial \alpha_1} = \frac{2\alpha_1}{n} + 2(\alpha_1 + \alpha_2 - 1) = 0$$

$$\frac{\partial E(T-\theta)^2}{\partial \alpha_2} = 2(c^2 - 1)\alpha_2 + 2(\alpha_1 + \alpha_2 - 1) = 0$$

$$\therefore \begin{cases} \alpha_1 = \frac{n(c^2 - 1)}{(n+1)(c^2 - 1) + 1} \\ \alpha_2 = \frac{1}{(n+1)(c^2 - 1) + 1} \end{cases}$$

$$A = \frac{\partial^2 E}{\partial \alpha_1^2} = \frac{2}{n} + 2, B = \frac{\partial^2 E}{\partial \alpha_1 \partial \alpha_2} = 2, C = \frac{\partial^2 E}{\partial \alpha_2^2} = 2c^2$$

$$\therefore A < 0, B^2 - AC = 4\left[1 - \frac{n+1}{n}c^2\right] < 0 \ (Using \ Wolfram Alpha)$$

 $\therefore E(T-\theta)^2$  gets the minimum when  $\alpha_1$  and  $\alpha_2$  equal the above values.

$$\therefore T = \frac{n(c^2 - 1)}{(n+1)(c^2 - 1) + 1}\bar{X} + \frac{1}{(n+1)(c^2 - 1) + 1}cS \text{ is the estimator that minimizes } E\left(T - \theta\right)^2$$