

$$\frac{99.2(91)}{A} - \frac{1}{A} \text{ حل تمرین سوره ۱}$$

مسئله ۱:

$$E[Y] = \int_{-1}^1 X' dX = E[X']$$

حال $cov(X, Y)$ را محاسبه می کنیم:

$$cov(X, Y) = \iint (X - E[X]) \cdot (Y - E[Y]) \cdot p(X, Y) dX dY$$

$$= \int_{-1}^1 X (X' - E[X']) dX = \Rightarrow \rho(X, Y) = \frac{cov(X, Y)}{\sqrt{Var(X) Var(Y)}} = 0$$

مسئله ۲:

$$E(XZ|Y) = \int xz p_{x,z|y}(x, z|y) dz dx = \int xz p_{x|y}(x|y) p_{z|y}(z|y) dz dx$$

$$= \int x p_{x|y}(x|y) dx \int z p_{z|y}(z|y) dz = E(X|Y) E(Z|Y) *$$

$$\Rightarrow E(XZ) = E_Y(E(XZ|Y)) = E_Y(E(X|Y) E(Z|Y)) \quad (1)$$

باتوجه به اینکه X, Z به صورت شرطی مستقل از Y هستند، می توان نوشت:

$$P_{X|Y}(X|Y) = N(\mu_{X|Y}, \Sigma_{X|Y}), \quad \Sigma_{X|Y} = cov(X, Y) = \rho_{X,Y} \sqrt{Var(X) Var(Y)}$$

$$\Sigma_{Y|Y}^{-1} = \frac{1}{Var(Y)} \xrightarrow{\text{باتوجه به اینکه } Y \text{ شرطی مستقل از } X, Z} E(X|Y) = \mu_X + \rho_{X,Y} \sqrt{\frac{Var(X)}{Var(Y)}} (Y - \mu_Y) \quad (2)$$

$$P_{Z|Y}(Z|Y) = N(\mu_{Z|Y}, \Sigma_{Z|Y}) \xrightarrow{\text{شرطی مستقل از } Y} E(Z|Y) = \mu_Z + \rho_{Z,Y} \sqrt{\frac{Var(Z)}{Var(Y)}} (Y - \mu_Y) \quad (3)$$

$$\begin{aligned} E(XZ) &= E_Y \left(\mu_X \mu_Z + (\mu_X \rho_{Z,Y} \sqrt{Var(Z)} + \mu_Z \rho_{X,Y} \sqrt{Var(X)}) \frac{Y - \mu_Y}{\sqrt{Var(Y)}} \right. \\ &\quad \left. + \rho_{X,Y} \rho_{Z,Y} \frac{\sqrt{Var(X) Var(Z)}}{Var(Y)} (Y - \mu_Y)^2 \right) \end{aligned}$$

$$= \mu_X \mu_Z + \frac{\mu_X \rho_{Z,Y} \sqrt{Var(Z)} + \mu_Z \rho_{X,Y} \sqrt{Var(X)}}{\sqrt{Var(Y)}} E_Y(Y - \mu_Y) + \frac{\rho_{X,Y} \rho_{Z,Y} \sqrt{Var(X) Var(Z)}}{Var(Y)} E((Y - \mu_Y)^2)$$

$$= \mu_X \mu_Z + \rho_{X,Y} \rho_{Z,Y} \sqrt{Var(X) Var(Z)} \Rightarrow \rho_{X,Z} = \frac{cov(X, Z)}{\sqrt{Var(X) Var(Z)}} = \frac{E(XZ) - E(X)E(Z)}{\sqrt{Var(X) Var(Z)}}$$

$$\Rightarrow \rho_{X,Z} = \frac{\mu_X \mu_Z + \rho_{X,Y} \rho_{Z,Y} \sqrt{Var(X) Var(Z)} - \mu_X \mu_Z}{\sqrt{Var(X) Var(Z)}} = \rho_{X,Y} \rho_{Z,Y}$$

سوال ۳۲

$$\bar{y}_1 = \frac{1}{n_1} \sum_{i=1}^{n_1} y_i^{(1)} \quad , \quad \bar{y}_r = \frac{1}{n_r} \sum_{i=1}^{n_r} y_i^{(r)}$$

prior $\rightarrow \mu_{\mu_2} = 0, \Sigma_{\mu_2} = 0$, likelihood : $A = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ و $\mu_y = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$, $\Sigma_y = \begin{bmatrix} \frac{V_1}{n_1} & 0 \\ 0 & \frac{V_r}{n_r} \end{bmatrix}$

از LGS می‌دانیم $\Sigma_{\mu_2|y}^{-1} = \Sigma_{\mu_2}^{-1} + A^T \Sigma_y^{-1} A = 0 + \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} \frac{V_1}{n_1} & 0 \\ 0 & \frac{V_r}{n_r} \end{bmatrix}^{-1} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{n_1}{V_1} + \frac{n_r}{V_r}$

$$\mu_{\mu_2|y} = \Sigma_{\mu_2|y}^{-1} \left[A^T \Sigma_y^{-1} \bar{y} + \Sigma_{\mu_2}^{-1} \mu_{\mu_2} \right] = \Sigma_{\mu_2|y}^{-1} \left[\frac{n_1 \bar{y}_1}{V_1} + \frac{n_r \bar{y}_r}{V_r} \right]$$

سوال ۲ الف) $NLL(\theta) \Rightarrow NLL(\theta) = -\log \prod_{n=1}^N P(y_n | \theta) = -\log \prod_{n=1}^N \theta^{I(y_n=1)} (1-\theta)^{I(y_n=0)}$

$$= -\sum_{n=1}^N [I(y_n=1) \log \theta + I(y_n=0) \log(1-\theta)] = -[N_1 \log(\theta) + N_0 \log(1-\theta)]$$

برای دست آوردن $\hat{\theta}_{MLE}$ داریم:

$$\frac{d}{d\theta} \cdot NLL(\theta) = \frac{-N_1}{\theta} + \frac{N_0}{1-\theta} = 0 \Rightarrow \hat{\theta}_{MLE} = \frac{N_1}{N_0 + N_1}$$

(۱-)

$$f(x; \theta) = \frac{1}{\theta} e^{-\frac{x}{\theta}}, 0 < x < \infty \text{ و } 0 < \theta < \infty$$

$$\Rightarrow L(\theta) = L(\theta; x_1, \dots, x_n) = \left(\frac{1}{\theta} e^{-\frac{x_1}{\theta}}\right) \left(\frac{1}{\theta} e^{-\frac{x_2}{\theta}}\right) \dots \left(\frac{1}{\theta} e^{-\frac{x_n}{\theta}}\right) = \frac{1}{\theta^n} e^{-\sum_{i=1}^n \frac{x_i}{\theta}}$$

$$\Rightarrow LL(\theta) = -n \log(\theta) - \frac{1}{\theta} \sum_{i=1}^n x_i \Rightarrow \frac{d(LL(\theta))}{d\theta} = \frac{-n}{\theta} + \frac{1}{\theta^2} \sum_{i=1}^n x_i = 0$$

$$\Rightarrow \hat{\theta}_{MLE} = \frac{\sum_{i=1}^n x_i}{n}$$

$$NLL(\mu, \sigma^r) = -\sum_{n=1}^N \log \left[\left(\frac{1}{\sqrt{2\pi}\sigma^r}\right)^{\frac{1}{r}} \exp\left(-\frac{(y_n - \mu)^r}{(\sigma^r)^r}\right) \right] = -\frac{1}{\sigma^r} \sum_{n=1}^N (y_n - \mu)^r - \frac{N}{r} \log(\sqrt{2\pi}\sigma^r)$$

$$\Rightarrow \frac{\partial}{\partial \mu} NLL(\mu, \sigma^r) = 0 \Rightarrow \hat{\mu}_{MLE} = \frac{1}{N} \sum_{n=1}^N y_n = \bar{y}$$

$$\Rightarrow \frac{\partial}{\partial \sigma^r} NLL(\mu, \sigma^r) = 0 \Rightarrow \hat{\sigma}_{MLE}^r = \frac{1}{N} \sum_{n=1}^N (y_n - \hat{\mu}_{MLE})^r = \frac{1}{N} \sum_{n=1}^N [y_n^r + \hat{\mu}_{MLE}^r - r \hat{\mu}_{MLE}^{r-1} y_n]$$

سوال ۸: ابتدا μ و Σ را برای $p(x_1, x_2)$ محاسبه می کنیم:

$$p(x_1, x_2) \propto \exp \left[-\frac{1}{2} \begin{pmatrix} x_1 - \mu_1 \\ x_2 - \mu_2 \end{pmatrix}^T \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}^{-1} \begin{pmatrix} x_1 - \mu_1 \\ x_2 - \mu_2 \end{pmatrix} \right]$$

می توان نوشت:

$$\rightarrow p(x_1, x_2) \propto \exp \left[-\frac{1}{2} \begin{pmatrix} x_1 - \mu_1 \\ x_2 - \mu_2 \end{pmatrix}^T \begin{pmatrix} I & 0 \\ -\Sigma_{21}^{-1} \Sigma_{11} & I \end{pmatrix} \begin{pmatrix} (\Sigma / \Sigma_{22})^{-1} & 0 \\ 0 & \Sigma_{22}^{-1} \end{pmatrix} \begin{pmatrix} I & -\Sigma_{12} \Sigma_{22}^{-1} \\ 0 & I \end{pmatrix} \begin{pmatrix} x_1 - \mu_1 \\ x_2 - \mu_2 \end{pmatrix} \right]$$

$$= \exp \left[-\frac{1}{2} (x_1 - \mu_1 - \Sigma_{12} \Sigma_{22}^{-1} (x_2 - \mu_2))^T (\Sigma / \Sigma_{22})^{-1} (x_1 - \mu_1 - \Sigma_{12} \Sigma_{22}^{-1} (x_2 - \mu_2)) \right]$$

$$\times \exp \left[-\frac{1}{2} (x_2 - \mu_2)^T \Sigma_{22}^{-1} (x_2 - \mu_2) \right] \quad (1)$$

$$\Rightarrow p(x_1, x_2) = p(x_1 | x_2) p(x_2) \Rightarrow p(x_1, x_2) \propto N(x_1 | \mu_{1|2}, \Sigma_{11|2}) N(x_2 | \mu_2, \Sigma_{22}) \quad (2)$$

$$\Rightarrow \mu_{1|2} = \mu_1 + \Sigma_{12} \Sigma_{22}^{-1} (x_2 - \mu_2), \quad \Sigma_{11|2} = \Sigma_{11} - \Sigma_{12} \Sigma_{22}^{-1} \Sigma_{21}$$