

## \* Exercise1

Write down the pdf form of a multivariate normal distribution.

$\mathbf{X} \sim N(\boldsymbol{\mu}, \Sigma)$  or  $\mathbf{X} \sim N_k(\boldsymbol{\mu}, \Sigma)$  標示出  $k$ -dimensional

Mean  $\boldsymbol{\mu} = E[\mathbf{X}] = [E[X_1], E[X_2], E[X_3], \dots, E[X_{k-1}], E[X_k], ]^T$ ,

$$\text{Covariance matrix } \Sigma = \begin{pmatrix} \text{var}(X_1) & \cdots & \text{cov}(X_1, X_k) \\ \vdots & \ddots & \vdots \\ \text{cov}(X_k, X_1) & \cdots & \text{var}(X_k) \end{pmatrix} = \begin{pmatrix} \sigma_1^2 & \cdots & \rho_{1k}\sigma_1\sigma_k \\ \vdots & \ddots & \vdots \\ \rho_{k1}\sigma_k\sigma_1 & \cdots & \sigma_k^2 \end{pmatrix}$$

The pdf of multivariate normal:

$$f_{\mathbf{X}}(x_1, x_2, x_3, \dots, x_{k-1}, x_k) = \frac{\exp\left(-\frac{1}{2}(\mathbf{X}-\boldsymbol{\mu})^T \Sigma^{-1}(\mathbf{X}-\boldsymbol{\mu})\right)}{\sqrt{(2\pi)^k |\Sigma|}}, \quad \mathbf{X} \in \mathbb{R}^k.$$

## \*Question A.1

Let  $\mathbf{y}$  be an  $N$  by 1 random vector with  $E\mathbf{y} = \mathbf{X}\boldsymbol{\beta}$ , and  $\text{Var}(\mathbf{y}) = \sigma^2 \mathbf{I}_N$ , where  $\mathbf{X}$  is  $N$  by  $p$  and  $\boldsymbol{\beta}$  is  $p$  by 1. Let  $\mathbf{Y} = \mathbf{P}\mathbf{y}$ , where  $\mathbf{P}$  is a projection (not necessarily orthogonal) onto the range of  $\mathbf{X}$ . (a) Find the mean and (co)variance of  $\mathbf{Y}$  and  $\mathbf{y}-\mathbf{Y}$ . (b) Prove that  $\text{Cov}(\mathbf{Y}, \mathbf{y}-\mathbf{Y})$  is 0 if and only if  $\mathbf{P}$  is an orthogonal projection.

(a) Find the mean and (co)variance of  $\mathbf{Y}$  and  $\mathbf{y}-\mathbf{Y}$ .

$$E(\mathbf{Y}) = E(\mathbf{P}\mathbf{y}) = \mathbf{P}E(\mathbf{y}) = \mathbf{P}\mathbf{X}\boldsymbol{\beta} = \mathbf{X}\boldsymbol{\beta};$$

$$\text{Var}(\mathbf{Y}) = \text{Var}(\mathbf{P}\mathbf{y}) = \mathbf{P}\text{Var}(\mathbf{y})\mathbf{P}' = \mathbf{P}\sigma^2 \mathbf{I}_N \mathbf{P}' = \mathbf{P}\mathbf{P}'\sigma^2$$

$$E(\mathbf{y} - \mathbf{Y}) = E(\mathbf{y} - \mathbf{P}\mathbf{y}) = E((\mathbf{I}_N - \mathbf{P})\mathbf{y}) = (\mathbf{I}_N - \mathbf{P})E(\mathbf{y}) = (\mathbf{I}_N - \mathbf{P})\mathbf{X}\boldsymbol{\beta} = (\mathbf{X} - \mathbf{P}\mathbf{X})\boldsymbol{\beta} = \mathbf{X}\boldsymbol{\beta} - \mathbf{P}\mathbf{X}\boldsymbol{\beta}$$

$$\text{Var}(\mathbf{y} - \mathbf{Y}) = \text{Var}(\mathbf{y}) + \text{Var}(\mathbf{Y}) - 2\text{Cov}(\mathbf{y}, \mathbf{Y}) = \sigma^2 \mathbf{I}_N + \mathbf{P}\sigma^2 \mathbf{P}' - \text{cov}(\mathbf{y}, \mathbf{P}\mathbf{y}) - \text{cov}(\mathbf{P}\mathbf{y}, \mathbf{y})$$

$$= \sigma^2 + \mathbf{P}\mathbf{P}'\sigma^2 - \mathbf{P}'\sigma^2 - \mathbf{P}\sigma^2$$

(b) Prove that  $\text{Cov}(Y, y - Y)$  is 0 if and only if  $P$  is an orthogonal projection.

$$\Rightarrow \text{cov}(Y, y - Y) = \text{cov}(Py, (I_N - P)y) = P\text{var}(y)(I_N - P)' = PI_N\sigma^2(I_N' - P') = (PI_N' - PP')\sigma^2,$$

$$(\text{If } P \text{ is orthogonal projection then } P = P', PP = P) (PI_N' - PP')\sigma^2 = (P - PP)\sigma^2 = (P - P)\sigma^2 = 0,$$

only when  $P$  is an orthogonal projection. Therefore,  $P$  is an orthogonal projection.

$$\Leftarrow P \text{ is an orthogonal projection, then } P' = P, PP = P, P - P'P = P - PP = P - P = 0$$

$$0 = (P - P'P)\sigma^2 I_N = (P - PP)\sigma^2 I_N = P(I_N - P')\sigma^2 I_N = P(I_N' - P')\sigma^2 I_N = P\sigma^2 I_N(I_N' - P')$$

$$= P\text{var}(y)((I_N' - P')) = P\text{cov}(y, y)(I_N' - P') = P\text{cov}(y, y)(I_N - P)' = \text{cov}(Py, (I - P)y) = \text{cov}(Y, y - Y)$$

Therefore,  $\text{cov}(Y, y - Y) = 0$

## **\*Question A.2**

Let  $y = X\beta + \epsilon$ , where  $\epsilon$  is iid  $N(0, I_N\sigma^2)$ ;  $y$  is  $N$  by 1,  $X$  is  $N$  by  $p$ , and  $\beta$  is  $p$  by 1. Let  $g$  be any  $N$  by 1 vector. What is the distribution of  $(g'y)^2$ ? What, if anything, changes when  $g'X$  is zero?

$$\text{Let } E(y) = E(X\beta + \epsilon) = E(X\beta) + E(\epsilon) = E(X\beta) = \mu$$

$$\text{Var}(y) = \text{Var}(X\beta + \epsilon) = \text{Var}(\epsilon) = I_N\sigma^2$$

$$y \sim N(\mu, I_N\sigma^2)$$

$$g'y \sim N(g'\mu, g'I_N\sigma^2g)$$

(a) What is the distribution of  $(g'y)^2$

$$\frac{g'y}{(g'I_N\sigma^2g)^{\frac{1}{2}}} \sim N\left(\frac{g'\mu}{(g'I_N\sigma^2g)^{\frac{1}{2}}}, 1\right)$$

$$\Rightarrow \left(\frac{g'y}{(g'I_N\sigma^2g)^{\frac{1}{2}}}\right)\left(\frac{g'y}{(g'I_N\sigma^2g)^{\frac{1}{2}}}\right) \sim \text{noncentral } \chi^2(1) \text{ and noncentrality parameter } \Omega = \frac{\mu'gg'\mu}{2}$$

(b)  $g'X = 0$

$$g'X = 0 \Rightarrow g'\mu = g'X\beta = 0 \Rightarrow \frac{g'y}{(g'I_N\sigma^2g)^{\frac{1}{2}}} \sim N(0, 1)$$

$$\Rightarrow \left(\frac{g'y}{(g'I_N\sigma^2g)^{\frac{1}{2}}}\right)\left(\frac{g'y}{(g'I_N\sigma^2g)^{\frac{1}{2}}}\right) \sim \chi^2(1) \Rightarrow (g'y)(g'y) \sim \text{Gamma}\left(\frac{1}{2}, 2(g'I_N\sigma^2g)\right)$$