

已知联合概率求边缘概率及条件概率

$$X \sim N(\mu, \Sigma) = \frac{1}{(2\pi)^{p/2} |\Sigma|^{1/2}} \cdot \exp\left[-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)\right]$$

2. $X = \begin{pmatrix} x_a \\ x_b \end{pmatrix} \begin{matrix} \rightarrow m \text{ 已知} \\ \rightarrow n \end{matrix} \quad \mu = \begin{pmatrix} \mu_a \\ \mu_b \end{pmatrix} \begin{matrix} p \\ q \end{matrix} \quad \Sigma = \begin{pmatrix} \Sigma_{aa} & \Sigma_{ab} \\ \Sigma_{ba} & \Sigma_{bb} \end{pmatrix}$

then ① x_1 和 $x_{2|} = x_2 - \Sigma_{21} \Sigma_{11}^{-1} x_1$ 是统计条件独立.

$$\textcircled{2} X_1 \sim N(\mu_1, \Sigma_{11}) \quad X_{2|} \sim N(\mu_{2|}, \Sigma_{2|})$$

$$\mu_{2|} = \mu_2 - \Sigma_{21} \Sigma_{11}^{-1} \mu_1 \quad \Sigma_{2|} = \Sigma_{22} - \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12}$$

记

$$x_a = \begin{pmatrix} 1_m & 0_n \end{pmatrix} \begin{pmatrix} x_a \\ x_b \end{pmatrix}$$

A 易求 X 已知量

$$E[x_a] = A E(X) = \mu_a$$

$$\text{Var}[x_a] = \begin{pmatrix} 1_m & 0_n \end{pmatrix} \begin{pmatrix} \Sigma_{aa} & \Sigma_{ab} \\ \Sigma_{ab} & \Sigma_{bb} \end{pmatrix} \begin{pmatrix} 1_m \\ 0_n \end{pmatrix} = \Sigma_{aa}$$

$$x_a \sim N(\mu_a, \Sigma_{aa})$$

$p(x_a)$

$$x_{b|a} = \begin{pmatrix} -\Sigma_{ba} \Sigma_{aa}^{-1} & 1 \end{pmatrix} \begin{pmatrix} x_a \\ x_b \end{pmatrix} \leftarrow X$$

$$E(x_{b|a}) = \begin{pmatrix} -\Sigma_{ba} \Sigma_{aa}^{-1} & 1 \end{pmatrix} \begin{pmatrix} \mu_a \\ \mu_b \end{pmatrix} = \mu_b - \Sigma_{ba} \Sigma_{aa}^{-1} \mu_a = \mu_{b|a}$$

$$\text{Var}[x_{b|a}] = \begin{pmatrix} -\Sigma_{ba} \Sigma_{aa}^{-1} & 1 \end{pmatrix} \begin{pmatrix} \Sigma_{aa} & \Sigma_{ab} \\ \Sigma_{ba} & \Sigma_{bb} \end{pmatrix} \begin{pmatrix} -\Sigma_{aa}^{-1} \Sigma_{ba} \\ 1 \end{pmatrix} = \Sigma_{bb|a}$$



