

# 高斯公式

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad-bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

这里有很多假设，

$$\Sigma_{11} \text{ 对称且 } \Sigma_{12}^T = \Sigma_{21}$$

把一个高斯用两个相互独立的高斯替代。

这里是将  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$  的空间到  $\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$  空间的线性变换。

Jacobian 矩阵为 B。

目前需要讨论的问题：

$$\Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix} \sim \begin{pmatrix} \Sigma_{11} & 0 \\ 0 & \Sigma_{22.1} \end{pmatrix} \quad \Sigma = A Q A^T$$

$$= \begin{pmatrix} I_p & 0 \\ -\Sigma_{21}\Sigma_{11}^{-1} & I_q \end{pmatrix} \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix} \begin{pmatrix} I_p & -\Sigma_{11}^{-1}\Sigma_{12} \\ 0 & I_q \end{pmatrix} \quad A Q A^T \text{ 分解}$$

$$= \begin{pmatrix} \Sigma_{11} & 0 \\ 0 & \Sigma_{22.1} \end{pmatrix} \text{ 块对角化}$$

化简得

$$\Sigma^{-1} = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}^{-1} = \begin{pmatrix} I_p & -\Sigma_{11}^{-1}\Sigma_{12} \\ 0 & I_q \end{pmatrix} \begin{pmatrix} \Sigma_{11}^{-1} & 0 \\ 0 & \Sigma_{22.1}^{-1} \end{pmatrix} \begin{pmatrix} I_p & 0 \\ -\Sigma_{11}^{-1}\Sigma_{12}^T & I_q \end{pmatrix}$$

$$(A - B D^{-1} C)^T = A^T + A^T B (D - C A^T B)^T C A^T$$

$n \times n \quad n \times 1 \quad 1 \times 1 \quad 1 \times n$

1x1 简化计算

线性变换

Jacobian 为

$$z = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ -\Sigma_{21}\Sigma_{11}^{-1} & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \longrightarrow dz = (\det B) dx$$

$$\downarrow z(x - \mu) = B^T z \longrightarrow (B^T z)^T z (B^T z)^T = z (B B^T)^T z$$

$$\downarrow z^T (B \Sigma B^T)^T z = (z_1, z_2) \begin{pmatrix} \Sigma_{11} & 0 \\ 0 & \Sigma_{22.1} \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$$

↓

将一个高斯分为  $z_1^T \Sigma_{11} z_1$  和  $z_2^T \Sigma_{22.1} z_2$  的两个独立高斯。

$$(B \Sigma B^T)^T = \begin{pmatrix} \Sigma_{11} & 0 \\ 0 & \Sigma_{22.1} \end{pmatrix}$$

条件概率分布

$$X_2 | X_1 \sim N(\mu_2 + \Sigma_{21} \Sigma_{11}^{-1} (X_1 - \mu_1), \Sigma_{22.1})$$

$$p(X_1, X_2) = p(X_1) p(X_2 | X_1)$$

$$\text{已知 } X_{21} = X_2 - \Sigma_{21} \Sigma_{11}^{-1} X_1$$

$$X_2 = X_{21} + \Sigma_{21} \Sigma_{11}^{-1} X_1$$

$$E(X_2) = \mu_{21} + \Sigma_{21} \Sigma_{11}^{-1} \mu_1$$











