

已知边缘和条件概率求联合分布

已知: $P(x) = N(x|\mu, \Lambda^{-1})$

$$P(y|x) = N(y|Ax+b, L^{-1})$$

求: ① $P(y)$, ② $P(x|y)$

解: ① $y = Ax + b + \varepsilon$

$$\varepsilon \sim N(0, L^{-1})$$

$$E(y) = E[Ax + b + \varepsilon]$$

$$= E(Ax + b) + E[\varepsilon]$$

$$= A\mu + b + 0$$

$$= A\mu + b$$

$$\text{Var}[y] = \text{Var}[Ax + b + \varepsilon]$$

$$= \text{Var}[Ax + b] + \text{Var}[\varepsilon]$$

$$= A\Lambda^{-1}A^T + L^{-1}$$

$$y \sim N(A\mu + b, A\Lambda^{-1}A^T + L^{-1})$$

$$\textcircled{2} z = \begin{pmatrix} x \\ y \end{pmatrix} \sim N \left(\begin{bmatrix} \mu \\ A\mu + b \end{bmatrix}, \begin{bmatrix} \Lambda^{-1} & 0 \\ 0 & L^{-1} + A\Lambda^{-1}A^T \end{bmatrix} \right)$$

$$\Delta = \text{cov}(x, y)$$

$$= E[(x - E(x))(y - E(y))^T]$$

$$= E[(x - \mu)(y - A\mu - b)^T]$$

$$= E[(x - \mu)(Ax + b + \varepsilon - A\mu - b)^T]$$

$$= E[(x - \mu)(Ax - A\mu + \varepsilon)^T]$$

$$= E[(x - \mu)(x - \mu)^T A^T + (x - \mu)\varepsilon^T]$$

$$= E[(x - \mu)(x - \mu)^T A^T] + E[(x - \mu)\varepsilon^T] \stackrel{=0}{=}$$

$$= E[(x - \mu)(x - \mu)^T] A^T$$

$$= \text{Var}[x] A^T = \Lambda^{-1} A^T$$

$$P(x|y) = N(_, _)$$

