

1 KL divergence

$$D_{KL}(P||Q) = \mathbb{E}_{x \sim P} \left[\log \left(\frac{p(x)}{q(x)} \right) \right]$$

2 Multivariate Normal

- **KL divergence**

$$\begin{aligned} D_{KL}(\mathcal{N}_a || \mathcal{N}_b) &= \frac{1}{2} \left(\text{Tr}(\Sigma_b^{-1} \Sigma_a) + (\mu_b - \mu_a)^T \Sigma_b^{-1} (\mu_b - \mu_a) - k + \log \frac{|\Sigma_b|}{|\Sigma_a|} \right) \\ &= \frac{1}{2} \left(\|\mathbf{L}_b \setminus \mathbf{L}_a\|_F^2 + \|\mathbf{L}_b \setminus (\mu_b - \mu_a)\|_2^2 - k + 2 \left(\log \text{diag}(\mathbf{L}_b)^T \mathbf{1} - \log \text{diag}(\mathbf{L}_a)^T \mathbf{1} \right) \right) \end{aligned}$$

- **Linear Gaussian systems** Given a linear system:

$$\begin{aligned} p(\mathbf{x}) &= \mathcal{N}(\mathbf{x} \mid \mu_x, \Sigma_x) \\ p(\mathbf{y} \mid \mathbf{x}) &= \mathcal{N}(\mathbf{y} \mid \mathbf{A}\mathbf{x} + \mathbf{b}, \Sigma_y) \end{aligned}$$

We have the following:

$$\begin{aligned} p(\mathbf{x} \mid \mathbf{y}) &= \mathcal{N}(\mathbf{x} \mid \mu_{x|y}, \Sigma_{x|y}) \\ \mu_{x|y} &= \Sigma_{x|y} \left[\mathbf{A}^T \Sigma_y^{-1} (\mathbf{y} - \mathbf{b}) + \Sigma_x^{-1} \mu_x \right] \\ \Sigma_{x|y} &= \Sigma_x^{-1} + \mathbf{A}^T \Sigma_y^{-1} \mathbf{A} \end{aligned}$$

$$p(\mathbf{y}) = \mathcal{N}(\mathbf{y} \mid \mathbf{A}\mu_x + \mathbf{b}, \Sigma_y + \mathbf{A}\Sigma_x\mathbf{A}^T)$$

- **quadratic relations**

$$- \mathbb{E}_{\mathbf{x} \sim \mathcal{N}(\mu, \Sigma)} [\mathbf{x}^T \mathbf{A} \mathbf{x}] = \text{Tr}(\mathbf{A} \Sigma) + \mu^T \mathbf{A} \mu$$

—

$$\mathbb{E}_{\mathbf{x} \sim \mathcal{N}(\mathbf{b}, \mathbf{B})} \left[(\mathbf{a} - \mathbf{A}\mathbf{x})^T \Sigma^{-1} (\mathbf{a} - \mathbf{A}\mathbf{x}) \right] = (\mathbf{a} - \mathbf{A}\mathbf{b})^T \Sigma^{-1} (\mathbf{a} - \mathbf{A}\mathbf{b}) + \text{Tr}(\mathbf{A}^T \Sigma^{-1} \mathbf{A} \mathbf{B})$$