

$$( \text{proof} \quad \int_{\mathbb{R}^k} f(x) dx = 1$$

$$\hat{=} y = x - u, \quad \int_{\mathbb{R}^k} f(x) dx = \int_{\mathbb{R}^k} \frac{1}{(2\pi)^k |\Sigma|} \exp\left(-\frac{1}{2} y^T \Sigma^{-1} y\right) dy$$

$$\text{其中, } \Sigma = Q \Lambda Q^T, \quad \Lambda = \text{diag}(\lambda_1, \dots, \lambda_k), \quad |\Sigma| = \prod_{i=1}^k \lambda_i$$

$$\hat{=} z = Q^T y \Rightarrow y^T \Sigma^{-1} y = z^T \Lambda^{-1} z = \sum_{i=1}^k \frac{z_i^2}{\lambda_i}$$

$$\Rightarrow \int_{\mathbb{R}^k} \frac{1}{(2\pi)^k \prod_{i=1}^k \lambda_i} \exp\left(-\frac{1}{2} \sum_{i=1}^k \frac{z_i^2}{\lambda_i}\right) dz$$

$$= \int_{\mathbb{R}^k} \left( \frac{1}{\sqrt{2\pi\lambda_i}} \int_{-\infty}^{\infty} \exp\left(-\frac{z_i^2}{2\lambda_i}\right) dz_i \right)$$

$$\because \int_{-\infty}^{\infty} \exp\left(-\frac{z^2}{2\lambda}\right) dz = \sqrt{2\pi\lambda}$$

$$\therefore \int_{\mathbb{R}^k} f(x) dx = 1$$

$$2. \quad (a) \quad \frac{\partial}{\partial A} \operatorname{tr}(AB) = B^T$$

$$\operatorname{tr}(AB) = \sum_{i=1}^n (AB)_{ii} = \sum_{i=1}^n \sum_{j=1}^n A_{ij} B_{ji}$$

$$\frac{\partial}{\partial A_{kl}} \operatorname{tr}(AB) = \sum_{i,j} \frac{\partial A_{ij}}{\partial A_{kl}} B_{ji} = B_{lk}$$

$$\Rightarrow \frac{\partial}{\partial A} \operatorname{tr}(AB) = B^T$$

$$(b) \quad \text{tr}(AB) = \text{tr}(BA)$$

$$\underline{x^T A x} = \text{tr}(x^T A x) = \text{tr}(A x x^T) = \underline{\text{tr}(x x^T A)}$$

(c)

1.

$$\frac{\partial \ell}{\partial \mu} = -\frac{1}{2} \sum_{i=1}^N (-2 \bar{x}^T (x_i - \mu)) = \bar{x}^T \sum_{i=1}^N (x_i - \mu)$$

$$\bar{x}^T \sum_{i=1}^N (x_i - \mu) = 0 \Rightarrow \sum_{i=1}^N x_i - N\mu = 0$$

$$\Rightarrow \hat{\mu}_n = \frac{1}{N} \sum_{i=1}^N x_i$$

2.

$$S = \sum_{i=1}^N (x_i - \hat{\mu}) (x_i - \hat{\mu})^T$$

$$\sum_{i=1}^N (x_i - \mu)^T \Sigma^{-1} (x_i - \mu) = \text{tr}(S \Sigma^{-1})$$

$$\frac{\partial \ell}{\partial \Sigma} = \frac{N}{2} \Sigma^{-1} - \frac{1}{2} S$$

$$\hat{\Sigma} \quad \frac{N}{2} \Sigma^{-1} - \frac{1}{2} S = 0 \Rightarrow N \Sigma = S$$

$$\therefore \Sigma_{ML} = \frac{1}{N} S = \frac{1}{N} \sum_{i=1}^N (x_i - \hat{\mu}) (x_i - \hat{\mu})^T$$