

Assignment 5

1.

$$\begin{aligned}
 \int_{\mathbb{R}^k} f(x) dx &= \int_{\mathbb{R}^k} \frac{1}{(2\pi)^{k/2} |\Sigma|^{1/2}} \exp\left(-\frac{1}{2} (x-\mu)^T \Sigma^{-1} (x-\mu)\right) dx, \text{ let } y = \sqrt{\Sigma^{-1}} (x-\mu) \\
 &= \int_{\mathbb{R}^k} \frac{1}{(2\pi)^{k/2} |\Sigma|^{1/2}} \exp\left(-\frac{1}{2} y^T y\right) \cdot \frac{dy}{\sqrt{|\Sigma^{-1}|}} \quad dy = \sqrt{|\Sigma^{-1}|} dx \\
 &= \int_{\mathbb{R}^k} \frac{1}{(2\pi)^{k/2} |\Sigma|^{1/2}} e^{-\frac{1}{2} \sum_{i=1}^k y_i^2} \cdot \frac{dy}{\sqrt{|\Sigma^{-1}|}} \\
 &= \frac{1}{(2\pi)^{k/2}} \int_{\mathbb{R}^k} e^{-\frac{1}{2} \sum_{i=1}^k y_i^2} dy \\
 &= \frac{1}{(2\pi)^{k/2}} \prod_{i=1}^k \int_{-\infty}^{\infty} e^{-\frac{1}{2} y_i^2} dy_i, \text{ note that } \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} dz = \sqrt{2\pi}, z \in \mathbb{R}^1 \\
 &= \frac{1}{(2\pi)^{k/2}} \cdot (\sqrt{2\pi})^k \\
 &= 1.
 \end{aligned}$$

Suppose $I = \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} dz$. Then $I^2 = \left(\int_{-\infty}^{\infty} e^{-\frac{z_1^2}{2}} dz_1\right) \left(\int_{-\infty}^{\infty} e^{-\frac{z_2^2}{2}} dz_2\right)$

$$\begin{aligned}
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-(z_1^2 + z_2^2)/2} dz_1 dz_2 \\
 &= \int_0^{\infty} \int_0^{2\pi} e^{-r^2/2} r d\theta dr, \text{ by polar coordinate} \\
 &= 2\pi \left(-e^{-\frac{r^2}{2}}\right) \Big|_0^{\infty} = 2\pi.
 \end{aligned}$$

Thus, $I = \sqrt{2\pi}$ \square

2.

(a) $\text{tr}(AB) = \sum_{i,j=1}^n a_{ij} b_{ji}$

$$\begin{aligned}
 \Rightarrow \frac{\partial}{\partial A} \text{tr}(AB) &= \begin{bmatrix} \frac{\partial}{\partial a_{11}} \text{tr}(AB) & \frac{\partial}{\partial a_{12}} \text{tr}(AB) & \dots \\ \frac{\partial}{\partial a_{21}} \text{tr}(AB) & & \\ & & \frac{\partial}{\partial a_{nn}} \text{tr}(AB) \end{bmatrix} \\
 &= \begin{bmatrix} b_{11} & b_{21} & \dots \\ b_{12} & & \\ \vdots & & b_{nn} \end{bmatrix} = B^T
 \end{aligned}$$

since $\frac{\partial}{\partial a_{pq}} \sum_{i,j=1}^n a_{ij} b_{ji} = b_{qp}$ for specific p, q .

$$(b) \operatorname{tr}(xx^T A) = \operatorname{tr}(A(xx^T)) \text{ by the property of trace}$$

$$= \sum_{i,j=1}^n a_{ij} x_j x_i$$

$$= \sum_{i,j=1}^n x_i a_{ij} x_j$$

$$= x^T A x$$

$$(c) f(x_i) = \frac{1}{(2\pi)^{n/2} |\Sigma|^{1/2}} \exp\left(-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)\right)$$

$$\text{take log} \Rightarrow L = -\frac{1}{2} \sum_{i=1}^n (x_i - \mu)^T \Sigma^{-1} (x_i - \mu) - \frac{n}{2} \ln(2\pi)^n - \frac{n}{2} \ln |\Sigma|$$

$$\stackrel{(b)}{=} -\frac{1}{2} \sum_{i=1}^n \operatorname{tr}((x_i - \mu)(x_i - \mu)^T \Sigma^{-1}) - \frac{n}{2} \ln(2\pi)^n + \frac{n}{2} \ln |\Sigma^{-1}| \text{ by } |A| = \frac{1}{|A^{-1}|}$$

A : invertible matrix

To find $\bar{\mu}, \bar{\Sigma}$ s.t. $L(\bar{\mu}, \bar{\Sigma}) = \max L(\mu, \Sigma)$, we have

$$\frac{\partial}{\partial \mu} L = \sum_{i=1}^n (x_i - \mu) \Sigma^{-1} = 0 \Rightarrow \bar{\mu} = \frac{1}{n} \sum_{i=1}^n x_i \text{ and}$$

$$\frac{\partial}{\partial \Sigma^{-1}} L = \frac{\partial}{\partial \Sigma^{-1}} \left[-\frac{1}{2} \sum_{i=1}^n \operatorname{tr}((x_i - \mu)(x_i - \mu)^T \Sigma^{-1}) \right] + \frac{n}{2} \Sigma$$

$$\stackrel{(a)}{=} -\frac{1}{2} \sum_{i=1}^n (x_i - \mu)^T (x_i - \mu) + \frac{n}{2} \Sigma = 0 \Rightarrow \bar{\Sigma} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{\mu})(x_i - \bar{\mu})^T$$

Thus, MLE gives $\bar{\mu} = \frac{1}{n} \sum_{i=1}^n x_i$ and $\bar{\Sigma} = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{\mu})(x_i - \bar{\mu})^T$ \square

3.

In GDA, we know that $\Sigma_k = \frac{1}{n_k} \sum_{i: y^{(i)}=k} (x^{(i)} - \mu_k)(x^{(i)} - \mu_k)^T$.

Note that $\operatorname{rank}(AA^T) = \operatorname{rank}(A)$. Then for $(x^{(i)} - \mu_k)$, $\operatorname{rank}(\Sigma_k) = \operatorname{rank}(x^{(i)} - \mu_k) \leq n_k - 1$

since $\sum_{i=1}^{n_k} (x^{(i)} - \mu_k) = 0$ implies that $x^{(i)} - \mu_k$ are dependent.

So if $n_k < n$, then $\operatorname{rank}(\Sigma_k) < n$ implies that Σ_k isn't invertible, and

we need Σ_k is invertible. So how to solve it? Is doing regularization

to Σ_k the solution?