

MATH 5473 HW3

1. (a) Let $f(x) = \frac{1}{(2\pi)^p |\Sigma|} \exp \left\{ -\frac{1}{2} (x_i - \mu)^T \Sigma^{-1} (x_i - \mu) \right\}$

$$\log f(x) = \sum_{i=1}^n \left[-\frac{p}{2} \log(2\pi) - \frac{1}{2} \log(|\Sigma|) - \frac{1}{2} (x_i - \mu)^T \Sigma^{-1} (x_i - \mu) \right]$$

$$f_n = \text{trace}(\log f(x))$$

$$\begin{aligned} &= \text{trace}\left(-\frac{np}{2} \log(2\pi)\right) - \frac{n}{2} \log \det(\Sigma) - \frac{1}{2} \sum_{i=1}^n \text{trace}[(x_i - \mu)^T \Sigma^{-1} (x_i - \mu)] \\ &= -\frac{n}{2} \text{trace}(\Sigma^{-1} S_n) - \frac{n}{2} \log \det(\Sigma) + C. \end{aligned}$$

here $S_n = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)(x_i - \mu)^T$.

(b) $f(x + \Delta) = \text{trace}[A(x + \Delta)^{-1}]$

$$= \text{trace}[A x^{-1} (I + \Delta x^{-1})^{-1}]$$

$$\approx \text{trace}[A x^{-1} (I - \Delta x^{-1})]$$

$$= \text{trace}(Ax^{-1}) - \text{trace}(A' x' \Delta x')$$

$$= \text{trace}(Ax^{-1}) - \text{trace}(x^{-1} A' x^{-1} \Delta)$$

$$= f(x) - \text{trace}(x^{-1} A' x^{-1} \Delta)$$

(c) $g(x + \Delta) = \log \det(x + \Delta)$

$$= \log \det X + \log \det(I + X^{-\frac{1}{2}} \Delta X^{-\frac{1}{2}})$$

$$= \log \det X + \sum_{i=1}^n \log(1 + \lambda_i).$$

here λ_i is the i -th eigenvalue of $X^{-\frac{1}{2}} \Delta X^{-\frac{1}{2}}$.

For small Δ , $\log(1 + \lambda_i) \approx \lambda_i$.

$$\begin{aligned}
 g(X + \Delta) &\approx \log \det(X) + \sum_{i=1}^n \lambda_i \\
 &= \log \det(X) + \text{trace}(X^{-\frac{1}{2}} \Delta X^{-\frac{1}{2}}) \\
 &= \log \det(X) + \text{trace}(X^{-1} \Delta)
 \end{aligned}$$

(d) By (a), $f(X) = -\frac{n}{2} \text{trace}(\Sigma^{-1} S_n) - \frac{n}{2} \log \det(\Sigma) + C$.

$$\begin{aligned}
 \Rightarrow 0 &= \frac{\partial f}{\partial \Sigma} = -\frac{n}{2} (-\Sigma^{-1} S_n \Sigma^{-1}) - \frac{n}{2} \Sigma^{-1} \\
 \Rightarrow \hat{\Sigma} &= S_n
 \end{aligned}$$

2. (a) Let $R(\mu) = \frac{1}{2} \|y - \mu\|_2^2 + \frac{\lambda}{2} \|\mu\|_2^2 = \frac{1}{2} (y^T y - 2y^T \mu + \mu^T \mu) + \frac{\lambda}{2} \mu^T \mu$

$$\begin{aligned}
 \Rightarrow 0 &= \frac{\partial R}{\partial \mu} = -\frac{1}{2} y + \frac{1+\lambda}{2} \hat{\mu} \\
 \Rightarrow \hat{\mu}_i &= \frac{1}{1+\lambda} y_i
 \end{aligned}$$

(b) Let $L(\mu) = \frac{1}{2} \|y - \mu\|_2^2 + \lambda \|\mu\|_1$

And $\partial_\mu L = \mu y + \lambda \text{sign}(\mu)$.
 $0 \in \partial_\mu L \Big|_{\mu=\mu^*} \Rightarrow \mu^* = \begin{cases} y & y > \lambda \\ 0 & -\lambda \leq y \leq \lambda \\ y + \lambda & y < -\lambda \end{cases}$

Thus, $\hat{\mu}_i = \text{sign}(y_i) (|y_i| - \lambda)_+$.

(c) Let $F(\mu) = \|y - \mu\|_2^2 + \lambda^2 \|\mu\|_0$

$$F(\mu_i) = \begin{cases} (y_i - \mu_i)^2 + \lambda^2 & \mu_i \neq 0 \\ y_i^2 & \mu_i = 0 \end{cases}$$

$$\Rightarrow \hat{\mu}(y) = y_i I(|y_i| > \lambda).$$

If $\hat{\mu}(y) = (1-g(y))y$,

$\Rightarrow g(y) = I(|y| < \lambda)$, $g(y)$ is not absolutely continuous.

Thus $g(x)$ is not weakly differentiable.

(d). Let $g(y) = -\frac{\alpha}{\|y\|^2}$ $\Rightarrow \hat{\mu} = (1+g)y$.

And $E\|\hat{\mu}(y) - \mu\|^2 = E U_\alpha(y)$. $U_\alpha(y) = \frac{\partial U_\alpha(y)}{\partial \alpha}$.

$\Rightarrow \alpha^* = \arg \min_{\alpha} U_\alpha(y) = p-2$.

And $\|y\|^2 \sim X^2(\|\mu\|^2, p)$. $N \sim \text{Poisson}(\frac{1}{2}\|\mu\|^2)$

$$E\left(\frac{1}{\|y\|^2}\right) = E\left(\frac{1}{p+2N-p}\right) \geq \frac{1}{p+\|\mu\|^2-p},$$

For $p > 3$, $\text{Risk}(\mu) < p - \frac{(p-2)^2}{p-2+\|\mu\|^2} < p$.

(e). The above four estimates are all shrinkage rules.

3. For any matrix C and orthogonal matrix P ,

$P^T C P$ is admissible iff Cy is admissible.

For given C , we have orthogonal matrix P and diagonal matrix $D = \text{diag}\{d_1, d_2, \dots\}$, s.t. $P^T D P = C$.

Then Dy is admissible if one or two $d_i = 1$ and others $0 \leq d_i < 1$.

Pf. Let $d_1 = d_2 = 1$, $0 \leq d_i < 1$, $i \geq 3$.

(Contradiction) There exists an estimate $H(y) = (h_1(y), h_2(y), \dots)$.

$$s.t. \rho(\theta, h(y)) \leq \rho(\theta, Dy)$$

$$\sum_{i=1}^P \int_{E_i} (h_i(y) - \theta_i)^2 dy \leq 2 + \sum_{i=1}^P [d_i^2 + \theta_i^2 (d_i - 1)^2]$$

$$\Rightarrow \int_{\mathbb{R}^P} [(h_1(y) - \theta_1)^2 + (h_2(y) - \theta_2)^2] dy \leq 2$$

$\Rightarrow Dy$ is admissible.

Besides, for generally $D = \text{diag}(d_i)$, let

$$d_i \geq 1 \Rightarrow \tilde{d}_i = 1, d_i < 0 \Rightarrow \tilde{d}_i = -d_i.$$

$$d_i = 1 \Rightarrow \tilde{d}_i = \left(1 - \frac{k-2}{\sum_{i=1}^k y_i}\right).$$

s.t $\tilde{D} = \text{diag}(\tilde{d}_i)$, and $\tilde{d}_i = 1$ for at most two i , while
the others satisfy $0 \leq \tilde{d}_i \leq 1$.

By the above proof, we know that $\tilde{D}y$ is admissible

$\Rightarrow Dy$ is admissible

$\Rightarrow Cy$ is admissible.