

# BIOS:7600 Homework 7

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## 1. Problem 5.1, Gaussian tail bound.

*Proof.* First, for any  $t > 0$ ,

$$\begin{aligned}\mathbb{P}(Z \geq \lambda) &\leq \exp(-t\lambda)\mathbb{E}(e^{tZ}) = \frac{e^{-t\lambda}}{\sqrt{2\pi(\sigma^2/n)}} \int_{-\infty}^{\infty} e^{tz} e^{-\frac{z^2}{2\sigma^2/n}} dz \\ &= \frac{e^{-t\lambda}}{\sqrt{2\pi(\sigma^2/n)}} \int_{-\infty}^{\infty} \exp\left(-\frac{(z - \frac{\sigma^2}{n}t)^2}{2\sigma^2/n}\right) \exp\left(\frac{\sigma^2 t^2}{2n}\right) dz \\ &= \exp\left(-t\lambda + \frac{\sigma^2 t^2}{2n}\right).\end{aligned}\tag{1}$$

Let

$$t = \frac{n\lambda}{\sigma} > 0,\tag{2}$$

then the inequality becomes

$$\mathbb{P}(Z \geq \lambda) \leq \exp\left(-\frac{n\lambda^2}{2\sigma^2}\right).\tag{3}$$

By symmetry,

$$\mathbb{P}(Z \leq -\lambda) \leq \exp\left(-\frac{n\lambda^2}{2\sigma^2}\right).\tag{4}$$

Hence we get the result.  $\square$

## 2. Problem 5.3, Prediction bound under RE condition.

(a) *Proof.* First, as proved in class, we have

$$\frac{1}{n}\|X\delta\|_2^2 \leq 3\lambda\sqrt{|\mathcal{S}|}\|\delta_S\|_2.\tag{5}$$

By the restricted eigenvalue condition, we have

$$\frac{1}{n}\delta^\top X^\top X\delta \geq \tau\|\delta\|_2^2 \geq \tau\|\delta_S\|_2^2,\tag{6}$$

so

$$\frac{1}{n}\|X\delta\|_2^2 \leq 3\lambda\sqrt{|\mathcal{S}|}\frac{1}{\sqrt{n\tau}}\|X\delta\|_2.\tag{7}$$

Hence,

$$\frac{1}{n}\|X\delta\|_2^2 \leq \frac{9}{\tau}\lambda^2|\mathcal{S}|.\tag{8}$$

$\square$

(b) *Proof.* Actually, by using the same equality in the last problem, we can get this result quickly.  $\square$

## 3. Problem 6.1, Conditional distribution for random knockoffs.

*Proof.* By the conditional distributions of multivariate normal distribution,

$$(\tilde{x}_i \mid x_i = x) \sim N((\Sigma - S)\Sigma^{-1}x)_i, (\Sigma - (\Sigma - S)\Sigma^{-1}(\Sigma - S))_{ii} \sim N((x - S\Sigma^{-1}x)_i, (2S - S\Sigma^{-1}S)_{ii}).\tag{9}$$

$\square$

## 4. Problem 6.2, Selective inference in the p=2 case.

*Proof.* The condition becomes

$$|\frac{1}{n}x_2^\top(1 - P_1)y + \lambda x_2^\top x_1(x_1^\top x_1)^{-1}| \leq \lambda. \quad (10)$$

By simplification,

$$-\lambda n(x_1^\top x_1 - x_2^\top x_1) \leq (x_1^\top x_1 x_2^\top - x_2^\top x_1 x_1^\top)y \leq \lambda n(x_1^\top x_1 - x_2^\top x_1). \quad (11)$$

Then

$$A = \begin{bmatrix} x_1^\top x_1 x_2^\top - x_2^\top x_1 x_1^\top & 0 \\ 0 & x_2^\top x_1 x_1^\top - x_1^\top x_1 x_2^\top \end{bmatrix}, \quad b = \lambda n \begin{bmatrix} x_1^\top x_1 - x_2^\top x_1 \\ x_1^\top x_1 - x_2^\top x_1 \end{bmatrix}, \quad (12)$$

where  $X = [x_1, x_2]$ .

□

## 5. Problem 6.3, HIV drug resistance study.

(a)