## MATH455(Statistical Learning)- Solution2

魏上傑

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Quiz Problem: Note that this exercise shows that the OLS estimate is essentially the MLE. sol:

We use the following notation to prove the result.

$$X = \begin{bmatrix} 1 & x_{11} & \dots & x_{1p} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_{N1} & \dots & x_{Np} \end{bmatrix} \text{ be a } N \times (p+1) \text{ matrix, } Y \text{ be a } N \times 1 \text{ vector, and } \beta \text{ be a } (p+1) \times 1 \text{ vector.}$$

Notice that

$$\hat{f} = \arg\min_{f \in \mathcal{F}} \frac{1}{N} \sum_{n=1}^N L(y, f(x)) \qquad = \arg\min_{f \in \mathcal{F}} \frac{1}{N} \sum_{n=1}^N -log(p(x, y)) \\ = \arg\max_{f \in \mathcal{F}} \frac{1}{N} \sum_{n=1}^N log(p(x, y))$$

That is, we want to maximize the log-likelihood function.

Also, note that

$$log(p(x,y)) \propto p(x,y) = p(y|x)p(x) \propto p(y|x)$$

$$\begin{split} Y|X \sim N(X^T\beta,\sigma^2) \\ \implies p(Y|X) = \frac{1}{\sqrt{2\pi}\sigma} \exp(\frac{-(Y-X\beta)^2}{2\sigma^2}) \propto -(Y-X\beta)^2 \end{split}$$

Hence, maximize  $p(Y|X) \iff \text{minimize } (Y-X\beta)^2$ 

$$\begin{split} \frac{\partial}{\partial \beta} &= 0 : -2X^T(Y - X\beta) = 0 \\ \implies X^TY - X^TX\beta &= 0 \implies \hat{\beta} = (X^TX)^{-1}X^TY \end{split}$$

**Selected Problems:** 

7.

$$\begin{split} Y &\sim N(X\beta, \sigma^2 I) \\ &\Longrightarrow (X^T X)^{-1} X^T Y \sim N((X^T X)^{-1} X^T X\beta, (X^T X)^{-1} X^T \sigma^2 I X (X^T X)^{-1}) \\ &\Longrightarrow \hat{\beta} \sim N(\beta, \sigma^2 (X^T X)^{-1}) \end{split}$$

We use the fact that  $\sigma^2$  is scalar.

8.

$$\begin{split} ||y - X\beta||^2 &= (y - X\beta)^T (y - X\beta) \\ &= (y^T - \beta^T X^T) (y - X\beta) \\ &= y^T y - y^T X\beta - \beta^T X^T y + \beta^T X^T X\beta \\ &= y^T y - 2y^T X\beta + \beta^T X^T X\beta \end{split}$$

Note that  $y^T X \beta$  is a scalar, so  $y^T X \beta = (y^T X \beta)^T$ 

9.

$$\frac{d}{d\beta}(y^T y) = 0$$

This is because  $y^Ty$  does not depend on  $\beta$ 

10.

$$\frac{d}{d\beta}(y^T X \beta) = X^T y$$

Check the dimension:  $\beta$  is  $(p+1)\times 1$ , and  $X^Ty$  is  $((p+1)\times N)\times (N\times 1)=(p+1)\times 1$ 

11.

$$\frac{d}{d\beta}(\beta^T X^T X \beta) = 2(X^T X)\beta$$

Check the dimension:  $\beta$  is (p+1)x1, and  $(X^TX)\beta$  is  $((p+1)\times N)\times (N\times (p+1))\times ((p+1)\times 1)=(p+1)\times 1$