

# Statistical Inference

## Midterm Examination I

2017/10/16

Time: 1:20 pm – 5:00 pm

1. Consider a linear regression model

$$y_i = \beta_0 + \sum_{j=1}^K \beta_j x_{ij} + \epsilon_i, \quad i = 1, \dots, n, \quad (*)$$

where  $\epsilon_i$ 's are independent and identically distributed (i.i.d.) with mean 0 and positive variance  $\sigma^2$ . Let  $\boldsymbol{\beta} = (\beta_0, \beta_1, \dots, \beta_K)'$ . We are interesting in estimating  $\boldsymbol{\beta}'\boldsymbol{\beta}/\sigma^2$  (which can be regarded as a signal-to-noise ratio), and a natural way to estimate it is

$$\frac{\hat{\boldsymbol{\beta}}'\hat{\boldsymbol{\beta}}}{\hat{\sigma}^2}$$

where  $\hat{\boldsymbol{\beta}} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y}$ ,

$$\mathbf{X} = \begin{pmatrix} 1 & x_{11} & \cdots & x_{1K} \\ \vdots & \vdots & & \vdots \\ 1 & x_{n1} & \cdots & x_{nK} \end{pmatrix}_{n \times (K+1)},$$

$\mathbf{y} = (y_1, \dots, y_n)'$ ,  $\hat{\sigma}^2 = (n - K - 1)^{-1}\mathbf{y}'(\mathbf{I}_n - \mathbf{M})\mathbf{y}$ ,  $\mathbf{I}_n$  is a  $n \times n$  identity matrix and  $\mathbf{M} = \mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'$  is the orthogonal projection matrix onto  $C(\mathbf{X})$ , the column space of  $\mathbf{X}$ .

- (1) Assuming  $\epsilon_1$  follows a normal distribution, compute  $E(\hat{\boldsymbol{\beta}}'\hat{\boldsymbol{\beta}}/\hat{\sigma}^2)$ . Is  $\hat{\boldsymbol{\beta}}'\hat{\boldsymbol{\beta}}/\hat{\sigma}^2$  an unbiased estimator of  $\boldsymbol{\beta}'\boldsymbol{\beta}/\sigma^2$ ? If no, please find an unbiased estimator for  $\boldsymbol{\beta}'\boldsymbol{\beta}/\sigma^2$ .
- (2) Does  $\hat{\boldsymbol{\beta}}'\hat{\boldsymbol{\beta}}/\hat{\sigma}^2$  converge in probability to  $\boldsymbol{\beta}'\boldsymbol{\beta}/\sigma^2$  when there is no normality assumption on  $\epsilon_1$ ? Why?

2. Consider model (\*) with  $K = 1$  and

$$x_{i1} = \begin{cases} 1, & \text{if } i = 1; \\ i^{-1/2}, & \text{if } i = 2, \dots, n. \end{cases}$$

Does  $\hat{\boldsymbol{\beta}}$  converge in probability to  $\boldsymbol{\beta}$ ? Why?

3. Consider model (\*) without normality assumption on the noise.

- (1) Find an asymptotic level 5% test of  $H_0 : \beta_0\beta_1 + \beta_1^2 = d$  versus  $H_a : \sim H_0$ .
- (2) Let  $\mathbf{x}^*$  be a  $(K + 1)$ -dimensional known vector. Assuming  $\epsilon_1$  follows a normal distribution, find a 95% confidence interval for  $\mathbf{x}^{*\prime}\boldsymbol{\beta}$ .
- (3) Construct a 95% confidence interval for  $(\mathbf{x}^{*\prime}\boldsymbol{\beta})^2$  when the normality assumption of the noise holds true. If you can not do that, please find an asymptotic 95% confidence interval for  $(\mathbf{x}^{*\prime}\boldsymbol{\beta})^2$ .

4. If  $X_n \sim \chi^2(n)$  and  $y_n \sim \chi^2(n^2)$ , prove that

$$n^{1/2} \left( \frac{X_n/n}{y_n/n^2} - 1 \right) \xrightarrow{d} N(0, 2).$$

5. Let  $(x_i, y_i)'$ ,  $i = 1, \dots, n$ , be i.i.d. bivariate normal with mean  $(\mu_x, \mu_y)'$  and covariance matrix

$$\begin{pmatrix} \sigma_x^2 & \sigma_{xy} \\ \sigma_{xy} & \sigma_y^2 \end{pmatrix}$$

where  $\sigma_x > 0$  and  $\sigma_y > 0$ . Define

$$\hat{\rho} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2}}.$$

- (1) Find the limiting distribution of  $\sqrt{n}(\hat{\rho} - \rho)$  where  $\rho = \sigma_{xy}/(\sigma_x \sigma_y)$ .  
 (2) Use the result of part (1) to find the limiting distribution of

$$\sqrt{n} \left( \ln \left( \frac{1 + \hat{\rho}}{1 - \hat{\rho}} \right) - \ln \left( \frac{1 + \rho}{1 - \rho} \right) \right)$$

where  $\rho \neq \pm 1$ .

- (3) Use the result of part (2) to establish an asymptotic level 5% test of  $H_0 : \rho = 0$  versus  $H_a : \rho \neq 0$ .
6. Let  $X_1, X_2, \dots, X_n$  be i.i.d. random variables with common probability density function  $f_\theta(x)$ . Find an maximum likelihood estimate (MLE) for  $\theta$  in each of the following cases:
- (a)  $f_\theta(x) = \frac{1}{2}e^{-|x-\theta|}$ ,  $-\infty < x < \infty$ .  
 (b)  $f_\theta(x) = e^{-x+\theta}$ ,  $\theta \leq x < \infty$ .  
 (c)  $f_\theta(x) = (\theta\alpha)x^{\alpha-1}e^{-\theta x^\alpha}$ ,  $x > 0$ , and  $\alpha$  known.

7. Find an MLE, if it exists, in each of the following cases:

- (a)  $X_1, X_2, \dots, X_n \sim N(\theta, \theta^2)$ ,  $\theta \in \mathbb{R}$ .  
 (b)  $X_1, X_2, \dots, X_n$  is a sample from

$$P(X = y_1) = \frac{1 - \theta}{2}, \quad P(X = y_2) = \frac{1}{2}, \quad P(X = y_3) = \frac{\theta}{2} \quad (0 < \theta < 1).$$

- (c)  $X_1, X_2, \dots, X_n \sim N(\theta, \theta)$ ,  $0 < \theta < \infty$ .

8. Let  $X_1, X_2, \dots, X_n$  be a sample from exponential density

$$f_\theta(x) = \theta e^{-\theta x}, \quad x \geq 0, \quad \theta > 0.$$

Find the MLE of  $\theta$ , and show that it is consistent and asymptotically normal.