STAT 260 - Nonparametric Statistics

Ying Sun

Statistics (STAT) Program

Computer, Electrical and Mathematical Sciences & Engineering (CEMSE) Division King Abdullah University of Science and Technology (KAUST)

HOMEWORK

Ι

Henrique Aparecido Laureano

Spring Semester 2018

Contents

| Problem 1 | 2 |
|-----------|---|
| Problem 2 | 3 |
| Problem 3 | 4 |
| Problem 4 | 4 |
| Problem 5 | 4 |
| Problem 6 | 4 |

Problem 1

Question Q1.1-1.3 on page 15 of Topic 2.

Q1.1

$$\mathbb{V}\hat{\beta}_0 = \sigma^2 \frac{\sum_{i=1}^n x_i^2}{nS_{xx}} ?$$

Solution:

$$\mathbb{V}\hat{\beta}_0 = \mathbb{V}(\bar{y} - \hat{\beta}_1 \bar{x}) = \mathbb{V}\bar{y} + \bar{x}^2 \mathbb{V}\hat{\beta}_1 - 2\bar{x} \operatorname{Cov}(\bar{y}, \hat{\beta}_1).$$

The variance terms are

$$\mathbb{V}\bar{y} = \frac{1}{n^2} \sum_{i=1}^n \mathbb{V} y_i = \frac{\sigma^2}{n}, \qquad \mathbb{V}\hat{\beta}_1 = \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{\sigma^2}{S_{xx}}.$$

The covariance term is

$$Cov(\bar{y}, \hat{\beta}_1) = \mathbb{E}\bar{y}\hat{\beta}_1 - \mathbb{E}\bar{y}\mathbb{E}\hat{\beta}_1 = \beta_1(\beta_0 + \beta_1\bar{x}) - (\beta_0 + \beta_1\bar{x})\beta_1 = 0.$$

So

$$\begin{split} \mathbb{V}\hat{\beta}_{0} &= \mathbb{V}(\bar{y} - \hat{\beta}_{1}\bar{x}) = \mathbb{V}\bar{y} + \bar{x}^{2}\mathbb{V}\hat{\beta}_{1} - 2\bar{x}\mathrm{Cov}(\bar{y}, \hat{\beta}_{1}) \\ &= \frac{\sigma^{2}}{n} + \frac{\bar{x}^{2}\sigma^{2}}{S_{xx}} - 0 = \sigma^{2}\frac{S_{xx} + n\bar{x}^{2}}{nS_{xx}} = \sigma^{2}\frac{\sum_{i=1}^{n}(x_{i} - \bar{x})^{2} + n\bar{x}^{2}}{nS_{xx}} \\ &= \sigma^{2}\frac{\sum_{i=1}^{n}x_{i}^{2} - n\bar{x}^{2} + n\bar{x}^{2}}{nS_{xx}} = \sigma^{2}\frac{\sum_{i=1}^{n}x_{i}^{2}}{nS_{xx}}. \end{split}$$

Therefore

$$\boxed{ \mathbb{V}\hat{\beta}_0 = \sigma^2 \frac{\sum_{i=1}^n x_i^2}{nS_{xx}}.}$$

Q1.2

$$Cov(\hat{\beta}_0, \hat{\beta}_1) = -\sigma^2 \frac{\bar{x}}{S_{xx}} ?$$

Solution:

$$\operatorname{Cov}(\hat{\beta}_{0}, \hat{\beta}_{1}) = \mathbb{E}(\hat{\beta}_{0} - \mathbb{E}\hat{\beta}_{0})(\hat{\beta}_{1} - \mathbb{E}\hat{\beta}_{1}) = \mathbb{E}(\hat{\beta}_{0} - \beta_{0})(\hat{\beta}_{1} - \beta_{1})$$

$$= \mathbb{E}(\bar{y} - \hat{\beta}_{1}\bar{x} - \beta_{0})(\hat{\beta}_{1} - \beta_{1})$$

$$= \mathbb{E}(\beta_{0} + \beta_{1}\bar{x} - \hat{\beta}_{1}\bar{x} - \beta_{0})(\hat{\beta}_{1} - \beta_{1})$$

$$= \mathbb{E}(-(\hat{\beta}_{1} - \beta_{1})\bar{x})(\hat{\beta}_{1} - \beta_{1})$$

$$= -\bar{x}\mathbb{E}(\hat{\beta}_{1} - \beta_{1})^{2}$$

$$= -\bar{x}\mathbb{E}(\hat{\beta}_{1}^{2} - 2\beta_{1}\hat{\beta}_{1} + \beta_{1}^{2})$$

$$= -\bar{x}[\mathbb{E}\hat{\beta}_{1}^{2} - 2\beta_{1}\mathbb{E}\hat{\beta}_{1} + \beta_{1}^{2}]$$

$$= -\bar{x}[\mathbb{V}\hat{\beta}_{1} + \mathbb{E}^{2}\hat{\beta}_{1} - \beta_{1}^{2}]$$

$$= -\bar{x}\left[\frac{\sigma^{2}}{S_{xx}} + \beta_{1}^{2} - \beta_{1}^{2}\right].$$

Therefore,

$$Cov(\hat{\beta}_0, \hat{\beta}_1) = -\frac{\sigma^2 \bar{x}}{S_{xx}}.$$

Q1.3

- What does mean $\mathbb{V}\hat{\beta}_1 \neq 0$?
- What is the source of randomness in $\hat{\beta}_1$?
- How can you reduce $\mathbb{V}\hat{\beta}_1$?

Solution:

Problem 2

Question Q2.1-2.2 on pages 16-17 of Topic 2.

Solution:

| Problem 3 |
|--|
| Question Q3 on page 19 of Topic 2. |
| Solution: |
| Problem 4 |
| Question Q4 on page 20 of Topic 2. |
| Solution: |
| Problem 5 |
| Write your own version of anova() function on page 14 of Topic 3 using R. Your function should not use lm() function. Must be able to compare two nested models. Replicate results on slides 14, 17, 18, 21 and 22 about the gala data set. |
| Solution: |
| Problem 6 |
| Write an R function that can be used to automatically determine the order d of the polynomial regression and replicate the results on pages 31 and 32 of Topic 3 about the savings data. (The output should be the chosen order d and the plot of the fitted polynomial function to the data.) |
| Solution: |
| |
| |