STAT 260 - Nonparametric Statistics
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HOMEWORK
I

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Problem 1

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

Q1.1

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

Solution:

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

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

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

So

$$\begin{aligned}
\mathbb{V}\hat{\beta}_0 = \mathbb{V}(\bar{y} - \hat{\beta}_1\bar{x}) &= \mathbb{V}\bar{y} + \bar{x}^2\mathbb{V}\hat{\beta}_1 - 2\bar{x}\mathrm{Cov}(\bar{y}, \hat{\beta}_1) \\
&= \frac{\sigma^2}{n} + \frac{\bar{x}^2\sigma^2}{S_{xx}} - 0 = \sigma^2\frac{S_{xx} + n\bar{x}^2}{nS_{xx}} = \sigma^2\frac{\sum_{i=1}^n (x_i - \bar{x})^2 + n\bar{x}^2}{nS_{xx}} \\
&= \sigma^2\frac{\sum_{i=1}^n x_i^2 - n\bar{x}^2 + n\bar{x}^2}{nS_{xx}} = \sigma^2\frac{\sum_{i=1}^n x_i^2}{nS_{xx}}.
\end{aligned}$$

Therefore

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

□

Q1.2

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

Solution:

$$
\begin{aligned}
\mathrm{Cov}(\hat{\beta}_0, \hat{\beta}_1) = \mathbb{E}(\hat{\beta}_0 - \mathbb{E}\hat{\beta}_0)(\hat{\beta}_1 - \mathbb{E}\hat{\beta}_1) &= \mathbb{E}(\hat{\beta}_0 - \beta_0)(\hat{\beta}_1 - \beta_1) \\
&= \mathbb{E}(\bar{y} - \hat{\beta}_1\bar{x} - \beta_0)(\hat{\beta}_1 - \beta_1) \\
&= \mathbb{E}(\beta_0 + \beta_1\bar{x} - \hat{\beta}_1\bar{x} - \beta_0)(\hat{\beta}_1 - \beta_1) \\
&= \mathbb{E}(-(\hat{\beta}_1 - \beta_1)\bar{x})(\hat{\beta}_1 - \beta_1) \\
&= -\bar{x}\mathbb{E}(\hat{\beta}_1 - \beta_1)^2 \\
&= -\bar{x}\mathbb{E}(\hat{\beta}_1^2 - 2\beta_1\hat{\beta}_1 + \beta_1^2) \\
&= -\bar{x}[\mathbb{E}\hat{\beta}_1^2 - 2\beta_1\mathbb{E}\hat{\beta}_1 + \beta_1^2] \\
&= -\bar{x}[\mathbb{V}\hat{\beta}_1 + \mathbb{E}^2\hat{\beta}_1 - \beta_1^2] \\
&= -\bar{x}\left[\frac{\sigma^2}{S_{xx}} + \beta_1^2 - \beta_1^2\right].
\end{aligned}
$$

Therefore,

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

□

Q1.3

What does mean $\mathbb{V}\hat{\beta}_1 \neq 0$?

Solution:

We don't know the real slope, β_1, so $\hat{\beta}_1$ is an estimative with mean β_1 and with a respective variance. If the variance of $\hat{\beta}_1$ is zero this means that, is this case, $\hat{\beta}_1$ is exactly equal to β_1.

□

What is the source of randomness in $\hat{\beta}_1$?

Solution:

□

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:

■