Chapter 2: Linear Regression

Distribution of the RSS Values and Hypothesis Testing for 수정하기

JuHyun, Kang

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The Three Sisters of Newton School of Mathematics, Statistics and Data Science Sungshin Women's University

Outline

1 Distribution of the RSS Values

 ${\color{red} {\bf 2}}$ Hypothesis Testing for $\hat{\beta}_j \neq 0$

Hat matrix

• Hat matrix defined by $\hat{y} = Hy$

$$\hat{y} = X\hat{\beta} = X(X^TX)^{-1}X^Ty = Hy$$

$$H \triangleq X(X^TX)^{-1}X^T$$

• Some properties

$$\begin{split} H^2 &= X(X^TX)^{-1}X^T \cdot X(X^TX)^{-1}X^T = X(X^TX)^{-1}X^T = H \\ (I-H)^2 &= I-2H+H^2 = I-H \\ HX &= X(X^TX)^{-1}X^T \cdot X = X \end{split}$$

Residual sum of square

• RSS defined

$$RSS \triangleq ||y - \hat{y}||^2$$

• Using hat matrix

$$y-\hat{y}=y-Hy=(I-H)y=(I-H)(X\beta+\varepsilon)$$

$$=(X-HX)\beta+(I-H)\varepsilon=(I-H)\varepsilon$$

$$\mathrm{RSS} \triangleq ||y - \hat{y}||^2 = \{(I - H)\varepsilon\}^T (I - H)\varepsilon = \varepsilon^T (I - H)^2 \varepsilon = \varepsilon^T (I - H)\varepsilon$$

Eigenvalues H and I-H

- They are only zeros and ones
- Dimensions of the eigenspaces of H and I-H are both p+1

Proof using
$$rank(X) = p + 1$$

$$\begin{aligned} & \operatorname{rank}(H) \leq \min\{\operatorname{rank}(X(X^TX)^{-1}), \operatorname{rank}(X)\} \leq \operatorname{rank}(X) = p+1 \\ & \operatorname{rank}(H) \geq \operatorname{rank}(HX) = \operatorname{rank}(X) = p+1 \end{aligned}$$

Hat matrix

• Hat matrix defined by

$$\hat{y} = X \hat{\beta} = X (X^T X)^{-1} X^T y = H y$$

$$H \triangleq X (X^T X)^{-1} X^T$$

• some properties

$$\begin{split} H^2 &= X(X^TX)^{-1}X^T \cdot X(X^TX)^{-1}X^T = X(X^TX)^{-1}X^T = H \\ (I-H)^2 &= I-2H+H^2 = I-H \\ HX &= X(X^TX)^{-1}X^T \cdot X = X \end{split}$$

Residual sum of square

• RSS defined

$$\mathrm{RSS} \triangleq ||y - \hat{y}||^2$$

• Using hat matrix

$$\begin{split} y - \hat{y} &= y - Hy = (I - H)y = (I - H)(X\beta + \varepsilon) \\ &= (X - HX)\beta + (I - H)\varepsilon = (I - H)\varepsilon \end{split}$$

$$RSS \triangleq ||y - \hat{y}||^2 = \{(I - H)\varepsilon\}^T (I - H)\varepsilon = \varepsilon^T (I - H)^2 \varepsilon = \varepsilon^T (I - H)\varepsilon$$

Eigenvalues of H and Null space of (I - H)

• Proof by contrapositive

$$Hx = x \Rightarrow (I - H)x = 0$$

 $(I - H)x = 0 \Rightarrow Hx = x$

• Dimensions of the eigenspaces of H is p+1

$$\mathbf{Proof} \text{ using } \mathrm{rank}(X) = p+1$$

$$\operatorname{rank}(H) \leq \min \{ \operatorname{rank}(X(X^TX)^{-1}), \operatorname{rank}(X) \} \leq \operatorname{rank}(X) = p+1$$

$$\operatorname{rank}(H) > \operatorname{rank}(HX) = \operatorname{rank}(X) = p+1$$

 \bullet Dimensions of the null space of I-H is N-(p+1)

$$P(I-H)P^T = \mathrm{diag}(\underbrace{1,\ldots,1}_{N-p-1},\underbrace{0,\ldots,0}_{p+1})$$

제목 뭐라고 하지..

• We define $v = P\varepsilon \in \mathbb{R}^N$, then from $\varepsilon = P^T v$

$$\begin{split} \text{RSS} &= \varepsilon^T (I - H) \varepsilon = (P^T v)^T (I - H) P^T v = v^T P (I - H) P^T v \\ &= [v_1, \cdots, v_{N-p-1}, v_{N-p}, \cdots, v_n] \begin{bmatrix} 1 & 0 & \cdots & \cdots & \cdots & 0 \\ 0 & \ddots & 0 & \cdots & \cdots & \vdots \\ \vdots & 0 & 1 & \cdots & \cdots & 0 \\ 0 & 0 & 0 & \cdots & \cdots & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & \cdots & \cdots & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ \vdots \\ v_{N-p-1} \\ v_{N-p} \\ \vdots \\ v_N \end{bmatrix} \\ &= \sum_{i=1}^{N-p-1} v_i^2 \end{split}$$

• Let
$$w \in \mathbb{R}^{N-p-1}$$

Average

$$E[v] = E[P\varepsilon] = 0$$
$$E[w] = 0$$

Covariance

$$\begin{split} E[vv^t] &= E[P\varepsilon(P\varepsilon)^T] = PE[\varepsilon\varepsilon^t]P = P\sigma^2IP^T = \sigma^2I\\ E[ww^T] &= \sigma^2I \end{split}$$

• We have RSS

$$\frac{RSS}{\sigma^2} \sim \chi^2_{N-p-1}$$

Outline

Distribution of the RSS Values

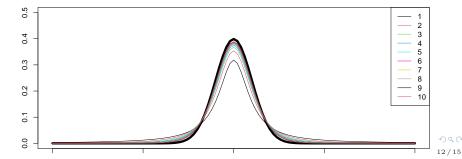
 ${\color{red} {\bf 2}}$ Hypothesis Testing for $\hat{\beta}_j \neq 0$

Test statistic T

- A t distribution with N-P-1 degrees of freedom
- We decide that hypothesis $\beta_i = 0$ should be rejected.
- $U \sim N(0,1), \ V \sim \chi_m^2$

$$T \triangleq U/\sqrt{V/m}$$

```
curve(dnorm(x), -10, 10, ann = FALSE, ylim = c(0, 0.5), lwd = 5)
for(i in 1:10)curve(dt(x, df= i), -10, 10, col = i, add = TRUE, ann = FALSE)
legend("topright", legend = 1:10, lty = 1, col = 1:10)
```



Significance level

•
$$\alpha = 0.01, 0.05$$

• Reject the null hypothesis

Q & A

Thank you:)