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# 고급회귀분석론

## 이영미 교수님

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|-------|----------------|
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| 과 제 명 | HW 02          |
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$$1. SSE = \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

$$= (y - \hat{y})^T (y - \hat{y})$$

$$= (y - Xb)^T (y - Xb)$$

$$= y^T y - 2b^T X^T y + b^T X^T X b$$

$$* b = (X^T X)^{-1} X^T y \text{ o1e2}$$

$$= y^T y - y^T X (X^T X)^{-1} X^T y$$

$$= y^T \underbrace{(I - X(X^T X)^{-1} X^T)}_{\substack{\text{B} \\ \text{B}}}$$

$$= y^T B y$$

$$E(SSE) = E(y^T [I - X(X^T X)^{-1} X^T] y)$$

$$= E(y^T B y) = \sigma^2 \text{tr}(B) + \beta^T X^T B X \beta$$

$$\text{tr}(B) = \text{tr}(I - X(X^T X)^{-1} X^T)$$

$$= \text{tr}(I) - \text{tr}(X(X^T X)^{-1} X^T)$$

$$= n - (K+1) = n - K - 1 \text{ o1e2}$$

$$X^T B X = X^T [I - X(X^T X)^{-1} X^T] X = 0 \text{ o1e2}$$

$$E(SSE) = (n - K - 1) \sigma^2,$$

2. 오차제곱합  $S = (y_1 - \beta_0)^2 + (y_2 - 2\beta_0 + \beta_1)^2 + (y_3 - \beta_0 - 2\beta_1)^2$

$$\frac{\partial S}{\partial \beta_0} = -2(y_1 - \beta_0) - 4(y_2 - 2\beta_0 + \beta_1) - 2(y_3 - \beta_0 - 2\beta_1) = 0$$

$$= y_1 - \beta_0 + 2y_2 - 4\beta_0 + 2\beta_1 + y_3 - \beta_0 - 2\beta_1 = 0$$

$$\therefore \beta_0 = \frac{y_1 + 2y_2 + y_3}{6}$$

$$\frac{\partial S}{\partial \beta_1} = 0 + 2(y_2 - 2\beta_0 + \beta_1) - 4(y_3 - \beta_0 - 2\beta_1) = 0$$

$$\therefore \beta_1 = \frac{y_2 - 2y_3}{5}$$

잔차제곱합은  $S$ 에  $\beta_0, \beta_1$  대입

$$\text{잔차제곱합} = \left( y_1 - \frac{y_1 + 2y_2 + y_3}{6} \right)^2 + \left( y_2 - \frac{y_1 + 2y_2 + y_3}{3} + \frac{y_2 - 2y_3}{5} \right)^2$$

$$+ \left( y_3 - \frac{y_1 + 2y_2 + y_3}{6} - \frac{2}{5} \times (y_2 - 2y_3) \right)^2$$

$$= \left( \frac{5y_1 - 2y_2 - y_3}{6} \right)^2 + \left( \frac{-5y_1 + 8y_2 + 11y_3}{15} \right)^2 + \left( \frac{-5y_1 - 22y_2 + 49y_3}{30} \right)^2$$

3. 단순회귀분석  $y_i = \beta_0 + \beta_1 x_i + \varepsilon_i$ ,  $\varepsilon_i \sim N(0, \sigma^2)$  IID 일때

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} \text{이며, } \hat{\beta}_1 = \frac{S(xy)}{S(xx)} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}$$

$$S_{y.x}^2 = \frac{\sum (y_i - \hat{y}_i)^2}{n-2}, R^2 = \frac{SSR}{SST} = \frac{(S(xy))^2}{S(xx)S(yy)}, \text{ t검정 } T = \frac{\hat{\beta}_1 - \beta_1^0}{\hat{\sigma} / \sqrt{S(xx)}}$$

$x_i$ 가  $Cx_i$ 로 대체된다면,

$$\hat{\beta}_1 = \frac{\sum (Cx_i - \overline{Cx_i})(y_i - \bar{y}_i)}{\sum (Cx_i - \overline{Cx_i})^2} = \frac{C S(xy)}{C^2 S(xx)} = \frac{1}{C} \frac{S(xy)}{S(xx)} \text{ 이다.}$$

$$\hat{\beta}_0 = \bar{y} - \frac{1}{C} \frac{S(xy)}{S(xx)} \cdot \overline{Cx} = \bar{y} - \hat{\beta}_1 \bar{x} \text{ 이므로 변하지 않는다.}$$

$S_{y.x}$ 도 변하지 않는다.

$$R^2 = \frac{SSR}{SST} = \frac{(S(xy))^2}{S(xx)S(yy)} = \frac{[\sum (Cx_i - \overline{Cx_i})(y_i - \bar{y}_i)]^2}{\sum (Cx_i - \overline{Cx_i})^2 \cdot \sum (y_i - \bar{y}_i)^2} = \frac{S(xy)}{S(xx)S(yy)} \text{ 이므로, } R^2 \text{도 변하지 않는다.}$$

t검정결과값은 분자  $\frac{1}{C}$ 배, 분모  $\frac{1}{\sqrt{C^2}}$ 배이므로, 변하지 않는다....

$$4. (1) H_0: y = \beta_0 + \beta_1 x + \varepsilon \quad \text{vs} \quad H_1: y \neq \beta_0 + \beta_1 x + \varepsilon$$

$$\sum x_i = 0 + 0 + 3 + \dots + 12 = 60$$

$$\sum x_i^2 = 0^2 + 0^2 + 3^2 + \dots + 12^2 = 540$$

$$\sum x_i y_i = 0 \times 8.5 + \dots + 12 \times 6.7 = 431.1$$

$$\sum y_i = 8.5 + 8.4 + \dots = 76.1$$

$$\bar{x} = 6, \quad \bar{y} = 7.511$$

$$\begin{aligned} \hat{\beta}_1 &= \frac{S_{xy}}{S_{xx}} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} = \frac{\sum x_i y_i - \frac{1}{n} \sum x_i \sum y_i}{\sum x_i^2 - \frac{1}{n} (\sum x_i)^2} \\ &= \frac{431.1 - \frac{1}{10} \times 60 \times 76.1}{540 - \frac{1}{10} \times 60^2} = -0.1417 \end{aligned}$$

$$\hat{\beta}_0 = \bar{y} - \bar{x} \hat{\beta}_1 = \frac{1}{10} \times 76.1 + 0.1417 \times 60 \times \frac{1}{10} = 8.4584$$

$$\hat{y} = 8.46 - 0.1417x$$

$$\begin{aligned} SSE &= \sum_{i=1}^K \sum_{j=1}^{n_i} (y_{ij} - \hat{y}_i)^2 = (8.5 - 0.04)^2 + (8.4 + 0.06)^2 + \dots + (6.7 + 0.0596)^2 \\ &\doteq 0.1165 \end{aligned}$$

$$\begin{aligned} SSPE &= \sum \sum (y_{ij} - \bar{y}_i)^2 = (8.5 - 8.45)^2 + (8.4 - 8.45)^2 + \dots + (6.7 - 6.75)^2 \\ &\doteq 0.095 \end{aligned}$$

$$SSLF = SSE - SSPE = 0.1165 - 0.095 = 0.0215$$

$$F_0 = \frac{\frac{SSLF}{df_L}}{\frac{SSPE}{df_P}} = \frac{\frac{0.0215}{3}}{\frac{0.095}{5}} \doteq 0.377 < 5.41 = F(3, 5, 0.95) \text{ 이므로}$$

$H_0$ 를 채택한다. 이귀측검은 타당하다.

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AA

BA

$$\hat{y}_{1j} = 6.756 + 0.2811x_{1j}$$

$$\hat{y}_{2j} = 12.85714 + 0.145x_{2j}$$

ANOVA

| 종   | 제곱합     | 자유도 |
|-----|---------|-----|
| SSR | 221.203 | 1   |
| SSE | 5.56    | 5   |
| SST | 226.76  | 6   |

ANOVA

| 종   | 제곱합      | 자유도 |
|-----|----------|-----|
| SSR | 58.87    | 1   |
| SSE | 8.5071   | 5   |
| SST | 67.37714 | 6   |

$$SSE(F) = 5.56 + 8.5071 = 14.0671$$

$$df_F = 5 + 5 = 10$$

중요한 변수를 찾기 위해

$$\sum x_i = 560, \quad \sum x_i^2 = 28000$$

$$\sum y_i = 256.6, \quad \sum x_i y_i = 11457$$

$$\hat{\beta}_1 = \frac{11457 - 560 \times 256.6 \times \frac{1}{14}}{28000 - 560^2 \times \frac{1}{14}} = 0.213036$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = 18.32857 - 0.21306 \times 40 = 9.80713$$

$$\hat{y}_{ij} = 9.807 + 0.213 \hat{x}_{ij}$$

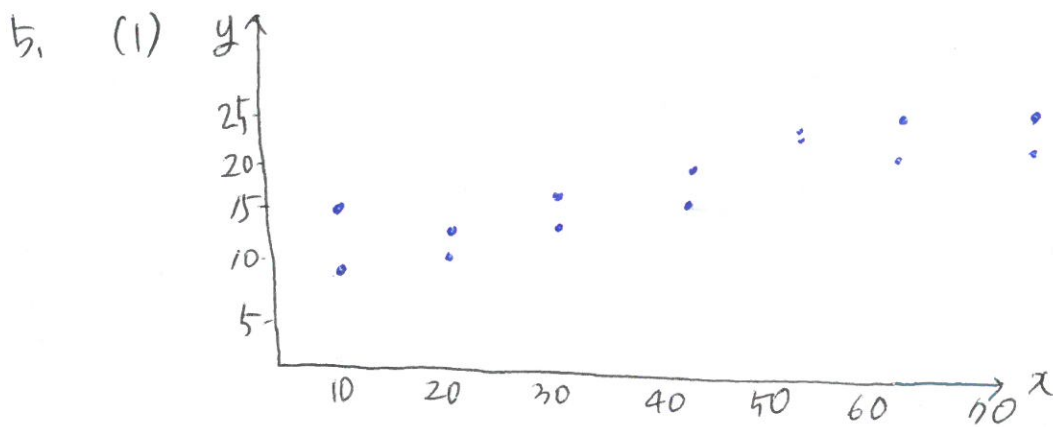
ANOVA

| 종   | 제곱합      | 자유도 |
|-----|----------|-----|
| SSR | 254.1516 | 1   |
| SSE | 41.49696 | 12  |
| SST | 295.6486 | 13  |

$$\begin{cases} SST = 4998.76 - \frac{1}{14} \times 256.6^2 = 295.6486 \\ SSR = \frac{1193^2}{5600} = 254.1516 \\ SSE = SST - SSR = 41.49696 \end{cases}$$

$$F_0 = \frac{41.49696 - 14.0671}{12 - 10} = \frac{14.0671}{10} = 9.74965 > F(2, 10; 0.95) = 4.10$$

이므로,  $H_0$ 를 기각한다. 즉 두 변수는 동일하지 않다.



(2)  $H_0: \beta_{01} = \beta_{02}$  and,  $\beta_{11} = \beta_{12}$  vs  $H_0: \beta_{01} \neq \beta_{02}$  or,  $\beta_{11} \neq \beta_{12}$ .

$$\sum x_i = 280 \quad \sum y_{1j} = 126 \quad \sum x_i y_{1j} = 5827$$

$$\sum x_i^2 = 14000 \quad \sum y_{2j} = 130.6 \quad \sum x_i y_{2j} = 5630$$

At 1% level  $\hat{y}_{1j} = 6.756 + 0.2811 x_{1j}$

$$\hat{\beta}_{11} = \frac{S(xy)}{S(xx)} = \frac{5827 - \frac{1}{n} \times 126 \times 280}{14000 - 280^2 \times \frac{1}{n}} \div 0.2811$$

$$\hat{\beta}_{01} = \bar{y}_1 - \hat{\beta}_{11} \bar{x} = 18 - 0.2811 \times 40 \div 6.756$$

At 1% level  $\hat{y}_{2j} = 12.857 + 0.145 x_{2j}$

$$\hat{\beta}_{12} = \frac{S(xy)}{S(xx)} = \frac{5630 - \frac{1}{n} \times 130.6 \times 280}{14000 - 280^2 \times \frac{1}{n}} \div 0.145$$

$$\hat{\beta}_{02} = \bar{y}_2 - \hat{\beta}_{12} \bar{x} = 18.65714 - 0.145 \times 40 \div 12.85714$$

At SST =  $\sum (y_i - \bar{y})^2 = \sum y_i^2 - \frac{1}{n} (\sum y_i)^2$

$$\div 2494.76 - \frac{1}{n} (126)^2 \div 226.76$$

At SSR =  $\sum (\bar{y} - \hat{y}_i)^2 = \frac{(S(xy))^2}{S(xx)} = \frac{787^2}{2800} = 221.203$

At SSE = SST - SSR = 5.56

At SST =  $\sum (y_i - \bar{y})^2 = 2504 - \frac{1}{n} (130.6)^2 = 67.37714$

At SSR =  $\sum (\bar{y} - \hat{y}_i)^2 = \frac{(S(xy))^2}{S(xx)} = \frac{406^2}{2800} \div 58.87$

At SSE = SST - SSR = 67.37714 - 58.87 \div 8.5071

$$5.(3) \quad H_0: \beta_{11} = \beta_{12} \quad \text{vs} \quad H_1: \beta_{11} \neq \beta_{12}$$

가설검정계량

$$t_0 = \frac{\hat{\beta}_{11} - \hat{\beta}_{12}}{\sqrt{\widehat{\text{var}}(\hat{\beta}_{11} - \hat{\beta}_{12})}} \sim_{H_0} t((n_1-2) + (n_2-2))$$

$$\hat{\beta}_{11} = 0.2811, \quad \hat{\beta}_{12} = 0.145$$

$$SSE(F) = 14.0671, \quad MSE(F) = \frac{SSE(F)}{df_F} = \frac{14.0671}{10} = 1.40671$$

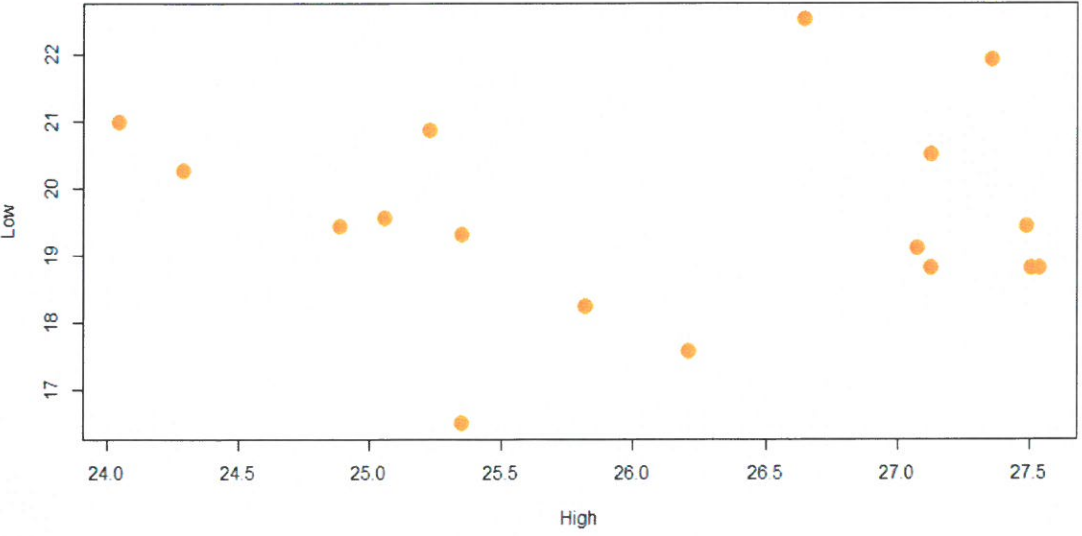
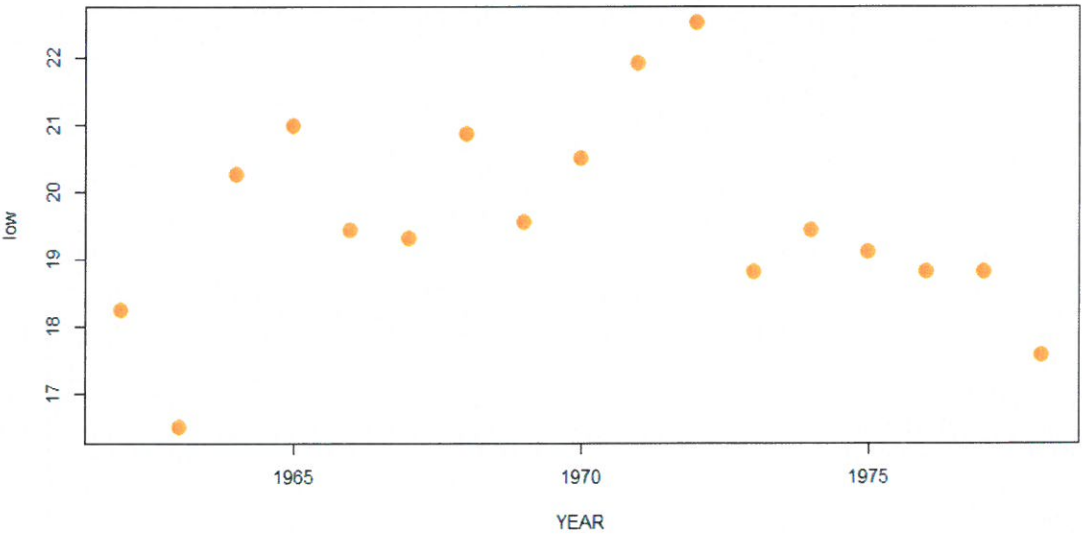
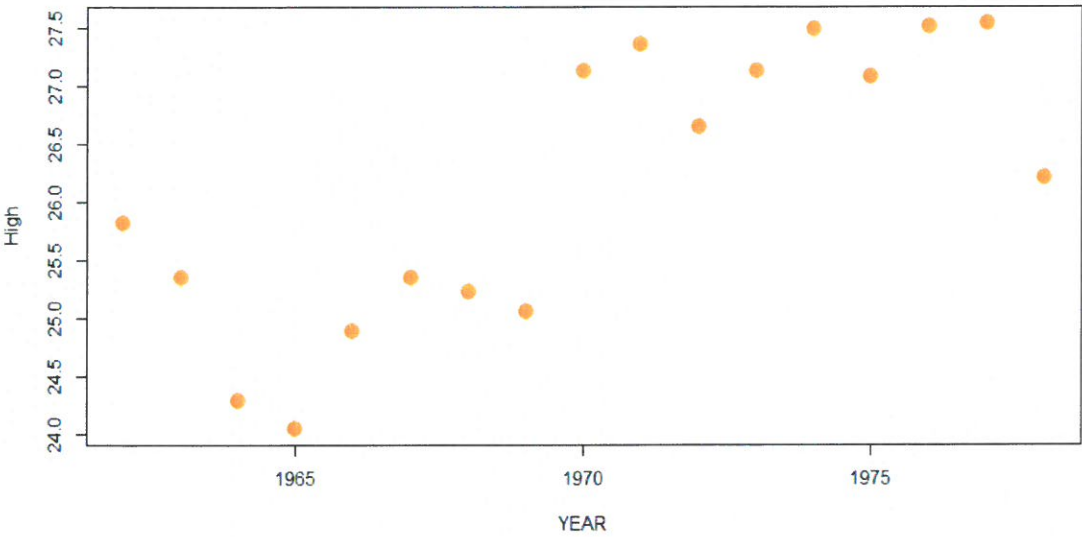
$$\widehat{\text{var}}(\hat{\beta}_{11} - \hat{\beta}_{12}) = 1.40671 \left[ \frac{1}{2800} + \frac{1}{2800} \right] = 0.001005$$

$$t_0 = \frac{0.2811 - 0.145}{\sqrt{0.001005}} \doteq 4.293583$$

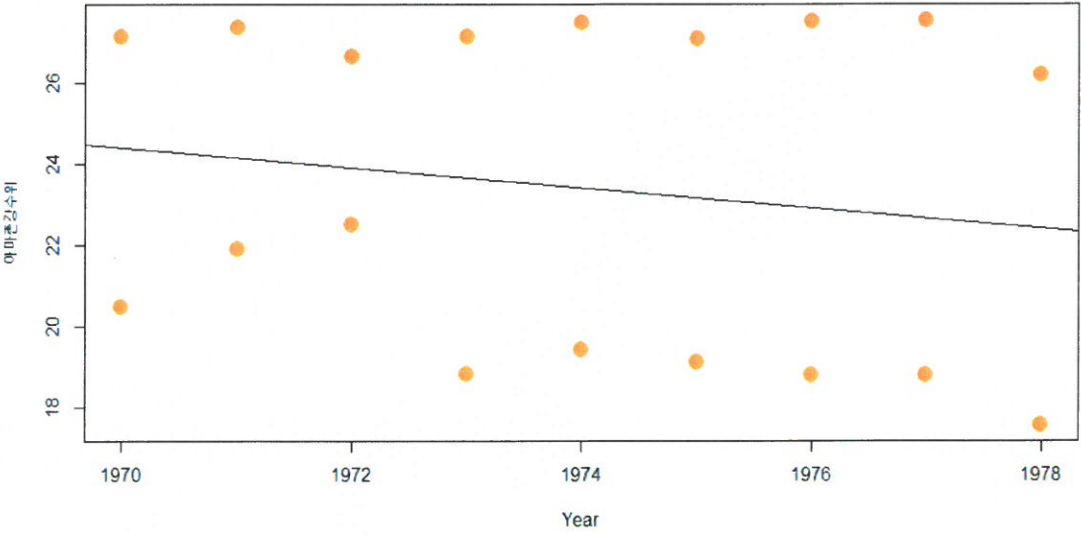
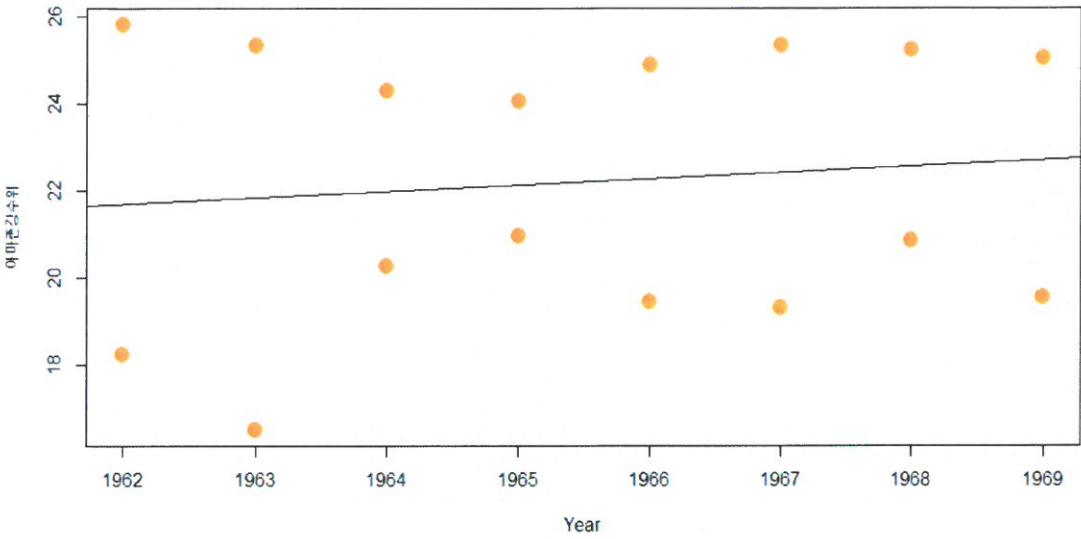
$|t_0| > t_{0.025}(10) = 2.228$  이므로  $H_0$ 를 기각한다.  
즉, 기원기초 값이 다르다.



6. (1)



6. (4)



6. (2) Year에 대한 High의 회귀분석 : Year가 1만큼 증가하면 문의수(High)는 0.180882만큼 증가한다.

$$\text{High} = -330.2124 + 0.180882 \cdot \text{Year}$$

Year에 대한 Low의 회귀분석 : Year가 1만큼 증가하면 문의수(Low)는 0.007892만큼 감소한다.

$$\text{Low} = 35.106961 - 0.007892 \cdot \text{Year}$$

Low에 대한 High의 회귀분석

$$\text{High} = 26.40088 - 0.01406 \cdot \text{Low}$$

(3) 아마존 검색 수의 변화를 예측하는 데 불충족.  $\rightarrow$  선형관계가 없다.

(4)  $H_0: \beta_{01} = \beta_{02}, \beta_{11} = \beta_{12}$  vs  $H_1: \beta_{01} \neq \beta_{02}$  or  $\beta_{11} \neq \beta_{12}$

1960년대 레지자수  $\hat{y}_{1j} = -265.1411 + 0.1462 \hat{x}_{1j}$

ANOVA

| 종   | 계산값      | 자유도 |
|-----|----------|-----|
| SSR | 1.795    | 1   |
| SSE | 141.912  | 14  |
| SST | 143.7076 | 15  |

1970년대 레지자수  $\hat{y}_{2j} = 510.5017 - 0.2467 \hat{x}_{2j}$

ANOVA

| 종   | 계산값     | 자유도 |
|-----|---------|-----|
| SSR | 7.306   | 1   |
| SSE | 262.144 | 16  |
| SST | 269.45  | 17  |

Step 1:  $SSE(F) = SSE_1 + SSE_2 = 141.912 + 262.144 = 403.336$   
 $df_F = 14 + 16 = 30$

측량값의 회귀식  $\hat{y} = -147.5527 + 0.0869 \hat{x}$

| ANOVA | 회귀     | 잔차 | 총 |
|-------|--------|----|---|
| SSR   | 6.10   | 1  |   |
| SSE   | 419.68 | 32 |   |
| SST   | 425.78 | 33 |   |

Step 2:  $SSE(R) = 419.68$

$df_R = 32$

Step 3: 검정통계량  $F_0 = \frac{419.68 - 403.336}{32 - 30} \div \frac{403.336}{30} = 0.6078307$

$F_0 < F(2, 32; 0.01) = 5.336$  이므로  $H_0$  채택

즉, 두 회귀모델은 동일하다.

(b)  $H_0: \beta_{11} = \beta_{12}$  vs  $H_1: \beta_{11} \neq \beta_{12}$

$\hat{\beta}_{11} = 0.1462, \hat{\beta}_{12} = -0.2467$

$SSE(F) = 403.336, MSE(F) = \frac{403.336}{30} = 13.44453$

$\widehat{Var}(\hat{\beta}_{11} - \hat{\beta}_{12}) = 13.4445 \left( \frac{1}{24} + \frac{1}{120} \right) = 0.272092$

검정통계량:  $t_0 = \frac{0.1462 - (-0.2467)}{\sqrt{0.272092}} = 0.753224$

$|t_0| < t_{0.005}(30) = 2.750$  이므로 유의수준 1%에서

귀무가설을 기각하지 못한다.