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Let
$$k=2$$
, and $Y_{\lambda} = b_1 + b_2 X_{\lambda} + e_{\lambda}$, $b_2 = \frac{\sum (\chi_{\lambda} - \overline{\chi})(y_{\lambda} - \overline{y})}{\sum (\chi_{\lambda} - \overline{\chi})^2}$, $b_1 = \overline{y} - b_2 \overline{\chi}$

$$\chi = \begin{cases} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} & = \begin{cases} \begin{pmatrix} r_1 \\ r_2 \\ r_3 \\ \vdots \\ r_n \end{pmatrix} = \chi \times = \begin{cases} 1 & 1 & \dots & 1 \\ 1 & x_n \\ \vdots & \vdots \\ 1 & x_n \end{cases} = \begin{cases} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \\ \vdots \\ x_n \end{pmatrix} = \begin{cases} x_1 \\ x_2 \\ \vdots \\ x_n \end{cases} = \begin{cases} x_$$

$$\left(\mathsf{X}'\mathsf{X}\right)^{-1} = \frac{1}{\mathsf{N}\mathsf{\Sigma}\mathsf{X}_{\mathsf{A}}^{2} - \mathsf{c}\mathsf{\Sigma}\mathsf{X}_{\mathsf{A}}^{2}} \left\{ \frac{\mathsf{\Sigma}\mathsf{X}_{\mathsf{A}}^{2} - \mathsf{\Sigma}\mathsf{X}_{\mathsf{A}}}{-\mathsf{\Sigma}\mathsf{X}_{\mathsf{A}}} \right\}$$

$$50 b = (\cancel{X}\cancel{X})^{1}(\cancel{X}\cancel{Y}) = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\lambda} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \\ - \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}} \begin{bmatrix} \cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2} \end{bmatrix} = \frac{1}{\cancel{\Sigma} \cancel{X_{\lambda}^{2}} - (\cancel{\Sigma} \cancel{X_{\lambda}^{2}})^{2}}$$

$$b_2 = \frac{h^{2} \chi_{i} \chi_{i} - i \chi_{i} \chi_{i}}{h^{2} \chi_{i}^{2} - (i \chi_{i})^{2}} = \frac{i \chi_{i} \chi_{i} - \frac{i \chi_{i} \chi_{i}}{h}}{i \chi_{i}^{2} - \frac{(i \chi_{i})^{2}}{h}} = \frac{i \chi_{i} \chi_{i} - \frac{i \chi_{i} \chi_{i}}{h}}{i (\chi_{i} - \chi_{i})^{2}}$$

$$b_{1} = \frac{IX_{\lambda}^{2}Y_{\lambda} - IX_{\lambda}IX_{\lambda}Y_{\lambda}}{nIX_{\lambda}^{2} - iIX_{\lambda}Y_{\lambda}} = \frac{\overline{Y}nIX_{\lambda}^{2} - n\overline{X} \cdot IX_{\lambda}Y_{\lambda}}{nIX_{\lambda}^{2} - n\overline{X}^{2}} = \frac{IX_{\lambda}^{2}Y_{\lambda} - IX_{\lambda}\overline{X}^{2}\overline{Y}}{I(X_{\lambda} - \overline{X})^{2}} - \overline{X}\frac{IX_{\lambda}Y_{\lambda} - n\overline{X}\overline{Y}}{I(X_{\lambda} - \overline{X})^{2}}$$

$$= \overline{Y} - b_{\lambda}\overline{X}$$

2.
$$Var(b) = \sigma^{2}(x'x)^{-1} = \sigma^{2} \frac{1}{h \Sigma x_{i}^{2} - (\Sigma x_{i})^{2}} \begin{Bmatrix} \Sigma x_{i}^{2} - \Sigma x_{i} \\ -\Sigma x_{i} & h \end{Bmatrix} = \frac{\sigma^{2}}{h \Sigma (x_{i} - \overline{x})^{2}} \begin{Bmatrix} \Sigma x_{i}^{2} - \Sigma x_{i} \\ -\Sigma x_{i} & h \end{Bmatrix}$$

$$= \begin{pmatrix} \sigma' \Sigma x_{i}^{2} & \sigma' 2 - \overline{x} \\ h \Sigma (x_{i} - \overline{x})^{2} & \overline{\Sigma} (x_{i} - \overline{x})^{2} \\ -\overline{\Sigma} (x_{i} - \overline{x})^{2} & \overline{\Sigma} (x_{i} - \overline{x})^{2} \end{pmatrix}$$

therefore:
$$Var(b_1|x) = \sigma_{11}^2 = \frac{\sigma^2 \Sigma x_i^2}{h I (x_i \cdot \overline{x})^2}$$

$$Var(b_2|x) = \sigma_{12}^2 = \frac{\sigma^2}{I (x_i \cdot \overline{x})^2}$$

$$\omega_V(b_1, b_2|x) = \sigma_{12}^2 = \frac{\sigma^2 (-\overline{x})}{I (x_i \cdot \overline{x})^2}$$

5.3

$$A. Li) t_{stat} = \frac{1.4515}{1.2019} = 0.6592$$

(ii)
$$se(b_2) = \frac{2.7648}{5.7103} = 0.4842$$

$$(iv) 55T = 6.39549 (/200-1) = 49041.5418$$

$$R^{2} = 1 - \frac{46221.62}{49041.5418} = 0.0595$$

(V)
$$\hat{G} = \sqrt{\frac{46221.62}{1200-4}} = 6.217$$

b. $b_2 = 2.7648 = |\text{unit increase in total expenditure will lead to 3.7448 unit increase on share of alcohol expenditure when other factors are held.$

of alcohol expenditure when other factors are held.

b = -1.15494 = /unit increase in number of children will lead to 1.4549 unit decrease on share of alcohol expenditure when other factors are held.

of alcohol expenditure when other factors are held.

by=-0.1503 =/unit ircrease in age of household will lead to 0.1503 and decrease on share of alcohol expenditure when other factors are held.

C. $CL = \{-0.1503 - 1.96 \times 0.0235, -0.1503 + 1.96 \cdot 0.0235\} = \{-0.1964, -0.1042\}$ There is a 95% confidence that increase I year of age the share of alwhole expenditure is estimated decrease between -0.1964 with -0.1042.

d. Except for the intercept All welkident are significantly different from 0 at 5% level.

C.
$$H_0: \beta_3 = -2$$

 $H_1: \beta_3 \neq -2$, $t = \frac{-1.4515 + 2}{0.3695} = 1.475 < 1.96$

=7 do not reject Ho. There is no sufficient evidence to say that the decrease is different from 2%

a. B. is negative, because larger quantity will make price lower B, is positive, because larger quality will make price larger By is not sure, because increasing supply leads fall in price and increasing demand lends fall in price

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Residuals:
Min 1Q Median 3Q Max
-43.479 -12.014 -3.743 13.969 43.753
        Residual standard error: 20.06 on 52 degrees of freedom
Multiple R-squared: 0.5097, Adjusted R-squared: 0.4814
F-statistic: 18.02 on 3 and 52 DF, p-value: 3.806e-08
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quantity increase I unit & price decrease 0.05977 with other factor is held. quality increase | unit => price increase 0.1162 | with other factor is held. time increase I unit & price decrease 2.35458 with other factor is held. =) All the signs ove same as expected.

PRICE = 90.14669 - 0.05997QUANT+0.1/621QUAL-2.3498TREND+e

C. R= 05097

d. Ho: B2≥0

 $H_0: \beta_1 = 0$ $= \frac{5}{100} t = -5.892 < -1.613b$

=> reject Ho, there is sufficient evidence to indicate that quality gets larger and the price of be lower > qt(0.05, df = 54)

e. $H_0: \beta_3 = 0$ $H_1: \beta_3 > 0 , t = 0.572 < 1.673b$

[1] -1.673565 > qt(0.95, df = 54)[1] 1.673565

=) do not reject Ho, there is no sufficient evidence to show that quality gots larger the price will be larger

f. average = -2.3548

it may because the development of technology.