$$\gamma_{1} = b_{1} + b_{2} \alpha_{1} + e_{1} \qquad \gamma = \begin{bmatrix} \gamma_{1} \\ \gamma_{2} \\ \vdots \\ \gamma_{n} \end{bmatrix} \qquad \beta = \begin{bmatrix} b_{1} \\ b_{2} \end{bmatrix} \qquad \chi = \begin{bmatrix} \alpha_{1} \\ \alpha_{2} \\ \vdots \\ \alpha_{N} \end{bmatrix}$$
matrix form: $\gamma = x \beta + e$

Qı

$$\beta = (x'x)^{-1} (x'Y)$$

$$(x'X) = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ 1 & x_1 & \cdots & x_n \end{bmatrix} \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \end{bmatrix} = \begin{bmatrix} n & \frac{2}{1}x_1 \\ 1 & x_2 \end{bmatrix} \begin{bmatrix} 1 & \cdots & 1 \\ 1 & x_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \frac{n}{2}x_1 \\ \frac{n}{2}x_1 \end{bmatrix}$$

$$(X_{1} X) = \begin{bmatrix} \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{1}} \\ \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{1}} \\ \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \end{bmatrix} = \begin{bmatrix} \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{1}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2}} \\ \frac{1}{\lambda_{2}} & \frac{1}{\lambda_{2$$

$$= \frac{1}{n \frac{c}{c} \chi_{i}^{2} - \frac{c}{c} \chi_{i}^{2}} \begin{bmatrix} \frac{1}{c} \chi_{i}^{2} \frac{c}{c} \chi_{i} + \frac{c}{c} \chi_{i} \frac{c}{c} \chi_{i} \\ \frac{1}{c} \chi_{i}^{2} \frac{c}{c} \chi_{i}^{2} + \frac{c}{c} \chi_{i} \frac{c}{c} \chi_{i}^{2} \end{bmatrix}$$

$$p' = \frac{\sum_{i=1}^{L} x_i}{\sum_{i=1}^{L} x_i} x_i - (u\underline{x})$$

$$= \frac{\sum_{i=1}^{L} x_i}{\sum_{i=1}^{L} x_i} x_i - u\underline{x} + \sum_{i=1}^{L} x_i x_i$$

$$= \frac{\sum_{i=1}^{L} x_i}{\sum_{i=1}^{L} x_i} x_i - u\underline{x}$$

$$b_{\lambda} = \frac{-n^{2} \overline{\chi} \overline{\gamma} + n \sum_{i=1}^{n} \gamma_{i} \gamma_{i}}{n \sum_{i=1}^{n} \chi_{i}^{2} - (n \overline{\chi})^{2}} = \frac{\sum_{i=1}^{n} \chi_{i} \gamma_{i} - n \overline{\chi} \overline{\gamma}}{\sum_{i=1}^{n} \chi_{i}^{2} - n \overline{\chi}^{2}}$$

The Ordinary Least Squares (OLS) Estimators

$$b_2 = \frac{\sum (x_i - \overline{x})(y_i - \overline{y})}{\sum (x_i - \overline{x})^2}$$
 (2.7)

$$b_1 = \overline{y} - b_2 \overline{x} \tag{2.8}$$

where $\bar{y} = \sum y_i/N$ and $\bar{x} = \sum x_i/N$ are the sample means of the observations on y and x.

$$(2.7)$$
式展開, $b_3 = \frac{2(x_i y_i - x_i \overline{y} - \overline{x} y_i + \overline{x} \overline{y})}{\overline{z} x_i^2 - 2\overline{z} x_i \overline{x} + \overline{z} \overline{x}^2} = \frac{\overline{z} x_i y_i - n \overline{x} \overline{y}}{\overline{z} x_i^2 - n \overline{x}^2} = \frac{\overline{z} x_i y_i - n \overline{x} \overline{y}}{\overline{z} x_i^2 - n \overline{x}^2}$

$$(2.8) \mathbb{R} | \mathbf{B} |_{b_1 = \overline{\gamma} - \left(\frac{\Sigma \chi_i \gamma_i - n \overline{\gamma} \overline{\gamma}}{\overline{z} \chi_i^2 - n \overline{\chi}^2}\right) \cdot \overline{\chi}} = \frac{\overline{\gamma} \overline{z} \chi_i^2 - \overline{\gamma} \overline{z} \chi_i^2 - \overline{\chi} \overline{z} \chi_i \gamma_i + n \overline{\chi}^2 \overline{\gamma}}{\overline{z} \chi_i^2 - n \overline{\chi}^2}$$

$$= \frac{\overline{\gamma} \overline{z} \chi_i^2 - \overline{\chi} \overline{z} \chi_i \gamma_i}{\overline{z} \chi_i^2 - n \overline{\chi}^2} \stackrel{\text{def}}{=} \mathbf{D} \mathbf{H} \mathbf{\tilde{z}}$$

$$V_{\alpha Y}(b) = 6^{2} \left(\frac{1}{X} \right)^{-1} = \frac{6^{2}}{n \frac{2}{i \cdot 1} x_{i}^{2} - \frac{2}{i \cdot 1} x_{i}^{2}} \begin{bmatrix} \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2} \\ -\frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2} \end{bmatrix} = \begin{bmatrix} \frac{6^{2} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2} - \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2}} & \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}^{2}} \\ \frac{-6^{3} \frac{n}{n} x_{i}^{2}}{n \frac{n}{n} x_{i}$$

$$V_{\alpha Y}(b_1|X) = \frac{6^2 \sum_{i=1}^{n} \chi_i^2}{n \sum_{i=1}^{n} \chi_i^2 - \sum_{i=1}^{n} \chi_i^2} = 6^2 \frac{\sum_{i=1}^{n} \chi_i^2}{n \sum_{i=1}^{n} \chi_i^2 - (n \bar{\chi})^2} \times$$

$$V_{OV}(b_{2}|\chi) = \frac{6^{2} h}{h_{\frac{1}{12}} \chi_{1}^{2} - \frac{1}{2} \chi_{1}^{2}} = \frac{6^{2} h}{h_{\frac{1}{12}} \chi_{1}^{2} - (m\bar{\chi})^{2}} = \frac{6^{2}}{\frac{5}{12} \chi_{1}^{2} - h\bar{\chi}^{2}}$$

$$Cov(b_{1},b_{2}|x) = \frac{-6^{2}\frac{h}{2}x_{1}}{n\frac{h}{2}x_{1}^{2} - \frac{h}{2}x_{1}^{2}} = \frac{-6^{2}\cdot hx}{n\frac{h}{2}x_{1}^{2} - n^{2}x_{2}^{2}} = \frac{-6^{2}x}{\frac{h}{2}x_{1}^{2} - n^{2}x_{2}^{2}} = \frac{-6^{2}x}{\frac{h}{2}x_{1}^{2} - n^{2}x_{2}^{2}}$$

$$\operatorname{var}(b_1|\mathbf{x}) = \sigma^2 \left[\frac{\sum x_i^2}{N\sum (x_i - \overline{x})^2} \right]$$
 (2.14)

$$\operatorname{var}(b_2|\mathbf{x}) = \frac{\sigma^2}{\sum (x_i - \overline{x})^2}$$
 (2.15)

$$\operatorname{var}(b_{2}|\mathbf{x}) = \frac{\sigma^{2}}{\sum (x_{i} - \bar{x})^{2}}$$

$$\operatorname{cov}(b_{1}, b_{2}|\mathbf{x}) = \sigma^{2} \left[\frac{-\bar{x}}{\sum (x_{i} - \bar{x})^{2}} \right]$$
(2.15)

$$\mathbb{Z}\left(\gamma_{1}^{2}-\overline{\chi}\right)^{2}=\mathbb{Z}\chi_{1}^{2}-2\mathbb{Z}\chi_{1}\overline{\chi}+\mathbb{Z}\overline{\chi}^{2}=\mathbb{Z}\chi_{1}^{2}-n\overline{\chi}^{2}$$

5.3 Consider the following model that relates the percentage of a household's budget spent on alcohol *WALC* to total expenditure *TOTEXP*, age of the household head *AGE*, and the number of children in the household *NK*.

$$WALC = \beta_1 + \beta_2 \ln(TOTEXP) + \beta_3 NK + \beta_4 AGE + e$$

This model was estimated using 1200 observations from London. An incomplete version of this output is provided in Table 5.6.

TABLE 5.6 Output for Exercise 5.3

Dependent Variable: WALC Included observations: 1200				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	1.4515	2.2019	0.6592	0.5099
ln(TOTEXP)	2.7648	0.4842	5.7103	0.0000
NK	-1,4549	0.3695	-3.9376	0.0001
AGE	-0.1503	0.0235	-6.4019	0.0000
R-squared	0.0575	Mean dependent var		6.19434
S.E. of regression	6.2161	S.D. dependent var		6.39547
Sum squared resid	46221.62			

- a. Fill in the following blank spaces that appear in this table.
 - i. The *t*-statistic for b_1 .
 - ii. The standard error for b_2 .
 - iii. The estimate b_3 .
 - iv. R^2 .
 - v. σ̂.
- **b.** Interpret each of the estimates b_2 , b_3 , and b_4 .
- c. Compute a 95% interval estimate for β_4 . What does this interval tell you?
- **d.** Are each of the coefficient estimates significant at a 5% level? Why?
- e. Test the hypothesis that the addition of an extra child decreases the mean budget share of alcohol by 2 percentage points against the alternative that the decrease is not equal to 2 percentage points. Use a 5% significance level.

(A)
$$\frac{\text{Coeff.(b)}}{5E} = \frac{1.4515}{2.2019} = 0.6529$$

(iv)
$$R^2 = 1 - \frac{45R}{55T} = 1 - \frac{46221.62}{\frac{55}{5}} = 1 - \frac{46221.62}{\frac{55}{5}} = 0.0575$$

(V)
$$\hat{G}^{2} = \frac{138.64}{10-16} = \frac{1300-4}{1200-4} = 38.64$$

(TOTEXP有力olog)

其他條件不變之下, TOTEXP 每增加 1%, WALC平均上升2.7648%

b3 = -1.4 549

其他條件不變之下,家中每多一個孩子(NK),WALC平均下降1.4549%

b4 = -0.1×03

其他條件不變之下, AGE每增一單位, WALC平均下降 0.1503%

(C) CI for by = $-0.1503 \pm 1.9619 \times 0.0235$ $t_{1196,0.025} = 1.9619$ => -0.1964 \leq by \leq -0.1042

在9×%的信心水準下,真實的好會被涵蓋在 interval中

- (d) 除了Intercept 外, 其他 beta的 p-value 都小於如了, 他們都顯著
- (e) $H_0 = b_3 = -2$, $H_1 = b_3 + -2$ $\frac{-1.4549 - (-2)}{0.3695} = 1.4752 < t_{1196,0.025} = 1.9619$

not veject H。,表示統計上沒有顕著證明支持額外NK 會對 WALC造成的 effect

不是一二%

(a)

b2 預期符號:負

販賣數量增加時,通常會有數量折扣,降低價格

b3 預期符號:正

純度較高的可卡因應該能賣得更高的價格

b4 預期符號:可能為負

市場供應增加或競爭變激烈導致價格下降

(b)

```
Call:
```

lm(formula = price ~ quant + qual + trend, data = cocaine)

Residuals:

Min 1Q Median 3Q Max -43.479 -12.014 -3.743 13.969 43.753

Coefficients:

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

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

- b2 數量每增加 1 克,價格會降低 0.05997 美元,符合數量折扣假設
- b3 純度每提高 1%,價格上升 0.11621 美元,符合預期
- b4 每年價格下降 2.35458 美元,可能反映市場供應增加

大致符合預期

(c)

R² = 0.5097, 代表該三個變數可以解釋 50.97%的價格的變異

(d)

H0:b2=0

H1:b2 < 0

 $t_{0.05, 52}$ = -1.6745

critical value = -0.05997/0.01018 = 5.89 < -1.6745。拒絕 H0,接受 H1

(e)

H0:b3=0

H1:b3>0

 $t_{0.05, 52}$ = 1.6745

critical value = 0.11621/0.20326 = 0.57< 1.6745。不拒絕 H0,沒有充足證據顯示 qual 對 price 有正面影響力

(f)

價格每年平均下降 2.35458, 在其他條件不變下。可能代表市場逐年的供給增加,讓價格下降