

Q1) a) $\sum y_i = \sum \hat{y}_i$

Proof:

consider, $\sum y_i - \sum \hat{y}_i$

$$= \sum (y_i - (b_0 + b_1 x_i))$$

$$= \sum y_i - n b_0 + b_1 \sum x_i$$

$$= 0 \quad (\because \text{Normal equation (1)}).$$

$$\therefore \sum y_i = \sum \hat{y}_i.$$

b) $\sum x_i e_i = 0$

Proof:

$$\sum x_i e_i = \sum x_i (y_i - b_0 - b_1 x_i)$$

$$= \sum x_i y_i - b_0 \sum x_i - b_1 \sum x_i^2$$

$$= 0 \quad (\because \text{Normal Equation (2)}).$$

c) $\sum \hat{y}_i e_i = 0$

Proof

$$\sum \hat{y}_i e_i = \sum (b_0 + b_1 x_i) e_i$$

$$= b_0 \underbrace{\sum e_i}_{=0} + b_1 \underbrace{\sum x_i e_i}_{=0}$$

$$= 0 \quad (\because \sum e_i = 0 \text{ and } \sum x_i e_i = 0).$$

Q2)

a)

X_i	Y_i	$X_i - \bar{X}$	$Y_i - \bar{Y}$	$(X_i - \bar{X})(Y_i - \bar{Y})$	$(X_i - \bar{X})^2$
				0	0
1	16	0	1.8	5.2	1
0	9	-1	-5.2	2.8	1
2	17	1	2.8	2.2	1
0	12	-1	-2.2	15.6	4
3	22	2	7.8	0	0
1	13	0	-1.2	6.2	1
0	8	-1	-6.2	0	0
1	15	0	0.8	4.8	1
2	19	1	4.8	3.2	1
0	11	-1	-3.2		

$$\sum X_i = 10 \quad \sum Y_i = 142$$

$$\Rightarrow \bar{X} = 1 \quad \rightarrow \bar{Y} = 14.2$$

$$\sum (X_i - \bar{X})(Y_i - \bar{Y}) = 40$$

$$\sum (X_i - \bar{X})^2 = 10$$

$$b_1 = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{\sum (X_i - \bar{X})^2} = \frac{40}{10} = 4$$

$$b_0 = \bar{Y} - b_1 \bar{X} = 14.2 - 4 \times 1 = 10.2$$

\therefore Estimated Regression Function is $\hat{Y} = 10.2 + 4X$.

$$b) E(Y) = 10.2 + 4(1) = 14.2$$

$$c) E(Y)_{X=2} - E(Y)_{X=1} = b_1 = 4$$

$$d) Y = 10.2 + 4(\bar{X}) = 10.2 + 4(1) = 14.2 = \bar{Y}$$

\therefore Regression line $Y = 10.2 + 4X$ goes through $(\bar{X}, \bar{Y}) = (1, 14.2)$.

Q3

(3)

X_i	Y_i	$X_i - \bar{X}$	$Y_i - \bar{Y}$	$(X_i - \bar{X})(Y_i - \bar{Y})$	$(X_i - \bar{X})^2$
16	199	-12	-26.5625	318.75	144
16	205	-12	-20.5625	246.75	144
16	196	-12	-29.5625	354.75	144
16	200	-12	-25.5625	306.75	144
24	218	-4	-7.5625	30.25	16
24	220	-4	-5.5625	22.25	16
24	215	-4	-10.5625	42.25	16
24	223	-4	-2.5625	10.25	16
32	237	4	11.4375	45.25	16
32	234	4	8.4375	33.75	16
32	235	4	9.4375	37.75	16
32	230	4	4.4375	17.75	16
40	250	12	24.4375	293.25	144
40	248	12	22.4375	269.25	144
40	253	12	27.4375	329.25	144
40	246	12	20.4375	245.25	144
				$\Sigma(X_i - \bar{X})(Y_i - \bar{Y})$	$\Sigma(X_i - \bar{X})^2$
				= 2604	= 1280

$$\Sigma X_i = 448 \quad \Sigma Y_i = 3609$$

$$\Rightarrow \bar{X} = 28 \quad \Rightarrow \bar{Y} = 225.5625$$

$$a) \quad b_1 = \frac{2604}{1280} = 2.034, \quad b_0 = \bar{Y} - b_1 \bar{X} = 225.5625 - 2.034 \times 28 = 168.61$$

\therefore Estimated Regression line: $\hat{Y} = 168.61 + 2.034X$,

b) When $X = 40$,

$$E(Y) = 168.61 + 2.034(40) = 249.97.$$

c) change in mean hardness when X is increased by 1 hour

$$\text{is } \hat{\beta}_1 = b_1 = 2.034_{hr}$$

Q4

(4)

y_i	199	205	196	200	218	220	215	223	237	234
\hat{y}_i	201.15	204.15	201.15	201.15	217.43	217.43	217.43	217.43	233.70	233.70
e_i	-2.15	3.85	-5.15	-1.15	0.57	2.57	-2.43	5.57	3.30	0.30

y_i	235	230	250	248	253	246
\hat{y}_i	233.70	233.70	249.98	249.98	249.98	249.98
e_i	1.30	-3.70	0.02	-1.98	3.02	-3.98

$$\sum e_i = -2.15 + 3.85 - 5.15 - 1.15 + 0.57 + 2.57 - 2.43 + 5.57 + 3.30 + 0.30 \\ + 1.30 - 3.70 + 0.02 - 1.98 + 3.02 - 3.98 = 0.00$$

Q5] If $\beta_1 = 0$, then there is no linear relationship between the response variable and the predictor variable.
Regression function with $\beta_1 = 0$ is a horizontal line.

$$Q6] \sum y_i = nb_0 + b_1 \sum x_i \longrightarrow (1)$$

$$\sum x_i y_i = b_0 \sum x_i + b_1 \sum x_i^2 \longrightarrow (2)$$

$$(2) \times n - (1) \times \sum x_i :$$

$$n \sum x_i y_i - \sum x_i \sum y_i = nb_1 \sum x_i^2 - b_1 (\sum x_i)^2$$

$$\Rightarrow b_1 = \frac{n \sum x_i y_i - \sum x_i \sum y_i}{n(\sum x_i)^2 - (\sum x_i)^2}$$

$$= \frac{\sum x_i y_i - \frac{1}{n} \sum x_i \sum y_i}{\sum x_i^2 - \frac{1}{n} (\sum x_i)^2}$$

$$\begin{aligned}
 b_1 &= \frac{\sum x_i y_i - \frac{1}{n} \sum x_i \sum y_i - \frac{1}{n} \sum x_i \sum y_i + \frac{1}{n} \sum x_i y_i}{\sum x_i^2 - \frac{1}{n} (\sum x_i)^2 - \frac{1}{n} (\sum x_i)^2 + \frac{1}{n} (\sum x_i)^2} \\
 &= \frac{\sum x_i y_i - \bar{x} \sum y_i - \bar{y} \sum x_i + n \bar{x} \bar{y}}{\sum x_i^2 - 2\bar{x} \sum x_i + n \bar{x}^2} \\
 &= \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2}
 \end{aligned}$$

Q7) $y_i = \beta_0 + \epsilon_i$

consider,

$$Q = \sum_{i=1}^n (y_i - \beta_0)^2$$

d.w.r.t. β_0 ,

$$\frac{dQ}{d\beta_0} = 2 \sum (y_i - \beta_0) (-1) \stackrel{\text{Set}}{=} 0$$

$$\Rightarrow \sum y_i - n\beta_0 = 0$$

$$\Rightarrow \hat{\beta}_0 = b_0 = \frac{\sum y_i}{n} = \bar{y}$$

$$\frac{d^2Q}{d\beta_0^2} = -2n < 0$$

i. Least square estimator of $\beta_0 = \hat{\beta}_0 = b_0 = \bar{y}$.

Q2

$$E(b_0) = E(\bar{y}) = E\left(\frac{\sum y_i}{n}\right)$$

$$= \frac{1}{n} \sum E(y_i) = \frac{1}{n} \sum E(\beta_0 + \xi_i) (\because y = \beta_0 + \xi_i)$$

$$= \frac{1}{n} \sum [\beta_0 + E(\xi_i)]$$

$$= \frac{1}{n} [n\beta_0 + 0] = \beta_0$$

$$\hat{Q} \quad E(b_0) = \beta_0$$

$\therefore b_0$ is an unbiased estimator.