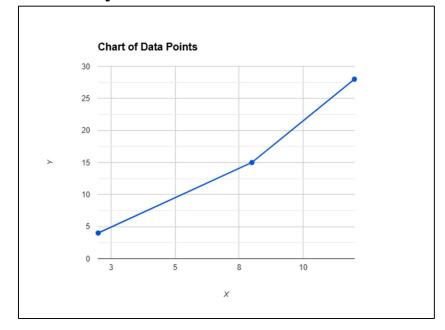


Linear Regression case study:

Input data:

| input datai | | |
|-------------|----|------------|
| i | Xi | y i |
| 0 | 2 | 4 |
| 1 | 8 | 15 |
| 2 | 12 | 28 |



Sum of squared Residuals (SSR):

$$SSR = \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$
 (1)

Mean squared error (MSE):

$$MSE = \frac{1}{n}SSR = \frac{1}{n}\sum_{i=1}^{n}(y_i - \hat{y}_i)^2$$
 (2)

Linear equation:

$$\hat{y}_i = k. \, x_i + q \tag{3}$$

Combine equations = cost function:

$$J(k,q) = \frac{1}{n} \sum_{i=1}^{n} [y_i - (k.x_i + q)]^2$$
 (4)

Find global minimum of the function:

$$\frac{\partial J}{\partial q} = 0$$

$$\frac{\partial J}{\partial k} = 0$$
(5)

- derivatives:



$$\frac{\partial J}{\partial q} = \frac{2}{n} \sum_{i=1}^{n} (y_i - k. x_i - q). (-x_i)$$

$$\frac{\partial J}{\partial k} = \frac{2}{n} \sum_{i=1}^{n} (y_i - k. x_i - q). (-1)$$

- after plugging the numbers into the equations, we get:

$$k \approx 2.355$$

$$q \approx -1.605$$
(7)

(6)

The best fitting line has equation:

$$\hat{y}_i = k. x_i + q = 2.355 x_i - 1.605$$
(8)

