

Practical Class 03

Inner Product and Linear Regression

1. Suppose that \mathbf{x} , \mathbf{y} and \mathbf{z} are three vectors in \mathbb{R}^3 such that

$$2\mathbf{x} + 2\mathbf{y} + \mathbf{z} = \mathbf{0}$$

and $|\mathbf{x}| = 3$, $|\mathbf{y}| = 4$, $|\mathbf{z}| = 7$. Use inner product (dot product) to find the angle between \mathbf{x} and \mathbf{y} .

Solution: Let

$$\mathbf{x} = (x_1, x_2, x_3), \mathbf{y} = (y_1, y_2, y_3) \text{ and } \mathbf{z} = (z_1, z_2, z_3).$$

Then the condition $2\mathbf{x} + 2\mathbf{y} + \mathbf{z} = \mathbf{0}$ can be written as

$$2x_1 + 2y_1 + z_1 = 0, \quad 2x_2 + 2y_2 + z_2 = 0, \quad 2x_3 + 2y_3 + z_3 = 0,$$

or equivalently,

$$2x_1 + 2y_1 = -z_1, \quad 2x_2 + 2y_2 = -z_2, \quad 2x_3 + 2y_3 = -z_3.$$

Recall that the inner(dot) product of \mathbf{x} and \mathbf{y} is defined to be $x_1y_1 + x_2y_2 + x_3y_3$, take squares of each of the above equation yields

$$4x_1^2 + 4y_1^2 + 8x_1y_1 = z_1^2, \quad 4x_2^2 + 4y_2^2 + 8x_2y_2 = z_2^2, \quad 4x_3^2 + 4y_3^2 + 8x_3y_3 = z_3^2.$$

Sum up these three equations and we get

$$4(x_1^2 + x_2^2 + x_3^2) + 4(y_1^2 + y_2^2 + y_3^2) + 8(x_1y_1 + x_2y_2 + x_3y_3) = z_1^2 + z_2^2 + z_3^2,$$

that is

$$4|\mathbf{x}|^2 + 4|\mathbf{y}|^2 + 8\langle \mathbf{x}, \mathbf{y} \rangle = |\mathbf{z}|^2.$$

Substitute the value in, we get

$$\langle \mathbf{x}, \mathbf{y} \rangle = -6.375 = |\mathbf{x}||\mathbf{y}|\cos(\theta),$$

where θ is angle between vector \mathbf{x} and \mathbf{y} . Therefore,

$$\cos(\theta) = \frac{\langle \mathbf{x}, \mathbf{y} \rangle}{|\mathbf{x}||\mathbf{y}|} = \frac{-6.375}{3 \times 4} = -0.53125,$$

and

$$\theta = 2.13.$$

2. As seen in the lecture slides, the parameter vector for a general linear regression with a bias term

$$h_{\theta}(\mathbf{x}) = \theta[0] + \theta[1]x[1] + \theta[2]x[2] + \cdots + \theta[n]x[n]$$

can be found in the matrix form by

$$\theta = (XX^T)^{-1}X\mathbf{y},$$

where

$$X = \begin{bmatrix} x_1[1] & x_2[1] & \cdots & x_m[1] \\ x_1[2] & x_2[2] & \cdots & x_m[2] \\ \vdots & \vdots & \ddots & \vdots \\ x_1[n] & x_2[n] & \cdots & x_m[n] \end{bmatrix}, \quad \mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}.$$

Now, consider a simple model with $n = 1$, that is,

$$y = \theta[0] + \theta[1]x.$$

Try to deduce the general formula for parameter $\theta[0]$ (bias) and $\theta[1]$ using matrix multiplication for a data set with m training samples.

Solution: Let's denote the training data as $\{(x_i, y_i)\}_{i=1}^m$. The augmented data matrix in this case is

$$X = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_1 & x_2 & \cdots & x_m \end{bmatrix},$$

and

$$XX^T = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_1 & x_2 & \cdots & x_m \end{bmatrix} \times \begin{bmatrix} 1 & x_1 \\ 1 & x_2 \\ \vdots & \vdots \\ 1 & x_m \end{bmatrix} = \begin{bmatrix} m & m\bar{x} \\ m\bar{x} & \sum_{i=1}^m x_i^2 \end{bmatrix},$$

noticing that $\sum_{i=1}^m x_i = m\bar{x}$. Recall that the inverse for 2×2 matrix is

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}.$$

Thus, we have

$$\begin{aligned} \theta &= \begin{bmatrix} \theta[0] \\ \theta[1] \end{bmatrix} = \frac{1}{m \sum_{i=1}^m x_i^2 - (m\bar{x})^2} \begin{bmatrix} \sum_{i=1}^m x_i^2 & -m\bar{x} \\ -m\bar{x} & m \end{bmatrix} \begin{bmatrix} 1 & 1 & \cdots & 1 \\ x_1 & x_2 & \cdots & x_m \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix} \\ &= \frac{1}{m \sum_{i=1}^m x_i^2 - (m\bar{x})^2} \begin{bmatrix} \sum_{i=1}^m x_i^2 - m\bar{x}x_1 & \cdots & \sum_{i=1}^m x_i^2 - m\bar{x}x_m \\ -m\bar{x} + mx_1 & \cdots & -m\bar{x} + mx_m \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix} \\ &= \frac{1}{m \sum_{i=1}^m x_i^2 - (m\bar{x})^2} \begin{bmatrix} m\bar{y} \left(\sum_{i=1}^m x_i^2 \right) - m\bar{x} \left(\sum_{i=1}^m x_i y_i \right) \\ m \sum_{i=1}^m x_i y_i - m^2 \bar{x} \bar{y} \end{bmatrix}, \end{aligned}$$

that is

$$\theta[0] = \frac{\bar{y} \left(\sum_{i=1}^m x_i^2 \right) - \bar{x} \left(\sum_{i=1}^m x_i y_i \right)}{\sum_{i=1}^m x_i^2 - m\bar{x}^2} \quad \text{and} \quad \theta[1] = \frac{\sum_{i=1}^m x_i y_i - m\bar{x}\bar{y}}{\sum_{i=1}^m x_i^2 - m\bar{x}^2}$$

3. A professor in the faculty of Engineering polled a dozen colleagues (m) about the number of professional meetings they attended in the past five years (x) and the number of papers they submitted to refereed journals (y) during the same period. The summary data are given as follows:

$$m = 12, \bar{x} = 4, \bar{y} = 12,$$

$$\sum_{i=1}^{12} x_i^2 = 232, \sum_{i=1}^{12} x_i y_i = 318.$$

Fit a simple linear regression model between x and y by finding the estimate of the intercept and slope using your results from the previous question. Comment on whether attending more professional meetings would result in publishing more papers.

Solution: From the previous question

$$\theta[0] = \frac{\bar{y} \left(\sum_{i=1}^m x_i^2 \right) - \bar{x} \left(\sum_{i=1}^m x_i y_i \right)}{\sum_{i=1}^m x_i^2 - m \bar{x}^2} = \frac{(12)(232) - (4)(318)}{232 - (12)(4^2)} = \frac{189}{5} = 37.8,$$

and

$$\theta[1] = \frac{\sum_{i=1}^m x_i y_i - m \bar{x} \bar{y}}{\sum_{i=1}^m x_i^2 - m \bar{x}^2} = \frac{318 - (12)(4)(12)}{232 - (12)(4^2)} = -\frac{129}{20} = -6.45.$$

So the simple linear regression line is

$$y = 37.8 - 6.45x.$$

It appears that attending professional meetings would not result in publishing more papers (due to the negative slope).

4. A study was made on the amount of converted sugar in a certain process at various temperatures. The data was recorded as follows:

Temperature (x)	1.0	1.1	1.2	1.3	1.4	1.5	1.6	1.7	1.8	1.9	2.0
Converted sugar (y)	8.1	7.8	8.5	9.8	9.5	8.9	8.6	10.2	9.3	9.2	10.5

- (a) Fit a linear regression line to the data using your result to Q2 and plot the regression line on a graph with the data.
 (b) Estimate the mean amount of converted sugar produced when the temperature is 1.75.
 (c) Plot the residuals versus temperature. Comment.

Solution:

- (a) With the data set, we have

$$n = 11, \sum_{i=1}^n x_i = 16.5, \sum_{i=1}^n y_i = 100.4, \sum_{i=1}^n x_i^2 = 25.85, \sum_{i=1}^n x_i y_i = 152.59.$$

Using previous results,

$$\theta[0] = \frac{(100.4/11)(25.85) - (16.5/11)(152.59)}{25.85 - 16.5^2/11} = 6.4132,$$

$$\theta[1] = \frac{152.59 - (16.5)(100.4)/11}{25.85 - 16.5^2/11} = 1.8091.$$

Hence

$$\hat{y} = 6.4136 + 1.8091x.$$

- (b) For $x = 1.75$, $\hat{y} = 6.4136 + 1.8091(1.75) = 9.580$.
 (c) See Figure below. Residuals appear to be random as desired.