

1. Let  $\vec{y} = A\vec{x} + \vec{b}$  be random vector.

a) Let's compute the expected value  $E(\vec{y})$  over the arbitrary interval  $\Omega$ :

$$E(\vec{y}) = \int_{\Omega} (A\vec{x} + \vec{b}) P(\vec{x}) d\vec{x}$$

where  $P(x)$  is the probability density function.

$$= A \underbrace{\int_{\Omega} \vec{x} P(\vec{x}) d\vec{x}}_{E(\vec{x})} + \underbrace{\vec{b} \int_{\Omega} P(\vec{x}) d\vec{x}}_{\equiv 1}$$

$$= A E(\vec{x}) + \vec{b} \quad \square$$

$$b) \text{ WTS } \text{cov}[\vec{y}] = A \underbrace{\text{cov}[\vec{x}]}_{\equiv \Sigma} A^T$$

Note  $\text{cov}[\vec{x}]$  is a square matrix which quantifies covariance b/w pairs of elements in  $\vec{x}$ .

$$\therefore \text{cov}[\vec{x}] \equiv \Sigma = E[(\vec{x} - E(\vec{x}))(\vec{x} - E(\vec{x}))^T].$$

$\therefore$  Substituting in,

$$\text{cov}[\vec{y}] = E[(A\vec{x} + \vec{b}) - E(A\vec{x} + \vec{b}) \{ (A\vec{x} + \vec{b}) - E(A\vec{x} + \vec{b}) \}^T]$$

$$= E[\{ A\vec{x} + \vec{b} - (A E(\vec{x}) + \vec{b}) \} \{ A\vec{x} + \vec{b} - (A E(\vec{x}) + \vec{b}) \}^T]$$

$$= E[\{ A(\vec{x} - E(\vec{x})) \} \{ A(\vec{x} - E(\vec{x})) \}^T]$$

$$= A E[(\vec{x} - E(\vec{x}))(\vec{x} - E(\vec{x}))^T] A^T$$

$$= A \text{cov}[\vec{x}] A^T \quad \square$$

by part a  
note  $(AB)^T = B^T A^T$

2. We have the data  $D = \{ (0, 1), (2, 3), (3, 6), (4, 8) \}$ .

a) Find  $\vec{y} = \Theta^T \vec{x}$  using Cramer's rule.

We are using a simple linear regression model:

$$y = \theta_0 + \theta_1 x$$

∴ The optimization problem is:

$$A\vec{\theta} = \vec{b}$$

for  $\vec{\theta} = \begin{bmatrix} \theta_0 \\ \theta_1 \end{bmatrix}$ ,  $A = \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}$ ,  $\vec{b} = \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix}$

the "target" y values

col of  
1s to reflect  
 $\theta_0$  term

col of x  
values

The normal equations are

$$A^T A \vec{\theta} = A^T \vec{b}$$

$A'$   $b'$  for use in Cramer's rule

$$A^T A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix} = \begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}$$

$$4 + 9 + 16 = 29$$

$$A^T \vec{b} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix} = \begin{bmatrix} 18 \\ 56 \end{bmatrix}$$

$$6 + 18 + 32 = 56$$

Cramer's rule says if  $A'$  non singular then solution components are:

$$x_i = \frac{\det(A'_i)}{\det(A')}$$

$A'_i$  denotes replacing  $i^{\text{th}}$  col of  $A'$  w/  $b'$  matrix

$\vec{b}'$

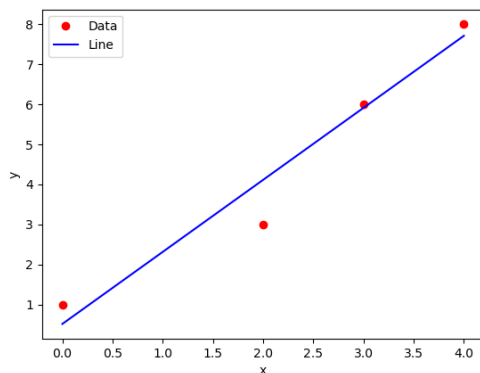
$$\therefore \theta_0 = \frac{\begin{vmatrix} 18 & 9 \\ 56 & 29 \end{vmatrix}}{\begin{vmatrix} 4 & 9 \\ 9 & 29 \end{vmatrix}} = \frac{18 \cdot 29 - 56 \cdot 9}{29 \cdot 4 - 9 \cdot 9} = \frac{18}{35}$$

$$\theta_1 = \frac{\begin{vmatrix} 4 & 18 \\ 9 & 56 \end{vmatrix}}{\begin{vmatrix} 4 & 9 \\ 9 & 29 \end{vmatrix}} = \frac{4 \cdot 56 - 18 \cdot 9}{35} = \frac{62}{35}$$

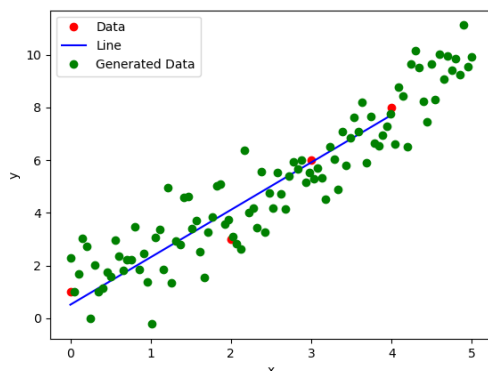
b) Now we use the actual solution to the normal equations, given by

$$\begin{aligned}\vec{\theta} &= (A^T A)^{-1} A^T \vec{b} \\ &= \begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix} \\ &= \frac{1}{4 \cdot 29 - 9 \cdot 9} \begin{bmatrix} 29 & -9 \\ -9 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix} \\ &= \frac{1}{35} \begin{bmatrix} 29 & 11 & 2 & -7 \\ -9 & -1 & 3 & 7 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix} \\ &= \frac{1}{35} \begin{bmatrix} 18 \\ 62 \end{bmatrix} = \begin{bmatrix} \frac{18}{35} \\ \frac{62}{35} \end{bmatrix}, \text{ Same as part a. } \square\end{aligned}$$

c) Plot data w/ fit ✓



d)



Sampled 100 points from  $x = 0 \rightarrow 5$ , computed the exact values using  $A\vec{\theta} + \vec{b}$ , then added random noise term drawn from Gaussian w/  $\mu = 0$ ,  $\sigma = 1$  for each point.