

Feel free to work with other students, but make sure you write up the homework and code on your own (no copying homework *or* code; no pair programming). Feel free to ask students or instructors for help debugging code or whatever else, though. The starter code for problem 2 part c and d can be found under the Resource tab on course website.

Note: You need to create a Github account for submission of the coding part of the homework. Please create a repository on Github to hold all your code and include your Github account username as part of the answer to problem 2.

1 (Linear Transformation) Let $\mathbf{y} = A\mathbf{x} + \mathbf{b}$ be a random vector. show that expectation is linear:

$$\mathbb{E}[\mathbf{y}] = \mathbb{E}[A\mathbf{x} + \mathbf{b}] = A\mathbb{E}[\mathbf{x}] + \mathbf{b}.$$

Also show that

$$\text{cov}[\mathbf{y}] = \text{cov}[A\mathbf{x} + \mathbf{b}] = A\text{cov}[\mathbf{x}]A^\top = A\Sigma A^\top.$$

Given that:

$$\mathbb{E}[\mathbf{x}] = \sum_{i=1}^k x_i p_i$$

$$\mathbb{E}[\mathbf{x}] = \int_{\text{Re}} x f(x) dx$$

$$\int_{\text{Re}} f(x) dx = 1$$

we can show the following:

$$\begin{aligned} \mathbb{E}[\mathbf{y}] &= \int_{\text{Re}} x f(x) dx \\ &= \int_{\text{Re}} (Ax + b) f(x) dx \\ &= A \int_{\text{Re}} x f(x) dx + b \int_{\text{Re}} f(x) dx \\ &= A\mathbb{E}[\mathbf{x}] + b \end{aligned}$$

Given that:

$$\text{cov}[\mathbf{y}] = \mathbb{E}[(\mathbf{y} - \mathbb{E}[\mathbf{y}])(\mathbf{y} - \mathbb{E}[\mathbf{y}])^\top]$$

we can show the following:

$$\begin{aligned} \text{cov}[\mathbf{y}] &= \text{cov}[A\mathbf{x} + \mathbf{b}] \\ &= \mathbb{E}[(A\mathbf{x} + \mathbf{b} - \mathbb{E}[A\mathbf{x} + \mathbf{b}])(A\mathbf{x} + \mathbf{b} - \mathbb{E}[A\mathbf{x} + \mathbf{b}])^\top] \end{aligned}$$

combines with linearity showed above, we see that:

$$\begin{aligned} \text{cov}[\mathbf{y}] &= \mathbb{E}[(A\mathbf{x} + \mathbf{b} - A\mathbb{E}[\mathbf{x}] - \mathbf{b})(A\mathbf{x} + \mathbf{b} - A\mathbb{E}[\mathbf{x}] - \mathbf{b})^\top] \\ &= \mathbb{E}[(A\mathbf{x} - A\mathbb{E}[\mathbf{x}])(A\mathbf{x} - A\mathbb{E}[\mathbf{x}])^\top] \\ &= A^\top A \mathbb{E}[(\mathbf{x} - \mathbb{E}[\mathbf{x}])(\mathbf{x} - \mathbb{E}[\mathbf{x}])^\top] \\ &= A\text{cov}[\mathbf{x}]A^\top \\ &= A\Sigma A^\top \end{aligned}$$

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2 Given the dataset $\mathcal{D} = \{(x, y)\} = \{(0, 1), (2, 3), (3, 6), (4, 8)\}$

- (a) Find the least squares estimate $y = \theta^\top x$ by hand using Cramer's Rule.
- (b) Use the normal equations to find the same solution and verify it is the same as part (a).
- (c) Plot the data and the optimal linear fit you found.
- (d) Find randomly generate 100 points near the line with white Gaussian noise and then compute the least squares estimate (using a computer). Verify that this new line is close to the original and plot the new dataset, the old line, and the new line.

(a)

From the dataset we get:

$x = 0, 2, 3, 4$ and $y = 1, 3, 6, 8$

putting these into the matrix forms we have:

$$x = \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}, y = \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix}$$

and so we can find:

$$x^\top x = \begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}, x^\top y = \begin{bmatrix} 18 \\ 56 \end{bmatrix}$$

Given these results use their relationship to find θ using Cramer's rule:

$$\theta_0 = \frac{\det \begin{bmatrix} 18 & 9 \\ 56 & 29 \end{bmatrix}}{\det \begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}}$$

$$\theta_1 = \frac{\det \begin{bmatrix} 4 & 18 \\ 9 & 56 \end{bmatrix}}{\det \begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}}$$

and so we find the following:

$$\theta_0 = \frac{18}{35}, \theta_1 = \frac{62}{35}$$

$$y = \theta_0 + \theta_1 x = \frac{18}{35} + \frac{62}{35}x$$

(b)

the normal equation is given by:

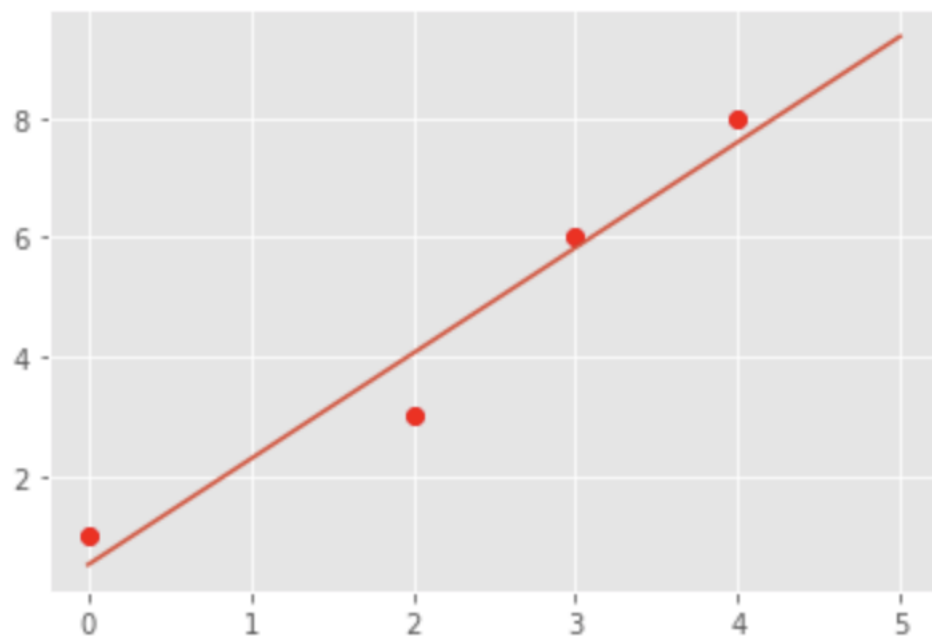
$$\theta = (x^\top x)^{-1} x^\top y$$

plugging in values we find:

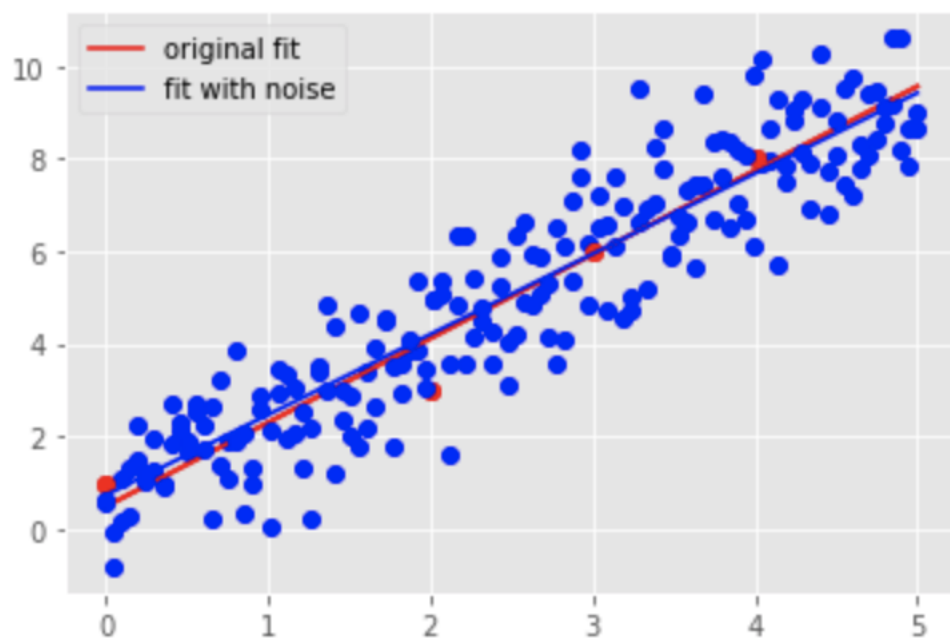
$$\theta = \begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}^{-1} \begin{bmatrix} 18 \\ 56 \end{bmatrix} = \frac{1}{35} \begin{bmatrix} 29 & -9 \\ -9 & 4 \end{bmatrix} \begin{bmatrix} 18 \\ 56 \end{bmatrix} = \begin{bmatrix} \frac{18}{35} \\ \frac{62}{35} \end{bmatrix}$$

there for, same as in part a:

$$\theta_0 = \frac{18}{35}, \theta_1 = \frac{62}{35}$$



(c)



(d)

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