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

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

We have

$$\mathbb{E}[\mathbf{y}] = \mathbb{E}[A\mathbf{x} + \mathbf{b}] = \int_{S} (A\mathbf{x} + \mathbf{b}) p(Y = A\mathbf{x} + \mathbf{b}) dx$$

$$= A \int_{S} \mathbf{x} p(Y = A\mathbf{x} + \mathbf{b}) dx + \mathbf{b} \int_{S} p(Y = A\mathbf{x} + \mathbf{b}) dx$$

$$= A \int_{S} \mathbf{x} p(X = \mathbf{x}) dx + \mathbf{b} \int_{S} p(X = \mathbf{x}) dx$$

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

(I got help for the previous problem from tutoring and solution) Now,

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

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

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

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

$$= A\mathbf{cov}[\mathbf{x}]A^{T}$$

$$= A\mathbf{\Sigma}A^{T}$$

1

- **2** Given the dataset  $\mathcal{D} = \{(x,y)\} = \{(0,1), (2,3), (3,6), (4,8)\}$ 
  - (a) Find the least squares estimate  $y = \theta^{\top} \mathbf{x}$  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 lecture slides)

We know

$$m = \frac{n \sum_{i=1}^{n} x_i y_i - (\sum_{i=1}^{n} x_i)(\sum_{i=1}^{n} y_i)}{n \sum_{i=1}^{n} x_i^2 - (\sum_{i=1}^{n} x_i)^2} \text{ and } b = \frac{n(\sum_{i=1}^{n} x_i)^2(\sum_{i=1}^{n} y_i) - \sum_{i=1}^{n} x_i \sum_{i=1}^{n} x_i y_i}{n \sum_{i=1}^{n} x_i^2 - (\sum_{i=1}^{n} x_i)^2}$$

We have 
$$\sum_{i=1}^{n} x_i = 9$$
,  $\sum_{i=1}^{n} x_i^2 = 29$ ,  $\sum_{i=1}^{n} x_i y_i = 56$ , and  $\sum_{i=1}^{n} y_i = 18$ 

Therefore 
$$m = \frac{4.56 - 9.18}{4.29 - 9.9} = \frac{62}{35}$$
 and  $b = \frac{29.18 - 9.56}{4.29 - 9.9} = \frac{18}{35}$ 

Hence 
$$y = \begin{bmatrix} \frac{18}{35} & \frac{62}{35} \end{bmatrix}^T \begin{bmatrix} 1 & \mathbf{x} \end{bmatrix}$$

(b) The normal equations state 
$$\boldsymbol{\theta} = (X^T X)^{-1} X^T \mathbf{y}$$
, where  $X = \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}$ . We have  $(X^T X) = \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}$ .

$$\begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}$$
, and thus  $(X^T X)^{-1} = \frac{1}{35} \begin{bmatrix} 29 & -9 \\ -9 & 4 \end{bmatrix}$ .

The product 
$$(X^T X)^{-1} X^T = \frac{1}{35} \begin{bmatrix} 29 & 11 & 2 & -7 \\ -9 & -1 & 3 & 7 \end{bmatrix}$$
.

Therefore 
$$\theta = \frac{1}{35}\begin{bmatrix} 29 & 11 & 2 & -7 \\ -9 & -1 & 3 & 7 \end{bmatrix}$$
  $\mathbf{y} = \begin{bmatrix} \frac{18}{35} & \frac{62}{35} \end{bmatrix}$ , and so we obtain the same result.

(c & d) In the python file.

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