

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

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

We know that

$$\mathbb{E}[\mathbf{y}] = \int \mathbf{y}p(x)dx = \int (Ax + b)p(x)dx$$

$$\mathbb{E}[\mathbf{y}] = A \int xp(x)dx + b \int p(x)dx$$

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

Now, let's solve the second part of the problem.

We know that:

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

Using the linearity we proved above:

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

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

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

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

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

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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 $y = \theta^\top \mathbf{x}$ 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) We know that:

$$X^\top X \theta = X^\top y$$

Using the given data, we can construct the following matrices:

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

Thus we have the following equation:

$$\begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix} \theta = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix}$$
$$\begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix} \theta = \begin{bmatrix} 18 \\ 56 \end{bmatrix}$$

Using Cramer's Rule:

$$\theta_0 = \det\left(\begin{bmatrix} 18 & 9 \\ 56 & 29 \end{bmatrix}\right) / \det\left(\begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}\right) = 18/35$$

$$\theta_1 = \det\left(\begin{bmatrix} 4 & 18 \\ 9 & 56 \end{bmatrix}\right) / \det\left(\begin{bmatrix} 4 & 9 \\ 9 & 29 \end{bmatrix}\right) = 62/35$$

Thus, the least square estimate is:

$$y = [18/35 \quad 62/35] \begin{bmatrix} 1 \\ x \end{bmatrix}$$

B)

Using the normal equation:

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

We can also calculate the least square estimate using the same matrices:

$$\boldsymbol{\theta} = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}^{-1} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 0 & 2 & 3 & 4 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix}$$

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

$$\boldsymbol{\theta} = \begin{bmatrix} 29/35 & -9/35 \\ -9/35 & 4/35 \end{bmatrix} \begin{bmatrix} 18 \\ 56 \end{bmatrix}$$

$$\boldsymbol{\theta} = \begin{bmatrix} 18/35 \\ 62/35 \end{bmatrix}$$

Thus, the two solutions are equivalent!

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