Math189R SU20.1 Homework 1 - Sam Mikes (smikes) Thursday, June 11, 2020

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

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

Expectation for a continuous random variable x with normalized probability distribution function  $f_x(x)$  is defined as

$$\mathbb{E}[x] = \int_{-\infty}^{\infty} x f_X(x) dx$$

Consider first a single random variable x, and y = ax + b

$$\mathbb{E}[x] = \mathbb{E}[ax+b] = \int_{-\infty}^{\infty} (ax+b) f_X(x) dx = a \int_{-\infty}^{\infty} x f_X(x) dx + b \int_{-\infty}^{\infty} f_X(x) dx = a \mathbb{E}[x] + b$$

In multiple variables, the  $x_i$  are independent, so it follows that

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

Covariance is defined

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

Again, since the expectation  $\mathbb{E}$  is linear in  $\mathbf{x}$ , the covariance and any linear transformation of  $\mathbf{x}$  commute and can be rearranged. The term in  $\mathbf{b}$  vanishes because the covariance of a constant term is zero.

$$\mathbf{v} = A\mathbf{x} + \mathbf{b}$$

$$\mathbb{C}(\mathbf{v}) = \mathbb{C}(A\mathbf{x} + \mathbf{b}) = A\mathbb{C}(x)A^T$$

 $Works\,used:\,https://www.probabilitycourse.com/chapter6/6\_1\_5\_random\_vectors.php$ 

Note: if

$$\mathbf{x} - > A\mathbf{x}$$

transforms a vector then

$$B->A^TBA$$

transforms a matrix

- 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.
- 1. By Cramer's rule, we have the closed forms

$$m = \frac{n\sum_{i=1}^{n} x_{i}y_{i} - (\sum_{i=1}^{n} x_{i})(\sum_{i=1}^{n} y_{i})}{n\sum_{i=1}^{n} x_{i}^{2} - (\sum_{i=1}^{n} x_{i})^{2}}$$

$$b = \frac{(\sum_{i=1}^{n} x_{i}^{2})(\sum_{i=1}^{n} y_{i}) - (\sum_{i=1}^{n} x_{i})(\sum_{i=1}^{n} x_{i}y_{i})}{n\sum_{i=1}^{n} x_{i}^{2} - (\sum_{i=1}^{n} x_{i})^{2}}$$

**Evaluation** gives

$$\sum(x) = 9, \sum(y) = 18, \sum x^2 = 29, \sum(xy) = 56, \sum(x^2y) = 194$$

error in calculation

 $\sum (x^2)(y) = 29 * 18 = 522$ 

Then

 $m = \frac{(4*56 - 9*18)}{4*29 - 81}$  $= \frac{62}{35} = 1.77$ 

and

$$b = \frac{29 * 18 - 9 * 56}{35}$$
$$= \frac{18}{35}$$
$$= 0.514$$

Numbers, on the other hand, finds m = b = 1.2 as the optimal fit (!)

2. Minimize the error *e*, where

$$e = \mathbf{e} \cdot \mathbf{e}$$

$$\mathbf{e} = \mathbf{y} - (m\mathbf{x} + b)$$

$$\mathbf{e} = \begin{bmatrix} 1 \\ 3 \\ 6 \\ 8 \end{bmatrix} - m * \begin{bmatrix} 0 \\ 2 \\ 3 \\ 4 \end{bmatrix} - b$$

$$\mathbf{e} = \begin{bmatrix} 1 - b \\ 3 - 2m - b \\ 6 - 3m - b \\ 8 - 4m - b \end{bmatrix}$$

$$e = (1-b)^2 + (3-2m-b)^2 + (6-3m-b)^2 + (8-4m-b)^2$$

Then find

$$\frac{\partial e}{\partial m} = -2m[(3 - 2m - b) + (6 - 3m - b) + (8 - 4m - b)]$$
$$= -2(17 - 9m - 3b)$$

$$\frac{\partial e}{\partial b} = -2b[(1-b) + (3-2m-b) + (6-3m-b) + (8-4m-b)]$$
$$= -2(18-9m-4b)$$

Used Mathematica to solve this system, with result

$$m = \frac{14}{9}, b = 1$$

Corrected:

$$X = \begin{bmatrix} 1 & 0 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}$$

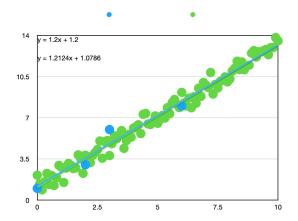
$$X^{\hat{}} = \mathbf{y}$$

3. Attached graph produced by Numbers.

×	L	.ine	Points	
	-5	-3.31		y = 1.77 x + 5.54
	10	23.24		
	0		1	
	2		3	
	3		6	
	6		8	
30 —				
22.5 R <sup>2</sup> = 0	).931			
15	J.931			
	J.931			
15	J.931			

## 4. Using (m, b) = (1.2, 1.2)

Generated 100 gaussian random numbers (mean = 0.0, variance = 1.0) from https://www.random.org/gaussian-distributions/?num=100&mean=0.0&stdev=0.5&dec=3&col=1&notation=scientific&format=html&rnd=new



(Data exported in csv format in git repository.)