

# CSE 847 Homework One

Reuben Lewis

05 February, 2021

## 1 Introduction

### 1.1 Pg 58, 1.3

Box	Apples	Oranges	Limes	Selection Prob $p$
$r$	3	4	3	0.2
$b$	1	1	0	0.2
$g$	3	3	4	0.6

(1) What is the probability of selecting is an apple?  $p(A) = p(A, r) + p(A, b) + p(A, g)$

$$p(A) = p(A|r) * p(r) + p(A|b) * p(b) + p(A|g) * p(g)$$

$$p(A) = 0.3 * 0.2 + 0.5 * 0.2 + 0.3 * 0.6$$

$$p(A) = 0.34$$

(2) If the selected fruit is an orange, what is the probability it came from the green box?

$$p(O) = p(O, r) + p(O, b) + p(O, g)$$

$$p(O) = p(O|r) * p(r) + p(O|b) * p(b) + p(O|g) * p(g)$$

$$p(O) = 0.4 * 0.2 + 0.5 * 0.2 + 0.3 * 0.6$$

$$p(O) = 0.36$$

$$p(g|O) = \frac{p(O|g)p(g)}{p(O)}$$

$$p(g|O) = \frac{0.3 * 0.6}{0.34}$$

$$p(g|O) = 0.5$$

### 1.2 Pg 59, 1.6

$$\text{cov}[x, y] = 0$$

$$\mathbb{E}[xy] - \mathbb{E}[x]\mathbb{E}[y] = 0$$

$$\mathbb{E}[xy] = \mathbb{E}[x]\mathbb{E}[y]$$

$$\sum_{x,y} xy P_{x,y}(x, y) = \sum_x x P(x) * \sum_y y P(y)$$

$$P(x, y) = P(x) * P(y)$$

This is the definition of independence, so, if  $x$  and  $y$  are independent, their covariance is going to be 0.

### 1.3 Pg 59, 1.11

Solve first for  $\mu_{ML}$ , then for  $\sigma^2$  to verify the results of (1.55) and (1.56). This involves first taking the partial derivative with respect to the value we are solving for, then setting that equation to 0 and solving for the value.

(1)  $\mu_{ML}$

$$\begin{aligned}
 & -\frac{1}{2\sigma^2} \sum_{n=1}^N (x_n - \mu)^2 - \frac{N}{2} \ln(\sigma^2) - \frac{N}{2} \ln(2\pi) \\
 & \frac{\partial}{\partial \mu} \Rightarrow -\frac{1}{2\sigma^2} \sum_{n=1}^N \frac{\partial}{\partial \mu} (x_n - \mu)^2 \\
 & -\frac{1}{2\sigma^2} \sum_{n=1}^N 2 * -1 * (x_n - \mu) \\
 & \quad \frac{1}{\sigma^2} \sum_{n=1}^N (x_n - \mu) \\
 & 0 = \frac{1}{\sigma^2} \sum_{n=1}^N (x_n - \mu) \\
 & 0 = \sum_{n=1}^N (x_n - \mu) \\
 & 0 = \sum_{n=1}^N x_n - N * \mu \\
 & N * \mu = \sum_{n=1}^N x_n \\
 & \mu_{ML} = \frac{1}{N} \sum_{n=1}^N x_n
 \end{aligned}$$

(2)  $\sigma^2$

$$\begin{aligned}
 & -\frac{1}{2\sigma^2} \sum_{n=1}^N (x_n - \mu)^2 - \frac{N}{2} \ln(\sigma^2) - \frac{N}{2} \ln(2\pi) \\
 & \frac{\partial}{\partial \sigma^2} \Rightarrow -\frac{\partial}{\partial \sigma^2} \frac{1}{2\sigma^2} \sum_{n=1}^N (x_n - \mu)^2 - \frac{\partial}{\partial \sigma^2} \frac{N}{2} \ln(\sigma^2) \\
 & \quad \frac{1}{\sigma^4} \sum_{n=1}^N (x_n - \mu)^2 - \frac{N}{2\sigma^2} \\
 & 0 = \frac{1}{2\sigma^4} \sum_{n=1}^N (x_n - \mu)^2 - \frac{N}{2\sigma^2} \\
 & \frac{N}{2\sigma^2} = \frac{1}{2\sigma^4} \sum_{n=1}^N (x_n - \mu)^2 \\
 & N * \sigma^2 = \sum_{n=1}^N (x_n - \mu)^2 \\
 & \sigma^2 = \frac{1}{N} \sum_{n=1}^N (x_n - \mu)^2
 \end{aligned}$$

## 2 Linear Algebra

1. Matrix Calculations

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 1 & 4 \end{bmatrix}, B = \begin{bmatrix} 1 & 0 \\ 2 & 1 \\ 3 & 2 \end{bmatrix}, C = \begin{bmatrix} 3 & -1 & 3 \\ 4 & 1 & 5 \\ 2 & 1 & 3 \end{bmatrix}, D = \begin{bmatrix} 2 & -4 & 5 \\ 0 & 1 & 4 \\ 3 & 2 & 1 \end{bmatrix}, E = \begin{bmatrix} 3 & -2 \\ 2 & 4 \end{bmatrix}.$$

(a)  $(2A)^T$  First, scale  $A$  by 2, then transpose it.

$$\begin{aligned}
 2 * A &= \begin{bmatrix} 2 & 4 & 6 \\ 4 & 2 & 8 \end{bmatrix} \\
 (2 * A)^T &= \begin{bmatrix} 2 & 4 \\ 4 & 2 \\ 6 & 8 \end{bmatrix}
 \end{aligned}$$

- (b)  $(A - B)^T$  Because  $A$  and  $B$  don't have the same dimension, this equation doesn't work.  $A - B^T$  would, however, since  $B$  would then have the same dimensions as  $A$ .
- (c)  $(3B^T - 2A)^T$  First, scale the matrices, then apply the transpose onto  $B$ . Next, subtract the matrices and transpose that output.

$$\begin{aligned}
 2 * A &= \begin{bmatrix} 2 & 4 & 6 \\ 4 & 2 & 8 \end{bmatrix} \\
 3 * B &= \begin{bmatrix} 3 & 0 \\ 6 & 3 \\ 9 & 6 \end{bmatrix} \\
 3 * B^T &= \begin{bmatrix} 3 & 6 & 9 \\ 0 & 3 & 6 \end{bmatrix} \\
 3 * B^T - 2 * A &= \begin{bmatrix} 1 & 2 & 3 \\ -4 & 1 & -2 \end{bmatrix} \\
 (3 * B^T - 2 * A)^T &= \begin{bmatrix} 1 & -4 \\ 2 & 1 \\ 3 & -2 \end{bmatrix}
 \end{aligned}$$

- (d)  $(-A)^T E$  First, distribute the  $-1 * A$ , and then transpose it. Next, multiply it by  $E$ , since their inner dimensions match ( $3 \times 2 * 2 \times 2$ ).

$$\begin{aligned}
 -1 * A &= \begin{bmatrix} -1 & -2 & -3 \\ -2 & -1 & -4 \end{bmatrix} \quad (-1 * A)^T = \begin{bmatrix} -1 & -2 \\ -2 & -1 \\ -3 & -4 \end{bmatrix} \\
 (-1 * A)^T E &= \begin{bmatrix} -1 * 3 + -2 * 2 & -2 * -1 + -2 * 4 \\ -2 * 3 + -1 * 2 & -2 * 2 + -1 * 4 \\ -3 * 3 + -4 * 2 & -3 * -2 + -4 * 4 \end{bmatrix} \\
 (-1 * A)^T E &= \begin{bmatrix} -7 & -6 \\ -8 & 0 \\ -17 & -10 \end{bmatrix}
 \end{aligned}$$

- (e)  $(C + D^T + E)^T$   
 Because  $E$  is not the same dimensions as  $C$  or  $D$ , it can not be added to the others, so this operation cannot be done.

2. Subspace of  $\mathbb{R}^2$ ? Justify your answer.

- (a)  $\{(x, y) \in \mathbb{R}^2 | x^2 + y^2 = 0\}$   
 This is not a subspace, since two vectors  $(1, 0) + (1, 1) = (2, 1)$  and show's it's not closed under addition.
- (b)  $\{(x, y) \in \mathbb{R}^2 | x^2 - y^2 = 0\}$   
 This is not a subspace, since two vectors  $(1, 1) + (1, -1) = (2, 0)$  and show's it's not closed under addition.
- (c)  $\{(x, y) \in \mathbb{R}^2 | x^2 - y = 0\}$   
 This is not a subspace, since two vectors  $(1, 1) + (1, -1) = (2, 0)$  and show's it's not closed under addition.

(d)  $\{(x, y) \in \mathbb{R}^2 | x - y = 0\}$

This is a subspace, since it satisfies the three main properties. The 0 vector is a part of the space, it's closed under vector addition, and it's closed under scalar multiplication. This is a special case of a line with a slope of 1.

Closed under Vector addition:

$$u = (x_1, y_1), v = (x_2, y_2)$$

$$u + v = (x_1 + x_2, y_1 + y_2)$$

$$u + v = (x_1 + x_2, x_1 + x_2)$$

This process is similar for scaling as well.

(e)  $\{(x, y) \in \mathbb{R}^2 | x - y = 1\}$

This is not a subspace because it doesn't contain the origin.

3.

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 2 \end{bmatrix}, B = \begin{bmatrix} 2 & -1 \\ -3 & 4 \end{bmatrix}.$$

Is  $AB = BA$ ? No.

$$AB = \begin{bmatrix} 1 * 2 + 2 * -3 & 1 * -1 + 2 * 4 \\ 3 * 2 + 2 * -2 & 3 * -1 + 2 * 4 \end{bmatrix}$$

$$AB = \begin{bmatrix} -4 & 7 \\ 2 & 5 \end{bmatrix}$$

$$BA = \begin{bmatrix} 2 * 1 + -1 * 3 & 2 * 2 + -1 * 2 \\ -3 * 1 + 4 * 3 & -3 * 2 + 4 * 2 \end{bmatrix}$$

$$BA = \begin{bmatrix} -1 & 2 \\ 9 & 2 \end{bmatrix}$$

4. (a) Because of how matrix multiplication works, you multiply the rows of the first matrix by the columns of the second. This means that if there is a row of 0s in the first matrix, that row would be multiplied against every column of the second matrix. This would cause a row of 0s in the resulting matrix, since multiplying any value by 0 results in 0. As an example, imagine  $A = 2 \times 3$  matrix and  $B = 3 \times 4$ , with a row of 0s at  $i = 2$ . Row 2 of the resulting matrix would be calculated like so:

$$\begin{aligned} A_{2k} &= \sum_j 0_{2,j} * b_{j,k} \\ [0 * b_{0,0} + 0 * b_{1,0} + 0 * b_{2,0} & \quad 0 * b_{0,1} + 0 * b_{1,1} + 0 * b_{2,1}] \\ [0 & \quad 0] \end{aligned}$$

- (b) The best way to show this is through an example. Let

$$A = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}, B = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

If we multiply  $BA$ , then we get the following matrix. This is because when we multiply rows of  $B$  through the columns of  $A$ , the column of 0s in  $A$  nullifies the values in the row of  $B$ , resulting in a column of 0s in the resulting matrix.

$$BA = \begin{bmatrix} 1 * 1 + 2 * 1 & 1 * 0 + 2 * 0 \\ 3 * 1 + 4 * 1 & 3 * 0 + 4 * 0 \end{bmatrix}$$

$$BA = \begin{bmatrix} 3 & 0 \\ 7 & 0 \end{bmatrix}$$

5. Show that  $\text{rank}(AB) \leq \min\{\text{rank}(A), \text{rank}(B)\}$ .

Remember that  $\text{rank}(AB) = \dim(\text{ran}(AB))$ ,  $\text{rank}(A) = \dim(\text{ran}(A))$   
 By definition,  $\text{ran}(AB) = \{y \in \mathbb{R}^n : y = ABx \text{ for some } x \in \mathbb{R}^n\}$  Substitute  
 $z = Bx \Rightarrow y = A(Bx) \Rightarrow y \in \mathbb{R}^n : y = Az \text{ for some } z \in \mathbb{R}^n = \text{ran}(A)$

This shows that the vector space  $\text{ran}(AB)$  is a subspace of  $\text{ran}(A)$ , which allows us to say  $\text{rank}(AB) = \dim(\text{ran}(AB)) \leq \dim(\text{ran}(A)) = \text{rank}(A)$ .  
 Now what about  $\text{rank}(B)$ ?

Remember that  $\text{rank}(A) = \text{rank}(A^T)$  Let  $C = A^T$  and  $D = B^T$   
 $\text{rank}(DC) \leq \text{rank}(D) \Rightarrow \text{rank}(C^T D^T) \leq \text{rank}(D^T)$  Translating that  
 back over, we get  $\text{rank}(AB) \leq \text{rank}(B)$

So, by that logic,  $\text{rank}(AB)$  is less than or equal to whichever of  $A$  or  $B$  has the lower rank.