# COMP1942 Exploring and Visualizing Data (Spring Semester 2024) Homework 2 Solution

Full Mark: 100 Marks

# **Q1 [20 Marks]**

(a) No. This new customer will not go to Disneyland.

(b) (i)

No. of Children	No. of Siblings	Go_Disneyland			
2	0	0			
0	2	0			
4	2	1			
2	4	1			

(ii)

### <u>Iteration 1</u>

$$x_1 x_2 y$$
  
(2, 0, 0)

b	$\mathbf{w}_1$	$W_2$
0.3	0.3	0.3

net = 
$$x_1w_1 + x_2w_2 + b$$
  
=  $2 \times 0.3 + 0 \times 0.3 + 0.3$   
=  $0.9$ 

$$w_1 = w_1 + \alpha (d - y) x_1$$
  
= 0.3 + 0.6 × (0 - 1) × 2  
= -0.9

$$w_2 = w_2 + \alpha (d - y) x_2$$
  
= 0.3 + 0.6 × (0 - 1) × 0  
= 0.3

$$b = b + \alpha (d - y)$$
  
= 0.3 + 0.6 × (0 - 1)  
= -0.3

## Iteration 2

$$x_1 x_2 y$$
  
(0, 2, 0)

ь	$\mathbf{w}_1$	$W_2$
-0.3	-0.9	0.3

net 
$$= x_1w_1 + x_2w_2 + b = 0.3$$

$$y = 1$$
 (Incorrect)

$$\begin{array}{ll} w_1 & = w_1 + \alpha (d - y) x_1 \\ & = -0.9 + 0.6 \times (0 - 1) \times 0 \\ & = -0.9 \end{array}$$

$$\begin{array}{ll} w_2 & = w_2 + \alpha \ (d - y) \ x_2 \\ & = 0.3 + 0.6 \times (0 \text{-} 1) \times 2 \\ & = \text{-} 0.9 \end{array}$$

$$\begin{array}{ll} b & = b + \alpha \ (d - y) \\ & = -0.3 + 0.6 \times (0 - 1) \\ & = -0.9 \end{array}$$

# <u>Iteration 3</u>

$$x_1 x_2 y$$
 (4, 2, 1)

b	$\mathbf{w}_1$	W <sub>2</sub>
-0.9	-0.9	-0.9

net = 
$$x_1w_1 + x_2w_2 + b$$
  
= -6.3

$$y = 0$$
 (Incorrect)

$$\begin{array}{ll} w_1 & = w_1 + \alpha (d - y) x_1 \\ & = -0.9 + 0.6 \times (1 - 0) \times 4 \\ & = 1.5 \end{array}$$

$$\begin{array}{ll} w_2 & = w_2 + \alpha (d - y) x_2 \\ & = -0.9 + 0.6 \times (1 - 0) \times 2 \\ & = 0.3 \end{array}$$

b = 
$$b + \alpha (d - y)$$
  
=  $-0.9 + 0.6 \times (1 - 0)$   
=  $-0.3$ 

# <u>Iteration 4</u>

 $x_1 x_2 y$ (2, 4, 1)

ь	$\mathbf{w}_1$	W <sub>2</sub>
-0.3	1.5	0.3

net 
$$= x_1w_1 + x_2w_2 + b$$
  
= 3.9

$$w_1 = w_1 + \alpha (d - y) x_1$$
  
= 1.5

$$w_2 = w_2 + \alpha (d - y) x_2$$
  
= 0.3

b = 
$$b + \alpha (d - y)$$
  
= -0.3

## <u>Iteration 5</u>

 $x_1 x_2 y$ (2, 0, 0)

b	$\mathbf{w}_1$	W2
-0.3	1.5	0.3

net 
$$= x_1w_1 + x_2w_2 + b$$
  
= 2.7

$$\begin{array}{ll} w_1 & = w_1 + \alpha (d - y) x_1 \\ & = 1.5 + 0.6 \times (0 - 1) \times 2 \\ & = 0.3 \end{array}$$

$$\begin{array}{ll} w_2 & = w_2 + \alpha \ (d - y) \ x_2 \\ & = 0.3 + 0.6 \times (0 - 1) \times 0 \\ & = 0.3 \end{array}$$

b = 
$$b + \alpha (d - y)$$
  
=  $-0.3 + 0.6 \times (0 - 1)$   
=  $-0.9$ 

ь	$\mathbf{w}_1$	$W_2$
-0.9	0.3	0.3

## **Q2 [20 Marks]**

For data (6, 6), difference from mean vector = 
$$\begin{pmatrix} 6-7 \\ 6-7 \end{pmatrix} = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$$

For data (8, 8), difference from mean vector 
$$=$$
  $\begin{pmatrix} 8-7 \\ 8-7 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ 

For data (5, 9), difference from mean vector 
$$=$$
  $\begin{pmatrix} 5-7 \\ 9-7 \end{pmatrix} = \begin{pmatrix} -2 \\ 2 \end{pmatrix}$ 

For data (9, 5), difference from mean vector 
$$=$$
  $\begin{pmatrix} 9-7 \\ 5-7 \end{pmatrix} = \begin{pmatrix} 2 \\ -2 \end{pmatrix}$ 

$$Y = \begin{pmatrix} -1 & 1 & -2 & 2 \\ -1 & 1 & 2 & -2 \end{pmatrix}$$

$$\Sigma = \frac{1}{4}YY^{T} = \frac{1}{4} \begin{pmatrix} -1 & 1 & -2 & 2 \\ -1 & 1 & 2 & -2 \end{pmatrix} \begin{pmatrix} -1 & -1 \\ 1 & 1 \\ -2 & 2 \\ 2 & -2 \end{pmatrix}$$

$$=\frac{1}{4} \begin{pmatrix} 10 & -6 \\ -6 & 10 \end{pmatrix}$$

$$= \begin{pmatrix} \frac{5}{2} & -\frac{3}{2} \\ -\frac{3}{2} & \frac{5}{2} \end{pmatrix}$$

$$\begin{vmatrix} \frac{5}{2} - \lambda & -\frac{3}{2} \\ -\frac{3}{2} & \frac{5}{2} - \lambda \end{vmatrix} = 0 \quad \Longrightarrow \quad \left(\frac{5}{2} - \lambda\right)^2 - \left(-\frac{3}{2}\right)^2 = 0 \quad \Longrightarrow \quad \lambda = 4 \quad or \quad \lambda = 1$$

when  $\lambda = 4$ ,

$$\begin{pmatrix} \frac{5}{2} - 4 & -\frac{3}{2} \\ -\frac{3}{2} & \frac{5}{2} - 4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{pmatrix} -\frac{3}{2} & -\frac{3}{2} \\ -\frac{3}{2} & -\frac{3}{2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow x_1 + x_2 = 0$$

We choose the eigenvector of unit length:  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ -\frac{\sqrt{2}}{2} \end{pmatrix}$ .

When  $\lambda = 1$ ,

$$\begin{pmatrix} \frac{5}{2} - 1 & -\frac{3}{2} \\ -\frac{3}{2} & \frac{5}{2} - 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{pmatrix} \frac{3}{2} & -\frac{3}{2} \\ -\frac{3}{2} & \frac{3}{2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow x_1 - x_2 = 0$$

We choose the eigenvector of unit length:  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{pmatrix}$ .

Thus, 
$$\Phi = \begin{pmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \\ -\frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix}$$
,  $Y = \Phi^T X = \begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix} X$ .

For data (6, 6), 
$$Y = \begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix} \begin{pmatrix} 6 \\ 6 \end{pmatrix} = \begin{pmatrix} 0 \\ 8.49 \end{pmatrix}$$

For data (8, 8), 
$$Y = \begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix} \begin{pmatrix} 8 \\ 8 \end{pmatrix} = \begin{pmatrix} 0 \\ 11.31 \end{pmatrix}$$

For data (5, 9), 
$$Y = \begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix} \begin{pmatrix} 5 \\ 9 \end{pmatrix} = \begin{pmatrix} -2.83 \\ 9.90 \end{pmatrix}$$

For data (9, 5), 
$$Y = \begin{pmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \end{pmatrix} \begin{pmatrix} 9 \\ 5 \end{pmatrix} = \begin{pmatrix} 2.83 \\ 9.90 \end{pmatrix}$$

Thus, (6, 6) is reduced to (0);

(8, 8) is reduced to (0);

(5, 9) is reduced to (-2.83);

(9, 5) is reduced to (2.83).

(Note: Another possible answer is

(6, 6) is reduced to (0);

(8, 8) is reduced to (0);

(5, 9) is reduced to (2.83);

(9, 5) is reduced to (-2.83).

This is because the eigenvectors used in this case are:

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \end{pmatrix}$$

(b)

- (2, 6) is reduced to (2.83); (3, 3) is reduced to (0); (5, 5) is reduced to (0); (6, 2) is reduced to (-2.83).

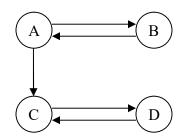
# Q3 [20 Marks]

The greedy algorithm discussed in class can be modified be changing the heuristics function from the computation of the benefit of a view to the computation of the benefit of a view per "unit space".

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i.e.
Let C(v) be the cost of view v(the number of rows in v)
Algorithm:
S \leftarrow \{top \ view\};
X \leftarrow X - C(v) where v is the top view;
While there exists a view v \in S \ s.t. \ C(v) \le X
Select the view v \in S \ s.t.
C(v) \le X
B(v,S)/C(v) is maximized
S \leftarrow S \cup \{v\}
X \leftarrow X - C(v)
output S.
```

# **Q4** [20 Marks]

(a)



Stochastic matrix:

(b) Equation to be solved:

$$\begin{pmatrix} A \\ B \\ C \\ D \end{pmatrix} = 0.8 \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0.5 & 0 & 0 & 0 \\ 0.5 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} A \\ B \\ C \\ D \end{pmatrix} + \begin{pmatrix} 0.2 \\ 0.2 \\ 0.2 \\ 0.2 \end{pmatrix}$$

(c)

(6)									
	1	2	3	4	5	6	7		13
A	1	1.0	0.68	0.68	0.58	0.58	0.54		0.53
В	1	0.6	0.6	0.47	0.47	0.43	0.43		0.41
С	1	1.4	1.4	1.53	1.53	1.57	1.57	•••	1.59
D	1	1.0	1.32	1.32	1.42	1.42	1.46		1.47

So the ranking is C, D, A, B.

### **Q5** [20 Marks]

(a)
$$P(SIR = Yes \mid AP = Yes, P = Yes, WBC = High)$$

$$= \frac{P(WBC = High \mid AP = Yes, P = Yes, SIR = Yes)}{P(WBC = High \mid AP = Yes, P = Yes)} P(SIR = Yes \mid AP = Yes, P = Yes)$$

$$= \frac{P(WBC = High \mid SIR = Yes) P(SIR = Yes \mid AP = Yes, P = Yes)}{\sum_{x \in \{Yes, No\}} P(WBC = High \mid SIR) P(SIR) P(SIR)} P(SIR) P(SIR) P(SIR)$$

$$= \frac{0.6*0.7}{0.6*0.7+0.3*0.3}$$

$$= 0.8235$$

$$P(SIR = No \mid AP = Yes, P = Yes, WBC = High) = 1-0.8235 = 0.1765$$

Since P(SIR = Yes | AP = Yes, P = Yes, WBC = High) > P(SIR = No | AP = Yes, P = Yes, WBC = High), it is more likely that the person has systemic inflammation reaction.

#### (b) Disadvantages:

The Bayesian Belief network classifier requires a predefined knowledge about the network. The Bayesian Belief Network classifier cannot work directly when the network contains cycles.