

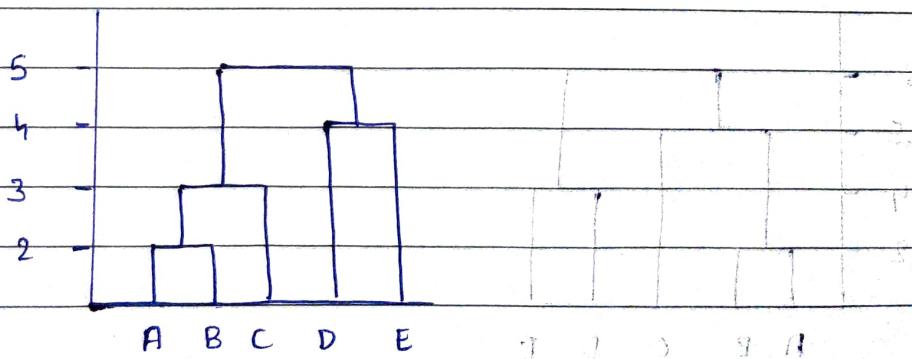
18DCS007

Q1. Applying single link clustering for above problem

	A	B	C	D	E	F	G	H	I	J	K	L
A	0	0										
B	2	9										
C	6	9	0									
D	10	8	7	0	0	F	P	9	d			
E	9	2	5	4	0			3	P	J		

	AB	C	D	E	F	G	H	JI
AB	0							
C	3	0						
D	9	7	0					
E	8	5	4	0	0	F	P	JI

	ABC	DE	E	JI	JI
ABC	0				
DE	5	0			



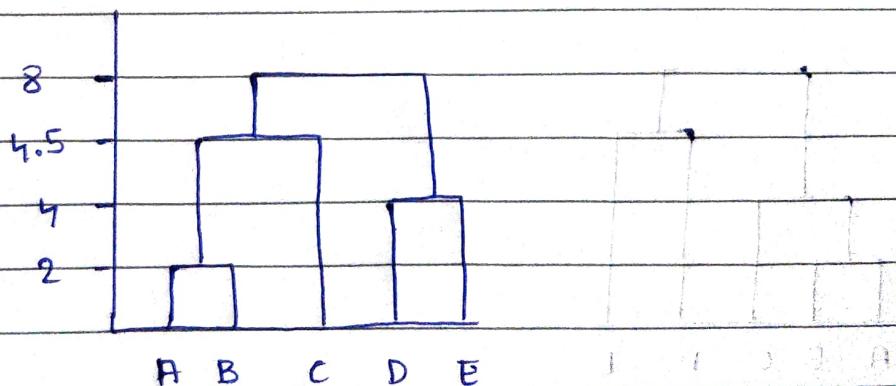
(2)

Applying average link clustering (ii) step by step

	A	B	C	D	E	I	J	K	L
A	0					0	0	0	0
B	2	0				0	0	0	0
C	6	3	0			0	0	0	0
D	10	9	7	0		0	0	0	0
E	9	8	5	4	0	0	0	0	0

	AB	C	D	E	I	J	K	L
AB	0				0	0	0	0
C	4.5	0			0	0	0	0
D	9.5	7	0		0	0	0	0
E	8.5	5	4	0	0	0	0	0

	ABC	DE	I	J	K	L
ABC	0				0	0
DE	8	0			0	0



(3)

Q2. Computer applications for which ML is appropriate

- (1) Stock prediction system / - prediction system which has high uncertainty & large data
- (2) Intrusion / Fraud Detection / - pattern recognition plays imp. role
- (3) Spam Filtering

→ Computer applications for which ML is inappropriate

- (1) Shortest path finding - we already have great algorithm like A* and dijkstra's algorithm
- (2) Simple face detection - image processing is capable of it
- (3) Rule based systems/games - We don't need AI to play tic-tac.

Q3. Min support = 30%, min confidence = 80%.

ITEM COUNT Support Threshold = $0.3 \times 7 = 2.1/3$

Button

4

$0.3 \times 7 = 2.1/3$

Bread

5

→ It shows biscuits does not meet

Milk

4

min support so it is deleted

Cheese

4

Biscuits

1

Clothes

3

(4)

→ JOIN STEP 1: (1) ITEMS COUNT in frequent

Button, Bread 3

Button, Milk 2

Button, Cheese 3

Button, Clothes 1

Bread, Milk 3

Bread, Cheese 2

Bread, Clothes 3

Milk, Cheese 1

Milk, Clothes 3

Cheese, Clothes 1

→ PRIME STEP: (2) ITEMS COUNT in frequent

Button, Bread 3

Button, Cheese 3

Bread, Milk 3

Bread, Clothes 3

Milk, Clothes 3

→ JOIN STEP: ITEMS COUNT

Button, Bread, Milk 2

Button, Bread, Cheese 2

Button, Bread, Clothes 0

Bread, Milk, Cheese 1

Bread, Milk, Clothes 3

Bread, Cheese, Clothes 1

{Bread, Milk, Clothes}

∴ Frequent item set = {Bread, Milk, Clothes}

(5)

→ Association Rule Mining - A technique for mining frequent item sets.

$$\{ \text{Bread, Clothes} \} \rightarrow \{ \text{Milk} \}$$

$$\text{confidence} = \frac{\text{support} \{ \text{Bread, Clothes, Milk} \}}{\text{support} \{ \text{Bread, Clothes} \}} = \frac{3}{3} = 100\%$$

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$$\{ \text{Bread, Milk} \} \rightarrow \{ \text{Clothes} \}$$

$$\text{confidence} = \frac{\text{support} \{ \text{Bread, Clothes, Milk} \}}{\text{support} \{ \text{Bread, Milk} \}} = \frac{3}{4} = 75\% < 80\%$$

$$\text{confidence} = \frac{\text{support} \{ \text{Bread, Milk} \}}{\text{support} \{ \text{Clothes} \}} = \frac{3}{4} = 75\% < 80\%$$

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$$\{ \text{Clothes, Milk} \} \rightarrow \{ \text{Bread} \}$$

$$\text{confidence} = \frac{\text{support} \{ \text{Bread, Clothes, Milk} \}}{\text{support} \{ \text{Clothes, Milk} \}} = \frac{3}{3} = 100\% > 80\%$$

∴ Strong association rules -

$$\{ \text{Bread, Clothes} \} \Rightarrow \{ \text{Milk} \}$$

$$\{ \text{Clothes, Milk} \} \Rightarrow \{ \text{Bread} \}$$

Q4. In feed forward networks, inputs are fed to the network and transformed into an output.

They are used for classification.

(6)

In Feed forward networks, predictions does not depend on the predictions in the previous moment.

But the situation is different in recurrent neural networks. The decision the recurrent neural net gets at this moment, depends on the decisions which the net got at the previous moment.

Moreover, the inputs and outputs to a feed forward network should be two dimensional with shape [no. of examples / input / output size]

And input and outputs for Recurrent neural network should be three dimensional with shape [no. of examples, input size / time series length]

Q5:

x	2	3	5	7	9	11	13	15
y	4	5	7	10	15	17	20	25

→ We know Linear regression hypothesis

$$h_0(x) = \theta_0 + \theta_1 x$$

→ Let's consider $y = a + bx$

→ Using method of least squares to find out the best fit

$$\Sigma y = a \Sigma 1 + b \Sigma x$$

$$\Sigma xy = a \Sigma x + b \Sigma x^2 \quad (\text{multiply with } x)$$

(7)

→ Calculating parameters (1) $\sum x = 26$ (2) $\sum x^2 = 263$

$$\sum y = 41 \quad \sum y^2 = 263$$

$$Equation 1: 5a + 26b = 41$$

$$Equation 2: 26a + 168b = 263$$

$$Hence \begin{cases} a = 0.305 \\ b = 1.518 \end{cases}$$

→ By solving the equation, $a = 0.305$

$$b = 1.518$$

$$\rightarrow \text{For } x = 8.3, y = 0.305 + 1.518 \\ = 12.905$$

$$= 13$$

$$Q6. H(\text{Diabetic}) = -\sum p(x_i) \log_2 p(x_i)$$

$$= -\frac{6}{9} \log_2 \left(\frac{6}{9}\right) - \frac{3}{9} \log_2 \left(\frac{3}{9}\right)$$

$$= -0.66(-0.599) - 0.33(-1.599)$$

$$= 0.9230$$

Weight

/ 80

$$H(\text{weight}) = -\frac{1}{5} \log_2 \left(\frac{1}{5}\right) - \frac{4}{5} \log_2 \left(\frac{4}{5}\right)$$

$$< 80 \geq 80$$

$$= 0.7219$$

YES/NO YES NO

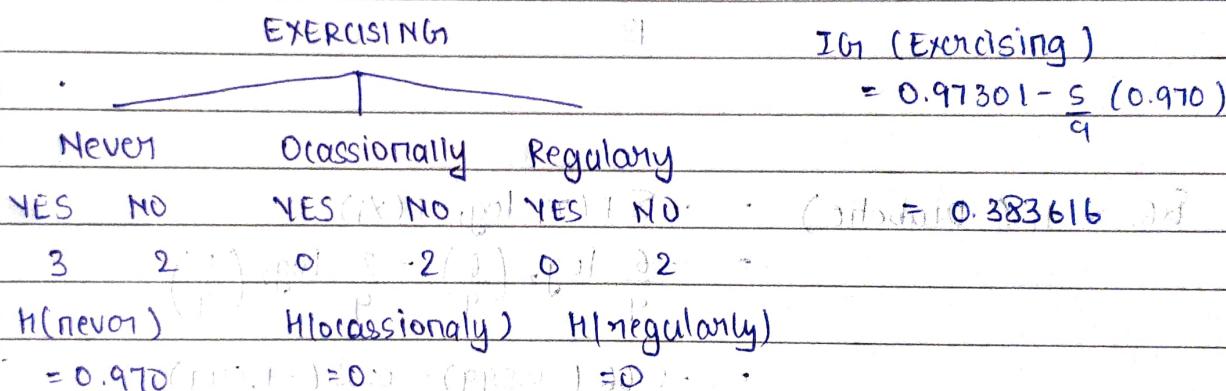
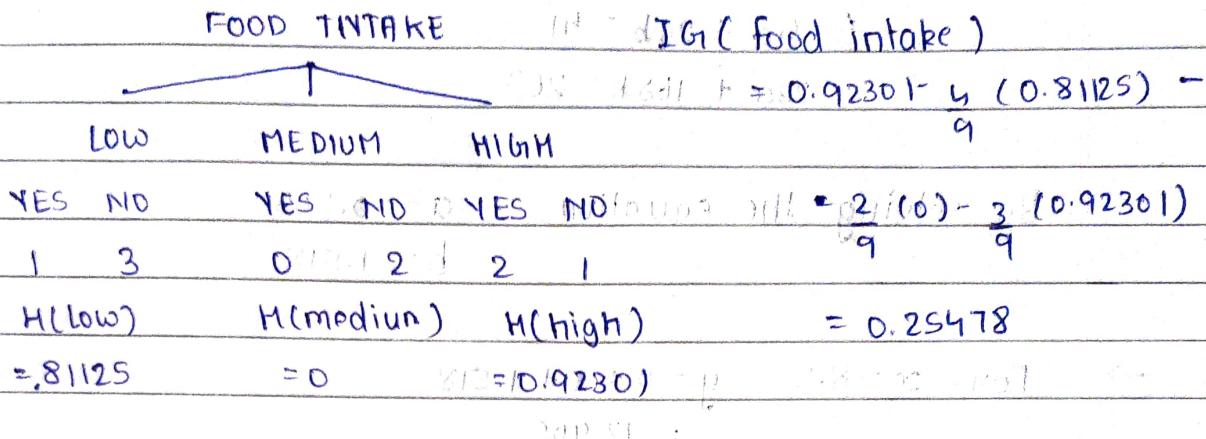
$$1 \oplus 4 \quad 2 \quad 2 \oplus (\text{weight} > 80) = 1$$

$$IG(\text{Diabetic, Weight}) = H(\text{Diabetic}) = \frac{1}{5} [w(\text{Weight} < 80) -$$

$$\frac{4}{5} H(\text{Weight} > 80)]$$

(8)

$$= 0.92301 - \frac{5}{9} (0.7219) - \frac{4}{9}(1) = 0.25478$$



→ Exercise will be root node.

IG₁ (Diabetic, exercising) is highest amongst all

Q7.

C₁ C₂ C₃

K=3

$$A_1 \quad 0 \quad 3.6 \rightarrow C_1 \quad C_1 = A_1$$

$$A_2 \quad 3.6 \quad 0 \quad 5.38 \rightarrow C_2 \quad C_2 = A_2$$

$$A_3 \quad 5.38 \quad 1.7 \quad 0 \rightarrow C_1 \quad C_3 = A_3$$

$$A_4 \quad 5.1 \quad 2.23 \quad 8.06 \rightarrow C_2$$

$$A_5 \rightarrow 3.16 \quad 4.12 \quad 2.23 \rightarrow C_3 \quad (\text{Unlabelled, unknown})$$

$$A_6 \quad 5 \quad 2 \quad 5.1 \rightarrow C_2$$

$$A_7 \rightarrow 6.08 \quad 4.47 \quad 3.16 \rightarrow C_3$$

$$A_8 \quad 8.24 \quad 6.08 \quad 5 \rightarrow C_3$$

(9)

→ New centroid values for each cluster? \rightarrow $C_1 = (1, 2)$

$$C_1 = (1, 2)$$

$$C_2 = ((3+2+5)/3, (5+7+5)/3) = (3.33), (5.66)$$

$$C_3 = ((6+4+7+9)/4, (0+1+3+4)/4) = (6.5, 2)$$

→ Taking distance from new centroids? \rightarrow $C_1 = (1, 2)$

	C_1	C_2	C_3	
A1	0	4.83	5.5	$\rightarrow C_1$
A2	3.6	0.73	4.6	$\rightarrow C_2$
A3	5.38	6.25	2.06	$\rightarrow C_3$
A4	5.1	1.88	6.72	$\rightarrow C_2$
A5	3.16	4.70	2.69	$\rightarrow C_4$
A6	5	1.79	3.35	$\rightarrow C_2$
A7	6.08	4.53	1.118	$\rightarrow C_3$
A8	8.246	5.9	3.2	$\rightarrow C_3$

→ No changes.
 Cluster 1 = $C_1 = (1, 2) = \{A1\}$
 Cluster 2 = $C_2 = (3.33, 5.66) = \{A2, A4, A6\}$
 Cluster 3 = $C_3 = (6.5, 2) = \{A3, A5, A7, A8\}$

Q8. K=5, Manhattan Distance :

	8	9
1	(19)	(17)
2	(18)	(34)
3	21	35
4	26	38
5	(11)	(5)
6	(7)	(19)
7	(7)	(23)

(10)

- The closest 5 neighbours of 8 are $\{1, 2, 5, 6, 7\}$
 which means Tiger: 3
 Leopard: 2
 $\therefore 8$ will be classified as tiger acc to KNN
- The closest 5 neighbours of 9 are $\{1, 2, 5, 6, 9\}$
 which means Tiger: 3
 Leopard: 2
 $\therefore 9$ will be classified as tiger acc to KNN