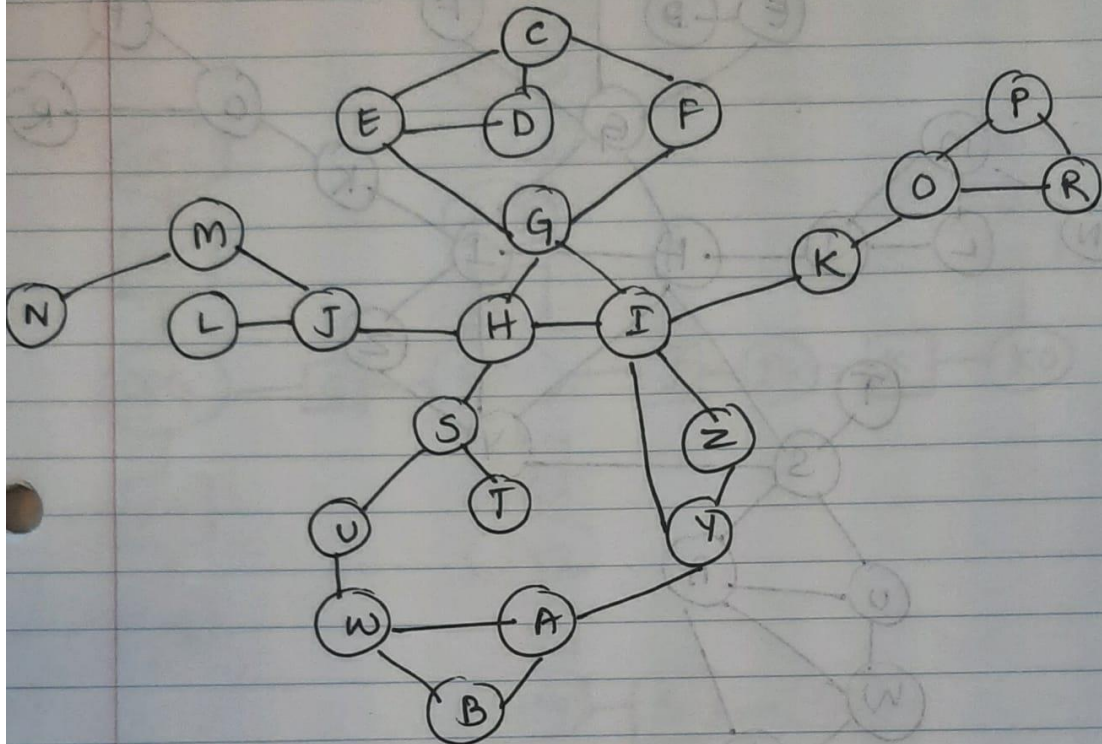


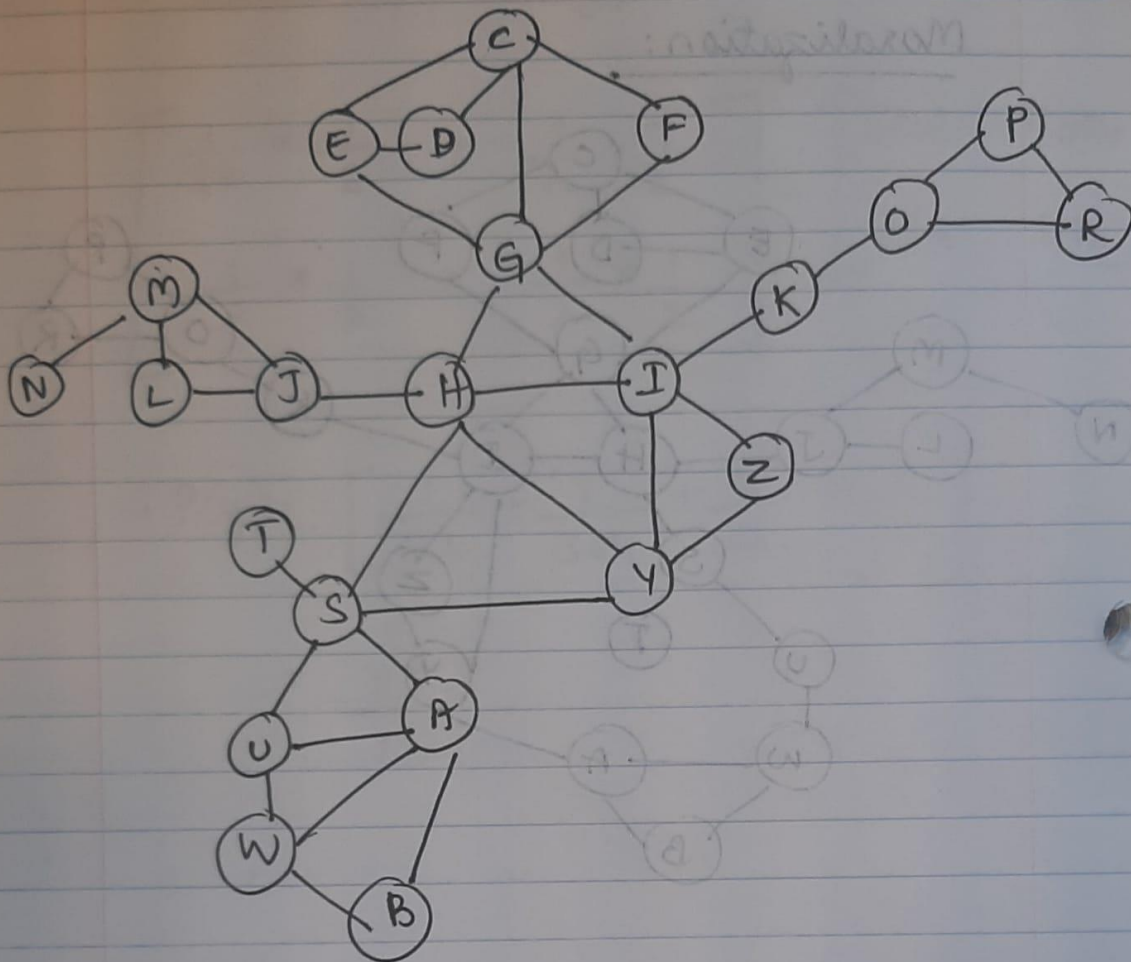
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Question-1:

Moralization:

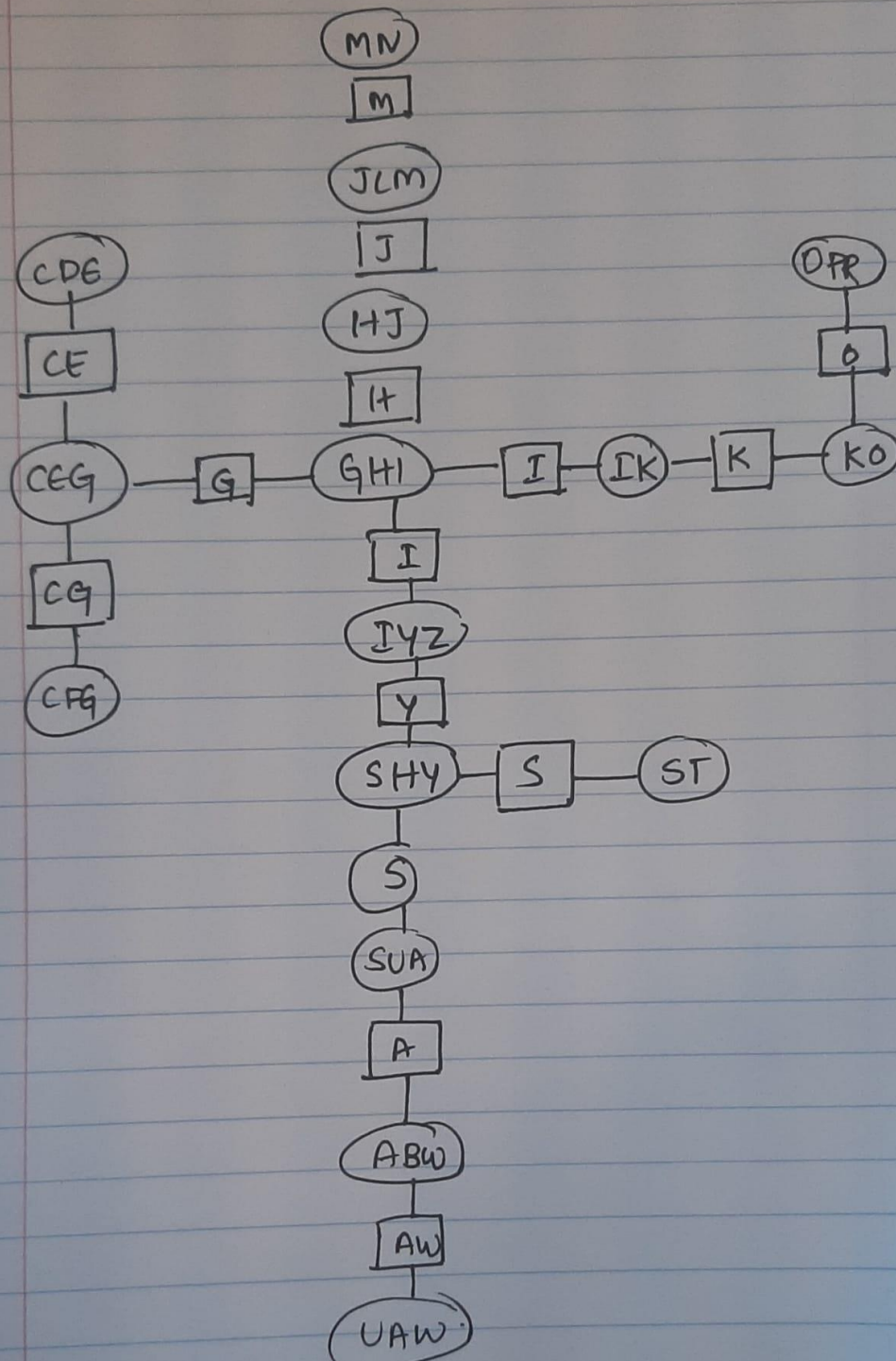


Triangulation:



Cliques: CDE, CEG, CFG, GHI, MN, MLJ,  
 JH, IK, KO, OPR, IZY, SHY, ST, SUA,  
 ABW, UAW.

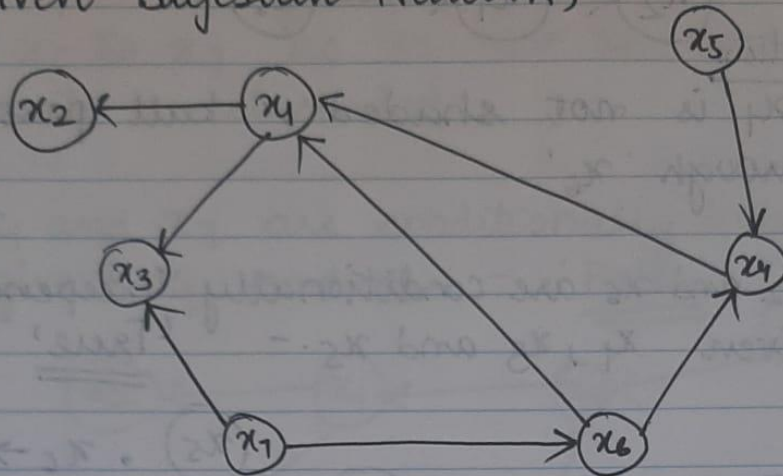
## Junction Tree:





Question 2:

Given Bayesian Network,



Factorization of probability distribution can be given as,

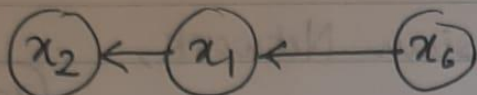
$$\begin{aligned}
 P(x_1, x_2, x_3, x_4, x_5, x_6, x_7) &= \prod_{i=1}^n P(x_i | pa_i) \\
 &= \prod_{i=1}^n P(x_i | \pi_i) = \prod_{i=1}^n P(x_i | \pi_i)
 \end{aligned}$$

$$P(x_1, x_2, x_3, x_4, x_5, x_6, x_7) = P(x_1 | x_4 x_6).$$

$$P(x_2 | x_1) P(x_3 | x_1 x_7) P(x_4 | x_5 x_6) P(x_5)$$

$$P(x_6 | x_7) P(x_7)$$

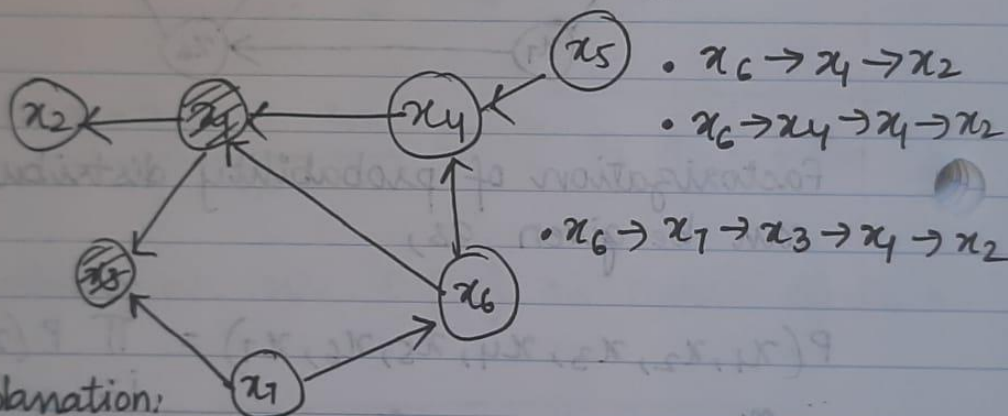
(a)  $x_2$  and  $x_6$  are independent - 'False'



Explanation:

$x_1$  is not shaded & ball passes through ' $x_6$ '

(b)  $x_2$  and  $x_6$  are conditionally independent given  $x_1, x_3$  and  $x_5$ . - 'True'



- $x_6 \rightarrow x_4 \rightarrow x_2$
- $x_6 \rightarrow x_4 \rightarrow x_3 \rightarrow x_1 \rightarrow x_2$
- $x_6 \rightarrow x_1 \rightarrow x_3 \rightarrow x_4 \rightarrow x_2$

Explanation:

- When chosen  $x_6 \rightarrow x_1 \rightarrow x_2$ .

Ball passes from  $x_1$  (not shaded  $\therefore$  Markov chain).

- When chosen  $x_6 \rightarrow x_4 \rightarrow x_1 \rightarrow x_2$

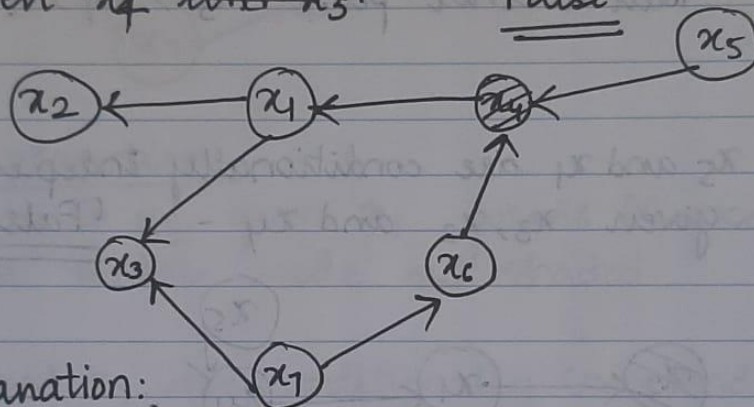
Ball moves from  $x_4$  to  $x_2$  passing through  $x_1$

• When chosen  $x_6 \rightarrow x_7 \rightarrow x_3 \rightarrow x_4 \rightarrow x_2$

$\rightarrow x_7$  to  $x_4$  from  $x_3$

$\rightarrow x_6$  to  $x_7$  as  $x_3, x_7 \neq x_6$  and middle node is not shaded.

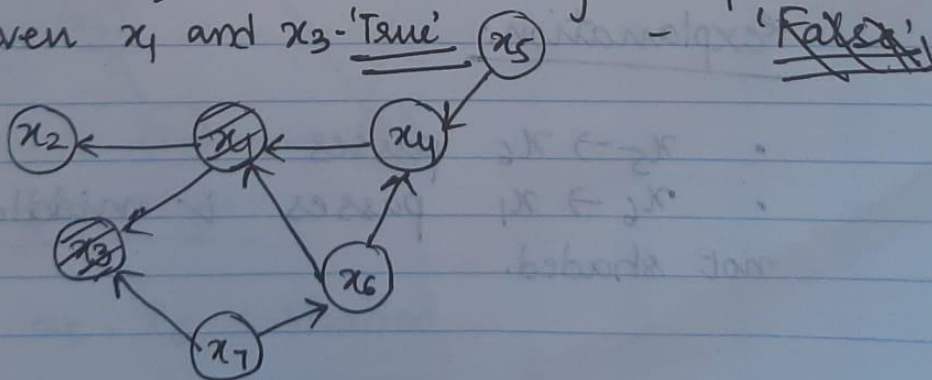
(c)  $x_1$  and  $x_7$  are conditionally independent given  $x_4$  and  $x_3$ . - 'False'



Explanation:

•  $x_7, x_6$  and  $x_4$  form Markov chain and middle node is not shaded here.

(d)  $x_5$  and  $x_2$  are conditionally independent given  $x_4$  and  $x_3$ . - 'True'

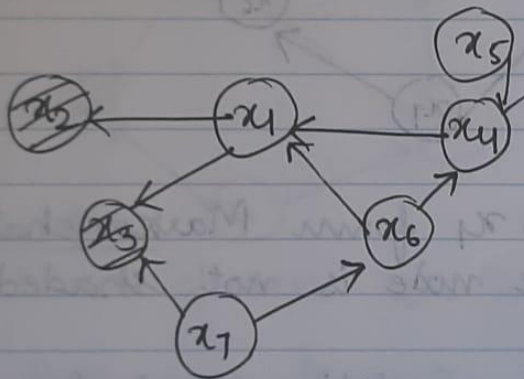




Explanation:

- $x_4 \rightarrow x_1 \rightarrow x_2$  Markov chain with middle node shaded
- $x_5 \rightarrow x_4 \rightarrow x_6$  middle node not shaded
- Ball does not pass  $x_5 \rightarrow x_4 \rightarrow x_6 \rightarrow x_7 \rightarrow x_3 \rightarrow x_2$   
 $\downarrow$   
 $x_2$ .

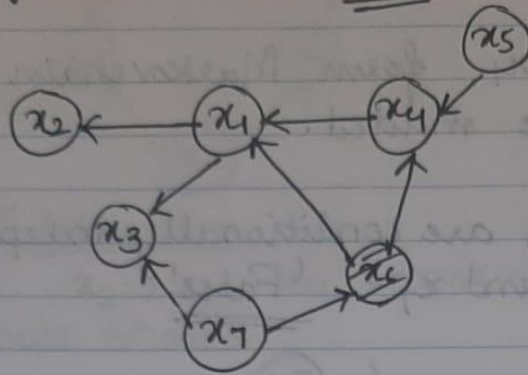
(e)  $x_5$  and  $x_1$  are conditionally independent given  $x_3, x_2$  and  $x_4$  - 'False'



Explanation:

- $x_5 \rightarrow x_6$  passes
- $x_6 \rightarrow x_1$  passes & middle node not shaded.

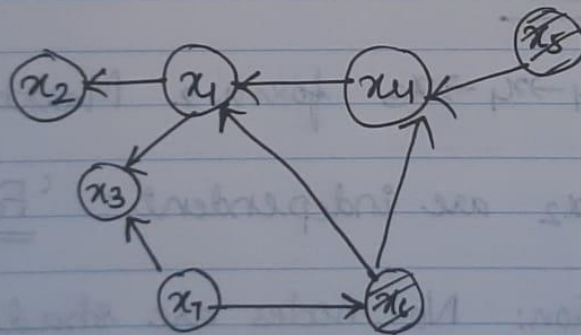
(f)  $x_4$  and  $x_3$  are conditionally independent given  $x_6$  - 'False'



Explanation:

- $x_4$  to  $x_3$  from  $x_1$  forms Markov chain with middle node not shaded.

(g)  $x_2$  and  $x_7$  are conditionally independent given  $x_5$  and  $x_6$  - 'True'



Explanation:

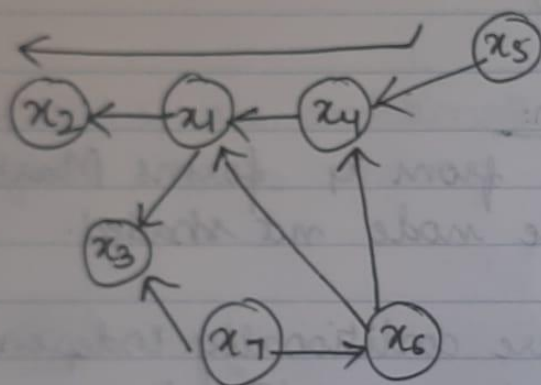
- $x_5, x_6$  : shaded



- Doesn't pass  $x_2 \rightarrow x_1 \rightarrow x_3 \rightarrow x_7$  middle node not shaded.

- $x_2 \rightarrow x_6 \rightarrow x_1$  form Markov chain & middle node shaded.

(b)  $x_5$  and  $x_3$  are conditionally independent given  $x_6$  and  $x_7$  - 'False'



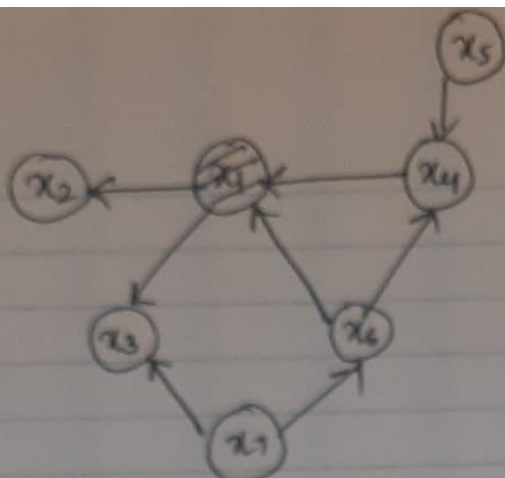
Explanation:

- $x_5 \rightarrow x_4 \rightarrow x_1 \rightarrow x_3$  form 2 Markov chains

(i)  $x_5$  and  $x_2$  are independent - 'False'

Explanation: No nodes are shaded.

(j)  $x_2$  and  $x_4$  are conditionally independent given  $x_1$  - 'True'



Explanation:

- $x_2 \rightarrow x_1 \rightarrow x_4$  Markov chain with middle node shaded.
- $x_2 \rightarrow x_1 \rightarrow x_3$  : doesnot pass
- $x_2 \rightarrow x_1 \rightarrow x_6 \rightarrow x_4$  : doesnot pass
- $x_2 \rightarrow x_1 \rightarrow x_6$  : Markov chain with middle node shaded.

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### Question-3

Given,

$$\pi = p(q_0) = \begin{matrix} & 1 & 2 \\ \begin{bmatrix} 1/3 & 2/3 \end{bmatrix} \end{matrix}$$

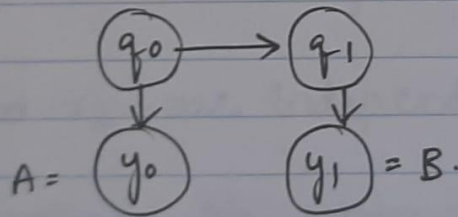
$$a^T = p(q_t | q_{t-1}) = \begin{matrix} & 1 & 2 \\ \begin{matrix} 1 \\ 2 \end{matrix} \begin{bmatrix} 1/8 & 1/2 \\ 7/8 & 1/2 \end{bmatrix} \end{matrix}$$

$$n^T = p(y_t | q_t) = \begin{matrix} & A & B \\ \begin{matrix} A \\ B \end{matrix} \begin{bmatrix} 1/4 & 3/4 \\ 3/4 & 1/4 \end{bmatrix} \end{matrix}$$

To find:

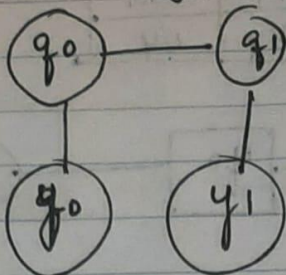
- (i) likelihood of  $p(y)$  using HMM
- (ii) Individual marginals of states where  $p(q_0/y)$  and  $p(q_1/y)$ .

Step 1: 2 state HMM

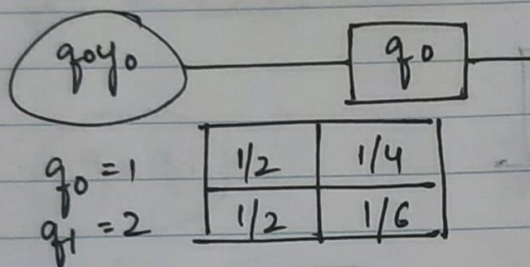
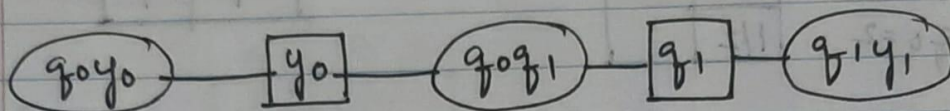




Step 2: Converting to junction tree



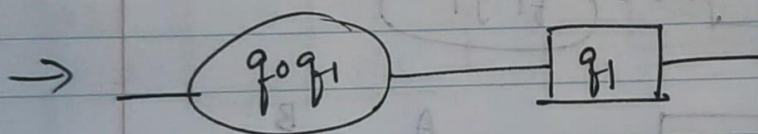
Implementing moralization & triangulation here.



$q_0 = 1$   
 $q_1 = 2$

1/2	1/4
1/2	1/6

- $q_1 y_1$ : emission matrix
- $q_0 q_1$ : transition matrix



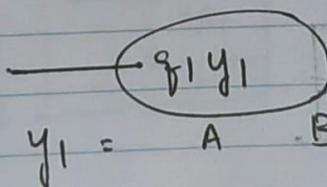
$q_1 =$

1	2
---	---

$q_0 = 1$   
 $q_1 = 2$

1/8	1/8
1/2	1/2

1	1
---	---

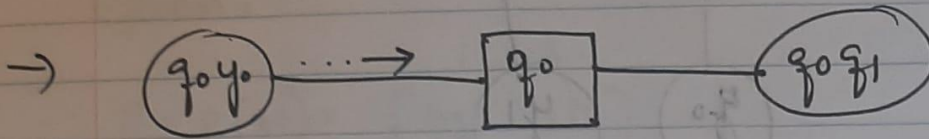


$y_1 =$  A B

$$g_0 = 1$$

$$g_1 = 2$$

1/4	3/4
3/4	1/4



$y_0$     A    B     $g_0$     1...2

$$g_0 = 1$$

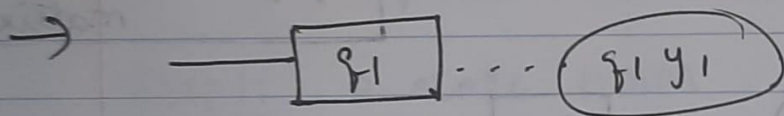
$$g_0 = 2$$

1/12	
1/2	

1	1
---	---

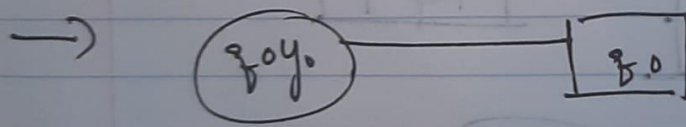
$$g_1 = 1 \dots 2$$

1/8	1/8
1/2	1/2



1	1
---	---

A	B
$g_1 = 1$	3/4
$g_1 = 2$	1/4



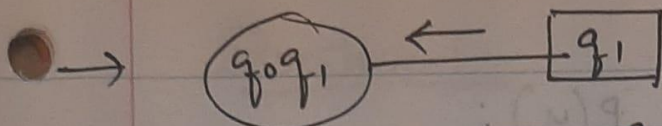
$$g_0 = 1$$

$$g_0 = 2$$

1/2	
1/2	

$$g_0 = 1 \dots 2$$

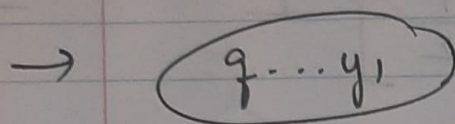
1	1
---	---



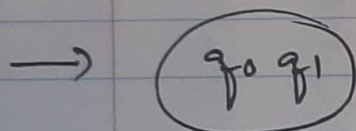
1	2
$1/8$	$7/8$
$1/2$	$1/2$

$g_1 = 1 \dots 2$

1	1
---	---



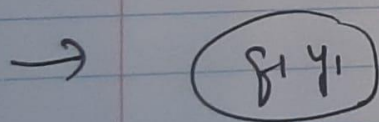
	A	B
$g_1 = 1$	$\diagup \diagdown$	$3/4$
$g_1 = 2$	$\diagup \diagdown$	$1/4$



	1	2
$g_0 = 1$	$1/8$	$7/8$
$g_0 = 2$	$1/2$	$1/2$

$g_1 = 1$

1	2
$3/4$	$1/4$



	A	B
$g_1 = 1$	.	$3/4$
$g_1 = 2$	.	$1/4$



∴ likelihood  $P(y)$ :

$$\begin{bmatrix} \frac{1}{12} \times \frac{1}{8} & , & \frac{1}{12} \times \frac{7}{8} \\ \frac{1}{2} \times \frac{1}{2} & , & \frac{1}{2} \times \frac{1}{2} \end{bmatrix} = \begin{bmatrix} 1/128 & 7/384 \\ 3/16 & 1/16 \end{bmatrix}$$

$$\therefore P(y) = \left[ \frac{1}{128} + \frac{7}{384} + \frac{3}{16} + \frac{1}{16} \right]$$

$$= \frac{53}{192}$$

Question-4

To solve K-means clustering algorithm.

Given that,

$$C_1 = (-4, -5), \quad C_2 = (5, 4)$$

Given 2-D dataset,

A(-3, -1) B(-1, -3) C(-2, -6) D(-5, -7) E(3, 1)  
F(2, 3) G(3, 6) H(8, 1)

Step 1: Initializing value of centroids

$$\text{i.e.; } C_1 = (-4, -5), \quad C_2 = (5, 4)$$

Step 2: objects - centroid distance.

In this step, we calculate the Euclidean distance between <sup>each object of</sup> the 2D dataset and the cluster centroid.

Iteration 0: When using  $C_1: (-4, -5)$

$$A = \sqrt{(-3+4)^2 + (-1+5)^2} = 4.12$$

$$B = \sqrt{(-1+4)^2 + (-3+5)^2} = 3.60$$

$$C = \sqrt{(-2+4)^2 + (-6+5)^2} = 2.23$$

$$D = \sqrt{(-5+4)^2 + (-7+5)^2} = 2.23$$

$$E = \sqrt{(3+4)^2 + (1+5)^2} = 9.21$$

$$F = \sqrt{(2+4)^2 + (3+5)^2} = 10.4$$

$$G = \sqrt{(3+4)^2 + (6+5)^2} = 13.03$$

$$H = \sqrt{(8+4)^2 + (1+5)^2} = 13.41$$

When using  $C_2 = (5, 4)$

$$A = \sqrt{(-3-5)^2 + (-1-4)^2} = 9.43$$

$$B = \sqrt{(-1-5)^2 + (-3-4)^2} = 9.21$$

$$C = \sqrt{(-2-5)^2 + (-6-4)^2} = 12.20$$

$$D = \sqrt{(-5-5)^2 + (-7-5)^2} = 15.62$$

$$E = \sqrt{(3-5)^2 + (1-4)^2} = 3.60$$



$$F = \sqrt{(2-5)^2 + (3-4)^2} = 3.16$$

$$G = \sqrt{(3-5)^2 + (6-4)^2} = 2.82$$

$$H = \sqrt{(8-5)^2 + (1-4)^2} = 4.24$$

Distance matrix can be given as,

$$D_0 = \begin{matrix} & \begin{matrix} A & B & C & D & E & F & G & H \end{matrix} \\ \begin{matrix} C_1 \\ C_2 \end{matrix} & \begin{bmatrix} 4.12 & 3.60 & 2.23 & 2.23 & 9.21 & 10 & 13.03 & 13.41 \\ 9.43 & 9.21 & 12.20 & 15.62 & 3.60 & 3.16 & 2.82 & 4.24 \end{bmatrix} \end{matrix}$$

$C_1$  is considered as group 1

$C_2$  is considered as group 2

Step 3: Object clustering:

We have to assign each object basing upon the minimum distance.

A, B, C, D belong to group 1

E, F, G, H belong to group 2

Then,

$$G^0 = \begin{matrix} & \begin{matrix} A & B & C & D & E & F & G & H \end{matrix} \\ \begin{matrix} \uparrow \\ \downarrow \end{matrix} & \begin{bmatrix} 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix} \end{matrix}$$

group 1  
group 2

Step 4: updating centroids

$$C_1 = \left( \frac{-3-1-2-5}{4}, \frac{-1-3-6-7}{4} \right) = \left( \frac{-11}{4}, \frac{-17}{4} \right) \\ = (-2.75, -4.25)$$

$$C_2 = \left( \frac{3+2+3+8}{4}, \frac{1+3+6+1}{4} \right) = \left( 4, \frac{11}{4} \right) \\ = (4, 2.75)$$

Iteration 1: using  $C_1$

$$A = \sqrt{(-3+2.75)^2 + (-1+4.25)^2} = 3.25$$

$$B = \sqrt{(-1+2.75)^2 + (-3+4.25)^2} = 2.15$$

$$C = \sqrt{(-2+2.75)^2 + (-6+4.25)^2} = 1.90$$

$$D = \sqrt{(-5+2.75)^2 + (-7+4.25)^2} = 3.55$$

$$E = \sqrt{(3+2.75)^2 + (1+4.25)^2} = 7.78$$

$$F = \sqrt{(2+2.75)^2 + (3+4.25)^2} = 8.66$$

$$G = \sqrt{(3+2.75)^2 + (6+4.25)^2} = 11.75$$

$$H = \sqrt{(5+2.75)^2 + (4+4.25)^2} = 11.31$$

When using  $G_2 = (4, 2.75)$

$$A = \sqrt{(-3-4)^2 + (-1-2.75)^2} = 7.94$$

$$B = \sqrt{(-1-4)^2 + (-3-2.75)^2} = 7.61$$

$$C = \sqrt{(-2-4)^2 + (-6-2.75)^2} = 10.60$$

$$D = \sqrt{(-5-4)^2 + (-7-2.75)^2} = 13.26$$

$$E = \sqrt{(3-4)^2 + (1-2.75)^2} = 2.01$$

$$F = \sqrt{(2-4)^2 + (3-2.75)^2} = 2.01$$

$$G = \sqrt{(3-4)^2 + (6-2.75)^2} = 3.40$$

$$H = \sqrt{(5-4)^2 + (4-2.75)^2} = 1.60$$

Distance matrix  $\begin{matrix} \nearrow \text{group 1} & & \nwarrow \text{group 2} \end{matrix}$

	A	B	C	D	E	F	G	H
A	3.25	2.15	1.90	3.55	7.78	8.66	11.75	11.31
B	7.94	7.61	10.60	13.26	2.01	2.01	3.40	1.60

→ updated centroid:

$$G_1 = \left( \frac{3+2+3+8}{4}, \frac{1+3+6+7}{4} \right) = \left( \frac{-11}{4}, \frac{-17}{4} \right)$$

$$G_2 = \left( \frac{3+2+3+8}{4}, \frac{1+3+6+7}{4} \right) = (4, 2.75)$$



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### Question - 5

Jensen's inequality:

To prove: for non-negative real numbers  $x_1, x_2, \dots, x_n$

$$\frac{x_1 + x_2 + \dots + x_n}{n} \geq \sqrt[n]{x_1 \cdot x_2 \cdot \dots \cdot x_n}$$

This is of the form that Arithmetic mean greater than the Geometric mean.

As 'log' function is an increasing function, we apply 'log' on both sides to proceed with Jensen's inequality.

$$\text{i.e.; } \log \left( \frac{x_1 + x_2 + \dots + x_n}{n} \right) \geq \frac{1}{n} \log(x_1 \cdot x_2 \cdot \dots \cdot x_n)$$

$$= \frac{\log x_1 + \dots + \log x_n}{n}$$

$$\text{i.e.; } \log \left( \frac{1}{n} \sum_{i=1}^n x_i \right) \geq \frac{1}{n} \log(x_1 \cdot \dots \cdot x_n)$$

$$\Rightarrow \log \left( \frac{1}{n} \sum_{i=1}^n x_i \right) \geq \frac{1}{n} \log \left( \prod_{i=1}^n x_i \right)$$

$$(or) \log \left( \frac{1}{n} \sum_{i=1}^n x_i \right) \geq \log \left( \prod_{i=1}^n x_i \right)^{1/n}$$

Then,

$$\frac{\sum_{i=1}^n x_i}{n} \geq \sqrt[n]{\prod_{i=1}^n x_i}$$

$$\therefore \frac{x_1 + x_2 + \dots + x_n}{n} \geq \sqrt[n]{x_1 \cdot x_2 \cdot \dots \cdot x_n}$$

When we use  $E[x]$  as mean of  $x_i$ 's.

Then,

$$\boxed{\log(E[x]) \geq E[\log(x)]}$$

Hence, it is proved using concavity of log given by Jensen's inequality.

Question.

6.)

Convolution layer:

$$8 \rightarrow 10, \quad 2 \times 2, \quad 2 \times 2$$

$$\Rightarrow \left( \frac{x}{2}, \frac{y-1}{2} \right)$$

$$\left( \frac{x}{2}, \frac{y-1}{2} \right)$$

is feature  
map here.

Max Pooling:

$$3 \times 3, \quad 2 \times 2.$$

$$\left( \frac{x-6+4}{4} \right) \left( \frac{y-3}{4} \right) = 3(101-2)+2$$

$$= 302 \times 154$$

Flattening:  $10 \times 12 \times 9 \rightarrow 1080$

$$\left( \frac{x-2}{4} \right)$$

$$\left[ \because \frac{x-2}{4} = 12 \right]$$

$$x-2 = 48$$

$$x = 50$$

$$\frac{y-3}{4} = 9$$

$$\Rightarrow y = 154$$

ReLU: No effect