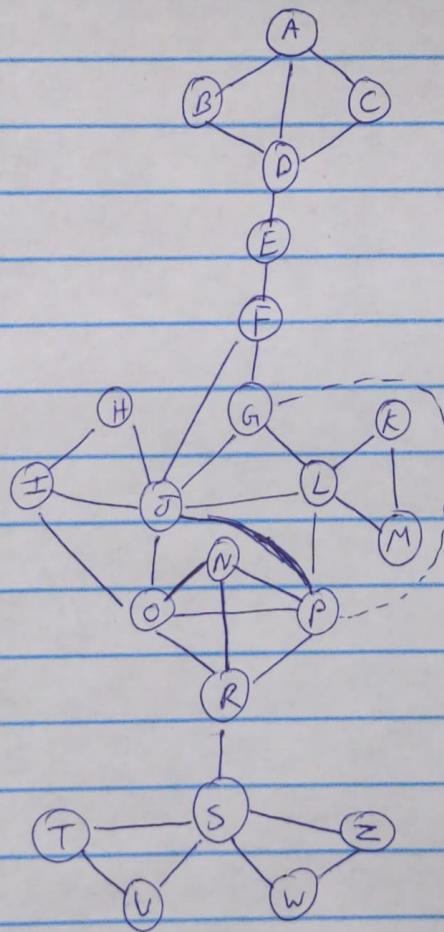


Problem 1

To convert the given graph into a junction tree we perform the following steps:-

Moralization

- Convert directed graph to undirected graph
- Connect nodes that have a common children.



Here, we introduced
~~few~~ new edges

→ J - P

→ O - N

→ A - D

→ F - J

→ I - O

→ G - P

→ N - R

After this we perform triangulation to eliminate any loop with more than or equal to 4 nodes.

This gives us the following graph.

Since there are no such loops in above graph, we can directly find out the cliques and apply kruskal's algorithm.

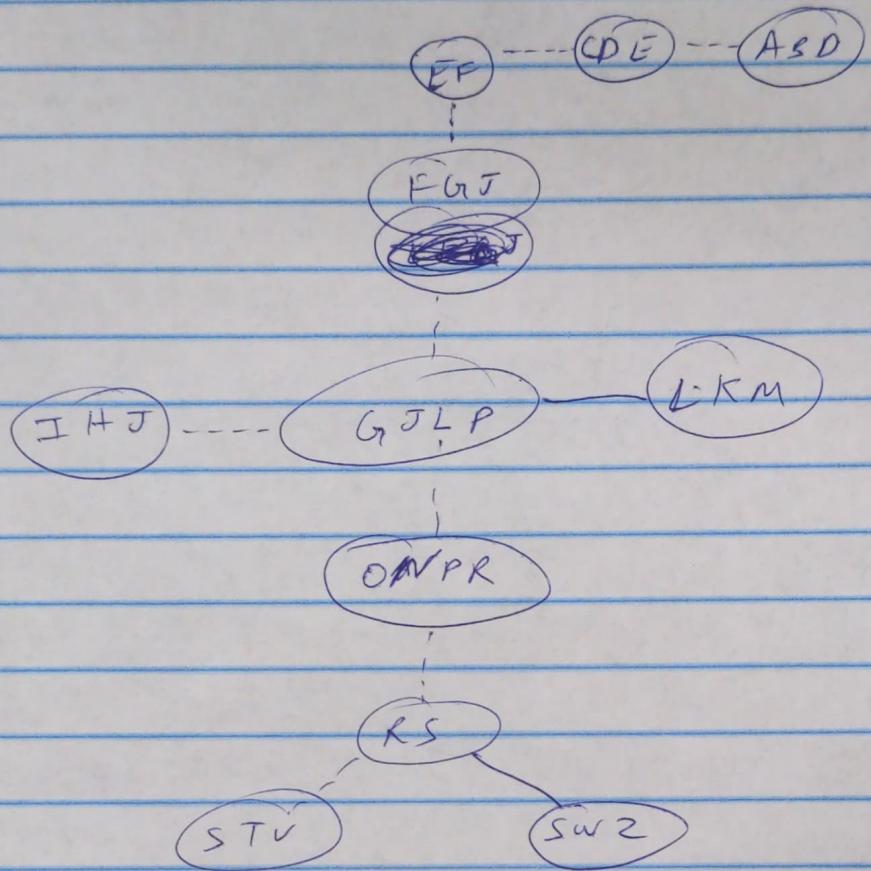
Cliques:

$\rightarrow ABD \rightarrow ACD \rightarrow DE \rightarrow EF \rightarrow FGJ$
 $\rightarrow GJLP \rightarrow LKM \rightarrow HIJ \rightarrow JPO \rightarrow ONPR$
 $\rightarrow JIO \rightarrow RS \rightarrow STV \rightarrow SWZ$

Using this we build Kruskal Table as follows:

	DE	EF	RS	ABD	ACD	FGJ	LKM	HIJ	JPO	JIO	STV	SWZ	GJLP	ONPR
DE	-	1	0											
EF		-	0											
RS			-											
ABD				-										
ACD					-									
FGJ						-								
LKM							-							
HIJ								-						
JPO									-					
JIO										-				
STV											-			
SWZ												-		
GJLP													-	
ONPR														-

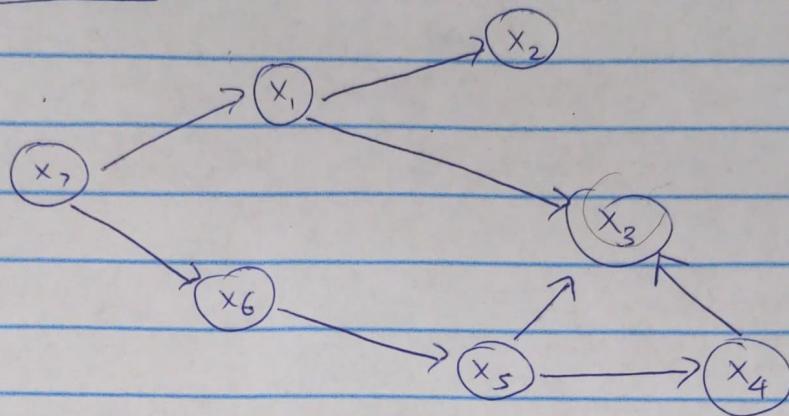
And then using the above we can convert it into JTA as below.



CD

Problem 2

Given Graph :-



$$p(x_1, \dots, x_7) = \prod_{i=1}^7 p(x_i | \text{parents}_i)$$

~~$p(x_7)$~~

$$= p(x_7) p(x_1 | x_7) p(x_2 | x_1) p(x_6 | x_7) p(x_5 | x_6)$$
 ~~$p(x_3 | x_8, x_7)$~~ ~~$p(x_4 | x_7)$~~

$$p(x_3 | x_1, x_4, x_5) p(x_4 | x_5)$$

Rearranging for simplicity we get :-

$$p(x_1, \dots, x_7) = p(x_1 | x_7) p(x_2 | x_1) p(x_3 | x_1, x_4, x_5) p(x_4 | x_5) p(x_5 | x_6) \\ p(x_6 | x_7) p(x_7)$$

Now using bayes ball theorem we find the following :-

(a) x_2 and x_6 are independent : FALSE

$$x_2 - x_1 - x_7 - x_6$$

$\rightarrow x_2 - x_1 - x_7$ Markov chain, Hence ball will go through

$\rightarrow x_1 - x_7 - x_6$ Two effects rule, ball goes through

Hence False x_2 and x_6 are dependent

(b) x_2 and x_6 are independent, given x_1, x_3, x_5 : TRUE
 $\rightarrow x_2 - x_1 - x_3 - x_6$: Ball stops at x_1 , (Markov chain)

$\rightarrow x_2 - x_1 - x_3 - x_5 - x_6$: Ball stops at x_1 ,

(c) x_1 and x_4 are conditionally independent given x_5 : TRUE

$\rightarrow x_1 \rightarrow x_3 \leftarrow x_4$: Two causes, Ball goes through stops
~~x₃ - x₄~~ since x_3 is not given

$\rightarrow x_1 - x_3 - x_5 - x_4$: Same as above

$\rightarrow x_1 - x_7 - x_6 - x_5 - x_4$: Ball stops at $(x_6 - x_5 - x_4)$
since x_5 is given (Markov chain)

(d) $x_5 \perp\!\!\!\perp x_2 | x_1, x_3$: TRUE

$\rightarrow x_5 - x_3 - x_1 - x_2$: Ball stops at x_1 , (Two effects)

$\rightarrow x_5 - x_6 - x_7 - x_1 - x_2$: Ball stops at x_1 , (Markov chain)

(e) $x_5 \perp\!\!\!\perp x_1 | x_3, x_2, x_4$: FALSE

$x_5 - x_6 - x_7 - x_1$: $x_5 - x_6 - x_7$ (Markov chain)

$x_6 - x_7 - x_1$, (Two effects) Ball goes through

(f) $x_4 \perp\!\!\!\perp x_7 | x_6$: TRUE

$x_4 - x_5 - x_6 - x_7$: Ball stops at x_6 (Markov chain)

$x_4 - x_3 - x_1 - x_7$: Ball stops at x_3 (Two causes)

$$(g) x_4 \perp\!\!\!\perp x_7 \mid x_5 : \boxed{\text{TRUE}}$$

$x_4 - x_5 - x_6 - x_7$: Ball stops at x_5 (Markov chain)

$x_4 - x_3 - x_1 - x_7$: Ball stops at x_3 (Two causes)

$$(h) x_1 \overset{\perp\!\!\!\perp}{x_5} \mid x_6, x_7 : \boxed{\text{TRUE}}$$

$x_1 - x_7 - x_6 - x_5$: Ball stops at x_7 (Two effects)

$x_1 - x_3 - x_5$: Ball stops at x_3 (Two causes)

$$(i) x_5 \perp\!\!\!\perp x_1 : \boxed{\text{FALSE}}$$

$x_1 - x_7 - x_6 - x_5$: Ball goes through (Two effects + Markov chain)

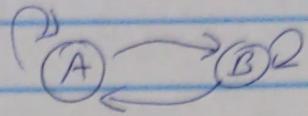
$$(j) x_2 \perp\!\!\!\perp x_4 \mid x_1 : \boxed{\text{TRUE}}$$

$x_2 - x_1 - x_3 - x_4$: Ball stops at x_1 (Two effects)

$x_2 - x_1 - x_7 - x_6 - x_5 - x_4$: Ball stops at x_1 (Markov chain)

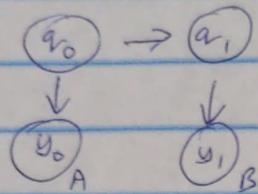
Problem 3

Given finite state machine is :-

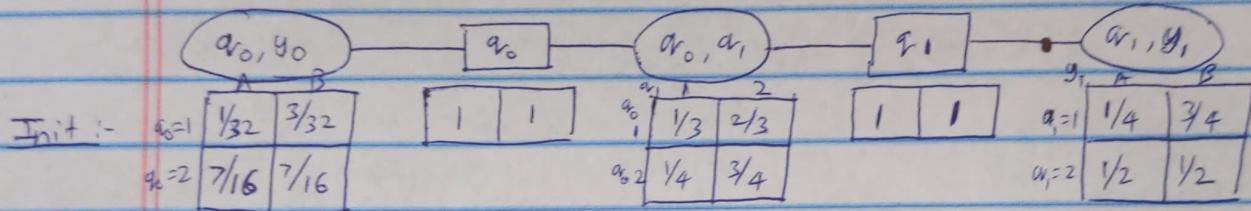


and we construct JTA as :-

for



we initialize the junction tree using the given values as :-



Now we apply the message passing iteration step by step :-

\rightarrow	\leftarrow
$q_0 = A \quad B$ $q_0 = 1 \quad 2$ $1 \quad \frac{1}{32} \quad \frac{3}{32}$ $2 \quad \frac{7}{16} \quad \frac{7}{16}$	$q_0 = A \quad B$ $q_0 = 1 \quad 2$ $1 \quad \frac{1}{3} \quad \frac{2}{3}$ $2 \quad \frac{1}{4} \quad \frac{3}{4}$

$q_0 = A \quad B$	$q_0 = 1 \quad 2$	$q_0 = 1 \quad 2$	$q_1 = A \quad B$
$1 \quad \frac{1}{32} \quad \frac{3}{32}$	$1 \quad \frac{1}{32} \quad \frac{7}{16}$	$1 \quad \frac{1}{3} \quad \frac{2}{3}$	$1 \quad \frac{1}{4} \quad \frac{3}{4}$
$2 \quad \frac{7}{16} \quad \frac{7}{16}$	$2 \quad \frac{1}{4} \quad \frac{3}{4}$	$2 \quad \frac{1}{2} \quad \frac{1}{2}$	$2 \quad \frac{1}{2} \quad \frac{1}{2}$

$q_0 = A \quad B$	$q_0 = 1 \quad 2$	$q_0 = 1 \quad 2$	$q_1 = A \quad B$
$1 \quad \frac{1}{32} \quad \frac{3}{32}$	$1 \quad \frac{1}{32} \quad \frac{7}{16}$	$1 \quad \frac{1}{128} \quad \frac{1}{96}$	$1 \quad \frac{1}{4} \quad \frac{3}{4}$
$2 \quad \frac{7}{16} \quad \frac{7}{16}$	$2 \quad \frac{1}{4} \quad \frac{3}{4}$	$2 \quad \frac{1}{256} \quad \frac{1}{192}$	$2 \quad \frac{1}{2} \quad \frac{1}{2}$

So the likelihood,

$$\begin{aligned} p(y) &= \frac{7}{384} + \frac{63}{256} = \frac{1}{128} + \frac{1}{96} + \frac{21}{256} + \frac{21}{128} \\ &= \frac{23}{256} + \frac{67}{384} \\ &= \frac{203}{768} = [0.264] \end{aligned}$$

$$p(a_0=1|y) = \frac{7/384}{7/384 + 63/256} = \frac{2}{29}$$

$$p(a_0=2|y) = \frac{63/256}{7/384 + 63/256} = \frac{27}{29}$$

$$p(a_1=1|y) = \frac{23/256}{23/256 + 67/384} = \frac{69}{203}$$

$$p(a_1=2|y) = \frac{67/384}{23/256 + 67/384} = \frac{134}{203}$$

Problem 4

$$\text{Initial centers} = \mu_1 = (-4, 2) \quad \mu_2 = (4, -5)$$

$$\text{Data-points} = \begin{matrix} \textcircled{6} \\ (-5, 3) \end{matrix}, \begin{matrix} \textcircled{1} \\ (-3, 2) \end{matrix}, \begin{matrix} \textcircled{2} \\ (-4, 5) \end{matrix}, \begin{matrix} \textcircled{3} \\ (-3, 4) \end{matrix}, \begin{matrix} \textcircled{4} \\ (3, -4) \end{matrix}, \begin{matrix} \textcircled{5} \\ (4, -2) \end{matrix}, \\ \begin{matrix} \textcircled{6} \\ (6, -6) \end{matrix}, \begin{matrix} \textcircled{7} \\ (3, -3) \end{matrix}$$

So lets say we have points $x_0 \dots x_7$ as above
we initially calculate distance of each point to the
clusters centers as.

Points	$\mu_1(-4, 2)$	$\mu_2(4, -5)$	$\min(\mu_1(-4, 2), \mu_2(4, -5))$ which clusters?
-5, 3	1.414	12.04	Cluster 1
-3, 2	1	9.89	Cluster 1
-4, 5	3	12.80	Cluster 1
-3, 4	2.23	11.40	Cluster 1
3, -4	9.21	1.4	Cluster 2
4, -2	8.94	3	Cluster 2
6, -6	12.80	2.2	Cluster 2
3, -3	13	4.4	Cluster 2

Now based on above we calculate new centers:-

$$\mu_1 = \left(\frac{-5 + -3 + -4 + -3}{4}, \frac{3 + 2 + 5 + 4}{4} \right) = \left(\frac{-15}{4}, \frac{15}{4} \right)$$

$$\mu_2 = \left(\frac{3 + 4 + 6 + 2}{4}, \frac{-4 + -2 + -6 + -3}{4} \right) = \left(\frac{21}{4}, \frac{-15}{4} \right)$$

P.T.O \rightarrow

With these new centers we do ~~the~~ the same as above and calculate distances.

Points	$M_1(-\frac{15}{4}, \frac{15}{4})$	$M_2(\frac{21}{4}, -\frac{15}{4})$	Which cluster?
-5, 3	1.45	12.27	1
-3, 2	1.90	10.05	1
-4, 5	1.27	12.73	1
-3, 4	0.79	11.31	1
3, -4	10.27	2.26	2
4, -2	9.65	2.15	2
6, -6	13.78	2.37	2
8, -3	13.55	2.85	2

New cluster centers:- They are same as before since cluster is same

$$M_1 = \left(-\frac{15}{4}, \frac{15}{4} \right) \quad M_2 = \left(\frac{21}{4}, -\frac{15}{4} \right)$$

Cluster 1 : $\left[(-5, 3), (-3, 2), (-4, 5), (-3, 4) \right]$

Cluster 2 : $\left[(3, -4), (4, -2), (6, -6), (8, -3) \right]$

Problem 5

$$\text{KL Divergence} := \text{KL}(p \parallel q) = \sum_{x \in \mathcal{X}} p(x) \log \frac{p(x)}{q(x)}$$

where both $p(x)$ and $q(x)$ are probability distribution
we have to use Jensen's inequality to prove ~~that~~ that
given KL Divergence is non-negative.

Jensen's inequality for distribution is as follows:-

$$f(E[x]) \leq E[f(x)] \quad \text{and for}$$

continuous densities it is :-

$$f\left(\int p(x) dx\right) \leq \int f(x)p(x)dx$$

Now, applying ~~the~~ above jensen's inequality to the given
KL - divergence we get :-

$$\text{KL}(p \parallel q) = \int_{-\infty}^{\infty} p(x) \log \frac{p(x)}{q(x)} dx$$

$$= - \int_{-\infty}^{\infty} p(x) \log \frac{q(x)}{p(x)} dx$$

$$\geq - \log \int_{-\infty}^{\infty} \frac{q(x)}{p(x)} p(x) dx \quad \text{--- Applying Jensen's inequality}$$

$$\geq 0 \quad \dots \text{since } [\log 1 = 0]$$

Hence Non-Negative

We first flip the log to cancel out $p(x)$ when jensen's
inequality is applied

Problem 6

Given the CNN architecture we have to find size of input image (x, y)

so lets assume size is x, y and calculate the output size (which is already known)

Conv layer 1 :- Input size = (x, y)

output channels = 5 , filter size = $(3, 3)$

stride = $(2, 3)$

$$\text{Output} = \left[\frac{x-3}{2} + 1, \frac{y-3}{3} + 1 \right]$$

Relu has no effect on size.

Maxpool layer 1 :- Region size = $(2, 2)$, stride = $(2, 2)$

Output size = $\frac{1}{2}$

$$= \left[\frac{x-3}{4} + \frac{1}{2}, \frac{y-3}{6} + \frac{1}{2} \right]$$

Now for next lets assume input as x, y again P-T-O \rightarrow

Conv layer 2 :- Input size = x, y

Output channels = \rightarrow Input channels = 5

filter = 3×3 stride = 2×2

$$\text{Output size} = \left[\frac{x-2}{4} + \frac{1}{2}, \frac{y-9}{6} + \frac{1}{2} \right]$$

After Maxpool with region = 2×2 and stride = 3×3

we have output size as

$$= \left[\frac{x-7}{12} + \frac{1}{3}, \frac{y-9}{18} + \frac{1}{3} \right]$$

Conv layer, 2 Input size = $\left[\frac{x-1}{4}, \frac{y}{6} \right]$

~~filter~~ filter = 3×3 stride = 2×2

$$\text{output size} = \left[\frac{\frac{x-1}{4} - 3}{2} + 1, \frac{\frac{y}{6} - 3}{2} + 1 \right]$$
$$= \left[\frac{x-13}{8} + 1, \frac{y-18}{12} + 1 \right]$$

Maxpool 1 : stride = 3×3 region : 2×2

$$\text{output size} = \left[\frac{x-13}{24} + \frac{1}{3}, \frac{y-18}{36} + \frac{1}{3} \right]$$

Now actual output size given is $(?, 14)$

$$\therefore \frac{x-13}{24} + \frac{1}{3} = ?$$

$$\therefore x = \left(? - \frac{1}{3} \right) \times 24 + 13$$

$$= 173$$

$$\therefore \frac{y-18}{36} + \frac{1}{3} = 14$$

$$\therefore y = \left(14 - \frac{1}{3} \right) \times 36 + 18$$

$$\therefore y = 510$$

$$\text{Image size} = (173, 510)$$