

Q2. In a big company, 60% of the employees are female, and the rest are male. Among all the female employees, 70% are married. Among all the male employees, 40% are married.

Question: Among employees who are married, how many percentage of them are female?

Employees \rightarrow $\begin{cases} 60\% \text{ Female} \\ 40\% \text{ Male} \end{cases}$

60% Female \rightarrow $\begin{cases} 70\% \text{ Married} \\ 30\% \text{ Not married} \end{cases}$

40% Male \rightarrow $\begin{cases} 40\% \text{ Married} \\ 60\% \text{ Not married} \end{cases}$

$$60\% = 100\%$$

$$x \quad 70\%$$

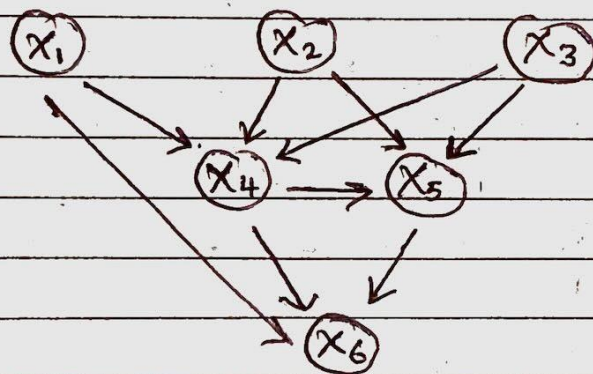
$$x = \frac{60 \times 70}{100} = 42\%$$

$$40\% = 100\%$$

$$x \quad 40\%$$

$$x = \frac{40 \times 40}{100} = 16\%$$

Q3. In the Bayesian network below, which random variables does X_4 directly depend on?



Ans: X_1 , X_2 and X_3

Q4. How many parameters are needed to specify the Bayesian network above.

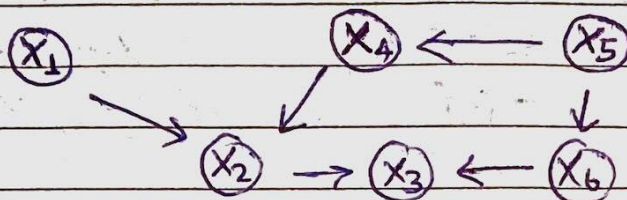
$2^0 \times 3$ X_1, X_2, X_3 do not depend on any other variable.

$2^3 \times 3$ X_4, X_5, X_6 depends on 3 other variable.

$$\therefore 2^0 \times 3 + 2^3 \times 3 = 1 \times 3 + 8 \times 3 = 3 + 24 = 27$$

Q5. At the final session of the lecture slides, we have presented a concrete example about ~~random~~ COVID, which is a Bayesian network with 6 random variables, and the network has 14 parameters.

For this Bayesian network, compute $P[X_3=1 | X_1 \neq X_4]$ to 5 d.p.



Step ① Total number of parameters needed

$2^0 \times 2$ X_1, X_5 do not depend on any other variable.

$2^1 \times 2$ X_4, X_6 depends on 1 other variable.

$2^2 \times 2$ X_2, X_3 depends on 2 other variables.

$$\therefore 2^0 \times 2 + 2^1 \times 2 + 2^2 \times 2 = 2 + 4 + 8 = 14$$

Step ②

$$P[X_1, X_2, X_3, X_4, X_5, X_6] = P[X_1] \cdot P[X_5] \cdot P[X_4 | X_5] \cdot P[X_6 | X_5] \\ \cdot P[X_2 | X_1, X_4] \cdot P[X_3 | X_2, X_6]$$

Note: $P[X_4 | X_5]$ means X_4 depends on X_5

Step ③ the Bayes formula. $P(A|E) = \frac{P(A \cap E)}{P(E)}$

$$P[X_3=1 | X_1 \neq X_4] = \frac{P[X_3=1, X_1 \neq X_4]}{P[X_1 \neq X_4]}$$

~~$P[X_3, X_4]$~~

$$\begin{aligned} P[X_3=1, X_1 \neq X_4] &= P[X_1=0, X_2=0, X_3=1, X_4=1, X_5=0, X_6=0] \\ &+ P[X_1=0, X_2=0, X_3=1, X_4=1, X_5=0, X_6=1] \\ &+ P[X_1=0, X_2=0, X_3=1, X_4=1, X_5=1, X_6=0] \\ &+ P[X_1=0, X_2=0, X_3=1, X_4=1, X_5=1, X_6=1] \\ &+ P[X_1=0, X_2=1, X_3=1, X_4=1, X_5=0, X_6=0] \\ &+ P[X_1=0, X_2=1, X_3=1, X_4=1, X_5=0, X_6=1] \\ &+ P[X_1=0, X_2=1, X_3=1, X_4=1, X_5=1, X_6=0] \\ &+ P[X_1=0, X_2=1, X_3=1, X_4=1, X_5=1, X_6=1] \end{aligned}$$

$$P[X_1, X_2, X_3, X_4, X_5, X_6] = P[X_1] \cdot P[X_5]$$

$$\cdot P[X_4|X_5] \cdot P[X_6|X_5]$$

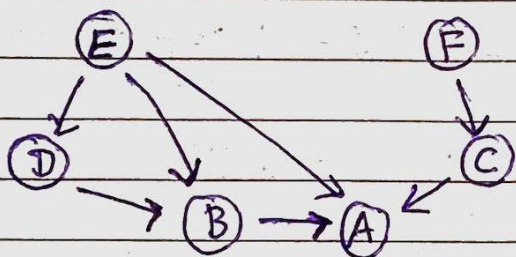
$$\cdot P[X_2|X_1, X_4] \cdot P[X_3|X_2, X_6]$$

Q4. b) Suppose there are 6 binary random variables:
A, B, C, D, E and F.

The conditional independence among these variables enables the joint probability distribution $P(A, B, C, D, E, F)$ to be defined as

$$P(A, B, C, D, E, F) = P(A|B, C, E) P(B|D, E) P(C|F) P(D|E) P(E) P(F)$$

i) Draw the Bayes network that corresponds to this conditional independence. Is it a polytree? Justify your answer.



A polytree (directed tree, single connected network) — is a directed acyclic graph whose underlying undirected graph is a tree.

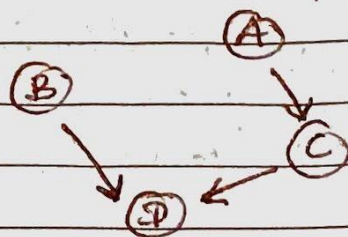
No, ~~it~~ it is not a polytree because there is a cycle between the variables.

ii) Specify the minimum number of parameters required to fully define the joint probability distribution.

$2^0 \times 2$	E, F	Do not depend on other variable
$2^1 \times 2$	D, C	Do depend on 1 other variable
$2^2 \times 1$	B	Does depend on 2 other variable
$2^3 \times 1$	A	Does depend on 3 other variable

$$\therefore 2^0 \times 2 + 2^1 \times 2 + 2^2 \times 1 + 2^3 \times 1 = 2 + 4 + 4 + 8 = 18$$

- c) Consider the following Bayes network where each random variable can take two possible values $\{T, F\}$:



The associated probability distributions for the binary random variables A, B, C and D are

$$P(A=T)=0.1, \quad P(B=T)=0.8, \quad \text{and} \\ P(C=T|A=T)=0.6 \quad P(C=T|A=F)=0.3$$

$$P(D=T|C=T, B=T)=0.1 \quad P(D=T|C=F, B=T)=0.5 \\ P(D=T|C=T, B=F)=0.3 \quad P(D=T|C=F, B=F)=0.8$$

- E) Write down an expression for full joint distribution of the random variables A, B, C and D.
Compute the probability that A and C are T while B and D are F.

$$P(A, B, C, D) = P(D|B, C) P(C|A) P(A) P(B)$$

$$P(A \neq C = T | B \neq D = F) =$$

- ii) Explain why the representation of the joint probability distribution of A, B, C, and D using the Bayes network is preferable to a direct tabular representation of the full joint probability distribution.