

Assignment 3

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Question 1

Consider the following Bayesian network, where variables A through E are all Boolean valued. Note: there is a typo in the image, it should be $P(A = \text{true}) = 0.2$ instead of $P(D = \text{true}) = 0.2$.

A)

$$P(A) * P(B) * P(C) * P(D | A, B) * P(E | B, C) = .2 * .5 * .8 * .1 * .3 = 0.0024$$

B)

$$\begin{aligned} P(\neg A, \neg B, \neg C, \neg D, \neg E) &= \neg P(A) * \neg P(B) * \neg P(C) * \neg P(D | A, B) * \neg P(E | B, C) \\ &= 0.8 * 0.5 * 0.2 * 0.1 * 0.8 = 0.0064 \end{aligned}$$

C)

$$\begin{aligned} P(\neg A | B, C, D, E) &= \frac{P(\neg A) * P(B) * P(C) * P(D | \neg A, B) * P(E | B, C)}{P(A, B, C, D, E) + P(\neg A) * P(B) * P(C) * P(D | \neg A, B) * P(E | B, C)} \\ &= \frac{0.8 * 0.5 * 0.8 * 0.6 * 0.3}{0.0024 + 0.8 * 0.5 * 0.8 * 0.6 * 0.3} = \frac{0.0576}{0.06} \\ &= 0.96 \end{aligned}$$

Question 2

A)

$$\begin{aligned} &P(B), P(E), P(A | B, E), P(J | A), P(M | A) \\ &\sum_B P(B) * \sum_E P(E) * \sum_A P(A | B, E) * P(J | A) * P(M | A) \\ &F1(B) * \sum_E F2(E) * \sum_A F3(A, B, E) * F4(A) * F5(A) \\ &F1(B) = P(B) = \begin{pmatrix} P(b) \\ P(\neg b) \end{pmatrix} = \begin{pmatrix} 0.9 \\ 0.05 \end{pmatrix} \\ &F2(E) = P(E) = \begin{pmatrix} P(e) \\ P(\neg e) \end{pmatrix} = \begin{pmatrix} 0.002 \\ 0.998 \end{pmatrix} \\ &F4(A) = P(J | A) = \begin{pmatrix} P(j|a) \\ P(j|\neg a) \end{pmatrix} = \begin{pmatrix} 0.90 \\ 0.05 \end{pmatrix} \\ &F5(A) = P(M | A) = \begin{pmatrix} P(m|a) \\ P(m|\neg a) \end{pmatrix} = \begin{pmatrix} 0.70 \\ 0.01 \end{pmatrix} \\ &F3(A, B, E) = P(A | B, E) = \begin{pmatrix} P(a|b,e) & P(a|b,\neg e) \\ P(a|\neg b,e) & P(a|\neg b,\neg e) \end{pmatrix} = \begin{pmatrix} 0.95 & 0.94 \\ 0.29 & 0.001 \end{pmatrix} \\ &F3(\neg A, B, E) = P(\neg A | B, E) = \begin{pmatrix} P(\neg a|b,e) & P(\neg a|b,\neg e) \\ P(\neg a|\neg b,e) & P(\neg a|\neg b,\neg e) \end{pmatrix} = \begin{pmatrix} 0.05 & 0.06 \\ 0.71 & 0.999 \end{pmatrix} \\ &F6(B, E) = \sum_A P(A | B, E) * P(J | A) * P(M | A) \\ &F6(B, E) = F3(A, B, E) * F4(A) * F5(A) \\ &F6(B, E) = F3(A, B, E) * F4(A) * F5(A) + F3(\neg A, B, E) * F4(A) * F5(A) \\ &F7(B) = F2(B) * F6(B, E) + F2(E) * F6(B, E) \\ &P(B | J, M) = \propto F1(B) * F7(B) \\ &F1(B) * F7(B) = \begin{pmatrix} 0.00059224 \\ 0.001491857 \end{pmatrix} \\ &\propto = 1 / (0.00059224 + 0.001491857) = 479.8233 \\ &F1(B) * F7(B) (\text{Normalizing with } \propto) = \begin{pmatrix} 0.2841718 \\ 0.7158281 \end{pmatrix} \end{aligned}$$

B)

Enumeration = $O(n^{2^{n-2}})$
Variable Elimination = $O(n)$

Question 3

1)

First we find the probability the rover is in a hot location.

Table 1: Probability of Rover, Hot

X_1	$\mathbf{P}(X_1 \parallel E_1 = \mathbf{Hot})$
A	1.0
B	0.0
C D E F	0.0

Then we find where $P(X_2 \parallel E_1 = \text{Hot}, E_2 = \text{Cold})$

Table 2: Probability of Rover, Hot Cold

X_2	$\mathbf{P}(X_2 \parallel E_1 = \mathbf{Hot}, E_2 = \mathbf{Cold})$
A	0.0
B	1.0
C D E F	0.0

Then we find the X_3 value and normalize it.

Table 3: Probability of Rover, Hot Cold Cold

X_3	$\mathbf{P}(X_3 \parallel E_1 = \mathbf{Hot}, E_2 = \mathbf{Cold}, E_3 = \mathbf{Cold})$
A	0.0
B	0.2
C	0.8
D E F	0.0

2)

The probability of position at the end of day 2 has to be B because if day 1 is hot, it must stay at A, and if day 2 is cold and it moves a mile a day, it must be at B.

Table 4: Probability of Position at End of Day 2

X_2	$\mathbf{P}(X_2 \parallel Hot_1, Cold_1, Cold_2)$
A	0.0
B	1.0
C D E F	0.0

3)

The most likely sequence of positions would be day 1 is A, day 2 is B, day 3 is B.

4)

Because we already know hot_1 , $cold_2$, and $cold_3$, we know that we have to be at C so we could have hot_4 . The probability of being hot_4 would be $0.64 * 0.8$, as we would have to be at C (0.64) to move to D and there is a 0.8 chance of moving to D instead of staying at C, so 0.8 times 0.64 is equal to 0.512.

Next, we have hot_5 . We have a 0.2 chance of staying at D instead of moving to E, so it is 0.2 times 0.512, or 0.1024.

Finally, to move to $cold_6$, we have a 0.8 chance. So 0.8 times 0.1024 would be 0.08192, or a 8.192% chance of this sequence.

5)

We start with the following probability distribution.

Table 5: Probability of Rover, Hot Cold Cold	
X_3	$P(X_3 \parallel E_1 = \text{Hot}, E_2 = \text{Cold}, E_3 = \text{Cold})$
A	0.0
B	0.2
C	0.8
D E F	0.0

Then we have the possibility of either moving to D or staying at C. If we're still at B on day 3, then we could either stay at B or move to C. If we're at C, then we could either stay at C or move to D. So we find the probability of location on day 4 as:

Table 6: Probability of Rover, Day 4 Unknown	
X_4	$P(X_3 \parallel E_1 = \text{Hot}, E_2 = \text{Cold}, E_3 = \text{Cold}, E_4 = \text{Unknown})$
A	0.0
B	0.04
C	0.32
D	0.64
E F	0.0

Moving on to day 5, we have a similar situation. We could stay at any of the locations or move forward.

Table 7: Probability of Rover, Day 5 Unknown	
X_5	$P(X_3 \parallel E_1 = \text{Hot}, E_2 = \text{Cold}, E_3 = \text{Cold}, E_4, E_5 = \text{Unknown})$
A	0.0
B	0.008
C	0.096
D	0.384
E	0.512
F	0.0

Question 4

a

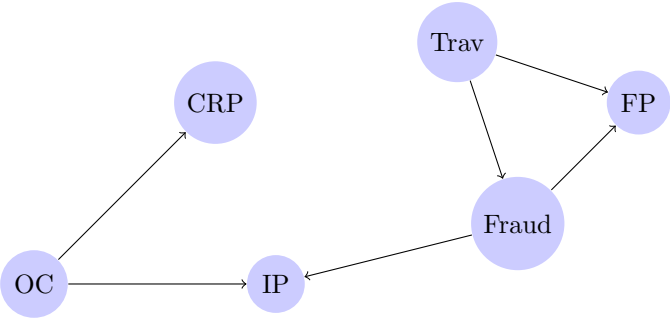


Table 8: OC

True	False
0.75	0.25

Table 9: CRP

OC	True	False
True	0.10	0.90
False	0.001	0.999

Table 10: Trav

True	False
0.05	0.95

Table 11: Fraud

Trav	True	False
True	0.01	0.99
False	0.004	0.996

Table 12: FP

Trav	Fraud	True	False
True	True	0.90	0.10
True	False	0.10	0.90
False	True	0.90	0.10
False	False	0.01	0.99

b

$P(\text{Fraud} = \text{true} \parallel \text{CRP} = \text{false}, \text{IP} = \text{false}, \text{FP} = \text{false}, \text{Trav} = \text{false})$ OR
 $P(\text{Fraud} = \text{true} \parallel \text{CRP} = \text{false}, \text{IP} = \text{false}, \text{FP} = \text{false}, \text{Trav} = \text{true})$
 $= 0.01 + 0.004 = 0.0104$ or 1.04%

Table 13: IP			
OC	Fraud	True	False
True	True	0.02	0.98
True	False	0.01	0.99
False	True	0.011	0.989
False	False	0.001	0.999

$P(\text{Fraud} = \text{true} \parallel \text{FP} = \text{true}, \text{IP} = \text{false}, \text{CRP} = \text{true})$
 $= 0.9 * (0.10 * 0.02) = 0.0018$ or 0.18%