

Exercise 1: TDT4171 Artificial Intelligence Methods

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Problem 1

a)

Let the random variable S denote the number of siblings a child has. Then

$$P(S \leq 2) = 0.15 + 0.49 + 0.27 = \underline{0.91} \quad (1)$$

b)

Using the same notation, we get the probability

$$P(S > 2 | S \geq 1) = \frac{P(S > 2, S \geq 1)}{P(S \geq 1)} = \frac{P(S > 2)}{P(S \geq 1)} = \frac{0.09}{0.85} = \underline{0.106}. \quad (2)$$

c)

Let S_1 , S_2 and S_3 denote each friend's number of siblings. We assume that these random variables are independent. We get:

$$P(S_1 + S_2 + S_3 = 3) = 0.49^3 + 0.15 \cdot 0.49 \cdot 0.27 \cdot 3! + 0.15^2 \cdot 0.06 \cdot 3 = \underline{0.241}. \quad (3)$$

The first term corresponds to everyone having 1 sibling, the second corresponds to the friends having 0, 1 and 2 siblings, and the last term corresponds to two of the friends having 0 siblings, while one has 3.

d)

Let E and J denote Emma and Jacob's number of siblings, respectively. We assume these random variables to be independent, and arrive at the following:

$$P(E = 0 | J + E = 3) = \frac{P(J = 3)}{P(J + E = 3)} = \frac{0.06}{0.15 \cdot 0.06 \cdot 2 + 0.49 \cdot 0.27 \cdot 2} = \underline{0.212} \quad (4)$$

Problem 2

a)

True. The CPT (Conditional Probability Table) of each node needs to store 2^k entries, where k is the number of parents. This results in $2 \cdot 1 + 4 \cdot 2 + 2 \cdot 4 = 18$ numbers. The reason why we don't need to store e.g. $P(A = \text{true})$ and $P(A = \text{false})$ is that we know these must sum to 1.

b)

True. A and G are marginally independent, as there is no way that information can "flow" between them given one or the other.

c)

True. H 's Markov blanket is given, which implies that it is conditionally independent of all other nodes in the network, including E .

d)

False. We cannot conclude that this statement is true. For example, if the value of E is given, this could impact the distribution of G and in turn the distribution of H .

Problem 3

a)

$$P(b) = P(b|\neg a)P(\neg a) + P(b|a)P(a) = \underline{0.44}$$

b)

First calculate $P(\neg b) = P(\neg b|\neg a)P(\neg a) + P(\neg b|a)P(a) = 0.56$. Then we get

$$P(d) = P(d|\neg b)P(\neg b) + P(d|b)P(b) = \underline{0.712}. \quad (5)$$

c)

$$P(c|\neg d) = \frac{P(c, \neg d)}{P(\neg d)} = \frac{\sum_b P(c, \neg d, b)}{1 - P(d)} = \dots = \underline{0.1778}. \quad (6)$$

See the appendix for details.

d)

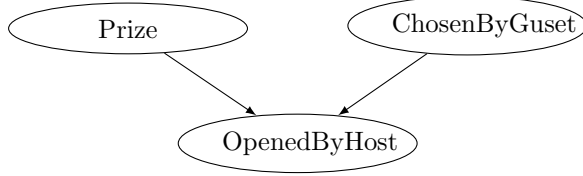
$$P(a|\neg c, d) = \frac{\sum_b P(a, \neg c, d, b)}{\sum_b P(\neg c, d, b)} = \dots = \underline{0.7983}. \quad (7)$$

See the appendix for details.

Problem 4

Below is the Bayesian network with CPTs depicted. 'P' is short for 'Prize', 'CBH' is short for 'ChosenByGuest' and 'OBH' is short for 'OpenedByHost'. Note: 0-indexing has been used, so the first door corresponds to 0, the second door corresponds to 2 and so on.

Prize		
0	1	2
1/3	1/3	1/3



ChosenByGuest		
0	1	2
1/3	1/3	1/3

		OBH		
P	CBG	0	1	2
0	0	0	0.5	0.5
1	0	0	0	1
2	0	0	1	0
0	1	0	0	1
1	1	0.5	0	0.5
2	1	1	0	0
0	2	0	1	0
1	2	1	0	0
2	2	0.5	0.5	0

Using the above model, the distribution in question can be calculated as

$$P(\text{Prize} | \text{ChosenByGuest} = 0, \text{OpenedByHost} = 2) = \langle \frac{1}{3}, \frac{2}{3}, 0 \rangle \\ \approx \langle 0.333, 0.667, 0 \rangle$$

My implementation yields the approximate result above, which clearly shows that it is advantageous to switch your choice.

Appendix

$$\begin{aligned}
 c) \quad P(c|\neg d) &= \frac{P(c, \neg d)}{P(\neg d)} = \frac{\sum_B P(c, \neg d, B)}{P(\neg d)} \\
 &= \frac{P(c, \neg d|b) \cdot P(b) + P(c, \neg d|\neg b) \cdot P(\neg b)}{P(\neg d)} \\
 &= \frac{P(c|b) \cdot P(\neg d|b) + P(c|\neg b) \cdot P(\neg d|\neg b) \cdot P(\neg b)}{1 - P(d)} \\
 &= \frac{0.1 \cdot 0.4 + 0.3 \cdot 0.2 \cdot (1 - 0.44)}{1 - 0.412} = \cancel{0.3472} = 0.1778
 \end{aligned}$$

Figure 1: Detailed calculations for 3c.

$$\begin{aligned}
 d) \quad P(a|\neg c, d) &= \frac{P(a, \neg c, d)}{P(\neg c, d)} \\
 &= \frac{\sum_B P(a, \neg c, d, B)}{P(\neg c, d)} \\
 &= \frac{P(\neg c, d|a, b) \cdot P(a, b) + P(\neg c, d|a, \neg b) \cdot P(a, \neg b)}{P(\neg c, d|b) + P(\neg c, d|\neg b)} \\
 &= \frac{[P(\neg c|b) \cdot P(d|b) \cdot P(b) + P(\neg c|\neg b) \cdot P(d|\neg b) \cdot P(\neg b)] \cdot P(a)}{P(\neg c|b) \cdot P(d|b) \cdot P(b) + P(\neg c|\neg b) \cdot P(d|\neg b) \cdot P(\neg b)} \\
 &= \frac{0.8 \cdot [0.9 \cdot 0.6 \cdot 0.5 + 0.7 \cdot 0.8 \cdot 0.5]}{0.9 \cdot 0.6 \cdot 0.44 + 0.7 \cdot 0.8 \cdot (1 - 0.44)} \\
 &= 0.7983.
 \end{aligned}$$

Figure 2: Detailed calculations for 3d.