

TDT4136 - ARTIFICIAL INTELLIGENCE METHODS

ASSIGNMENT 1: UNCERTAINTY, BAYESIAN NETWORKS

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Problem 1

The probability that a person has 0, 1, 2, 3, 4, or 5 or more siblings is 0.15, 0.49, 0.27, 0.06, 0.02, 0.01, respectively.

a) *What is the probability that a child has at most two siblings?*

$$P(X \leq 2) = P(X = 0) + P(X = 1) + P(X = 2) = 0.15 + 0.49 + 0.27 = 0.91$$

b) *What is the probability that a child has more than 2 siblings given that he has at least 1 sibling?*

$$P(X > 2 | X \geq 1) = \frac{P(X > 2 \wedge X \geq 1)}{P(X \geq 1)} = \frac{P(X > 2)}{P(X \geq 1)} = \frac{1 - P(X \leq 2)}{1 - P(X = 0)} = \frac{1 - 0.91}{1 - 0.15} = \frac{0.09}{0.85} = 0.11$$

c) *Three friends who are not siblings are gathered. What is the probability that they combined have three siblings?*

$$\begin{aligned} P(X+Y+Z = 3) &= P(X = 1 \wedge Y = 1 \wedge Z = 1) + 3P(X = 3 \wedge Y = 0 \wedge Z = 0) + 6P(X = 0 \wedge Y = 1 \wedge Z = 2) \\ &= P(X = 1) \times P(Y = 1) \times P(Z = 1) + 3 \times P(X = 3) \times P(Y = 0) \times P(Z = 0) \\ &\quad + 6 \times P(X = 0) \times P(Y = 1) \times P(Z = 2) = 0.15^3 + 3 \times 0.06 \times 0.15^2 + 6 \times 0.15 \times 0.49 \times 0.27 \\ &= 0.12 \end{aligned}$$

d) *Emma and Jacob are not siblings, but combined they have a total of 3 siblings. What is the probability that Emma has no siblings?*

$$\begin{aligned} P(X = 0 | X+Y = 3) &= \frac{P(X = 0 \wedge X+Y = 3)}{P(X+Y = 3)} = \frac{P(X = 0 \wedge X+Y = 3)}{2 \times P(X = 3 \wedge Y = 0) + 2 \times P(X = 2 \wedge Y = 1)} \\ &= \frac{P(X = 0) \times P(Y = 3)}{2 \times P(X = 3) \times P(Y = 0) + 2 \times P(X = 2) \times P(Y = 1)} = \frac{0.15 \times 0.06}{2 \times 0.06 \times 0.15 + 2 \times 0.27 \times 0.49} \\ &= 0.032 \end{aligned}$$

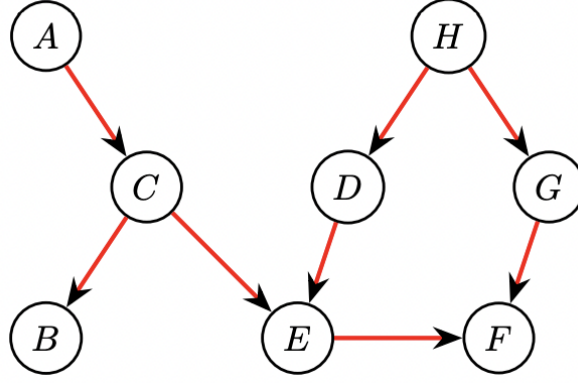


Figure 1: Bayesian network for problem 2.

Problem 2

a) *If every variable in the network has a Boolean state, then the Bayesian network can be represented with 18 numbers.*

Variables A and H have zero parents each which gives $1 \times 1 = 2$ numbers. B , C , D and G have one parent each which gives $4 \times 2 = 8$ numbers. E and F have two parent each which gives $2 \times 4 = 8$ numbers. In total the Bayesian network can be represented with $2 + 8 + 8 = 18$ numbers.

b) $G \perp\!\!\!\perp A$

We have that $P(G|A) = P(G)$ and $P(A|G) = P(A)$, thus the random variables G and A are independent.

c) $E \perp\!\!\!\perp H | \{D, G\}$

Path $E \leftarrow D \leftarrow H$ is blocked by conditioning on D . Path $E \rightarrow F \leftarrow G \rightarrow H$ is blocked by conditioning on G . E and H are d-separated by D and G , consequently, E and H are conditional independent given D and G .

d) $E \perp\!\!\!\perp H | \{C, D, F\}$

By introducing the conditioning on F , the second path is not blocked by a collider around F since it is being conditioned and not d-separated. Therefore one can not guarantee that E and H are conditional independent given C , D and G .

Problem 3

The Bayesian network below contains only binary states. The conditional probability for each state is listed. From the Bayesian network, calculate the following probabilities:

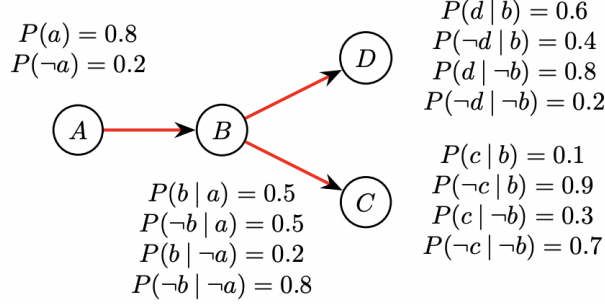


Figure 2: Bayesian network for problem 3.

a) $P(b) = P(b|a) \times P(a) + P(b|\neg a) \times P(\neg a) = 0.5 \times 0.8 + 0.2 \times 0.2 = 0.44$

b) It follows from a) that $P(\neg b) = 1 - 0.44 = 0.56$

$P(d) = P(d|b) \times P(b) + P(d|\neg b) \times P(\neg b) = 0.6 \times 0.44 + 0.8 \times 0.56 = 0.712$

c) Similarly from b) we have that $P(\neg d) = 1 - 0.712 = 0.288$. D and C are conditional independent given B which results in $P(C, D|B) = P(C|B) \times P(D|B)$.

$$P(c|\neg d) = \frac{P(c, \neg d)}{P(\neg d)} = \frac{P(c, \neg d) \times P(b) + P(c, \neg d) \times P(\neg b)}{P(\neg d)} = \frac{P(c|b) \times P(\neg d|b) \times P(b) + P(c|\neg b) \times P(d|\neg b) \times P(\neg b)}{P(\neg d)} = \frac{0.1 \times 0.4 \times 0.44 + 0.3 \times 0.2 \times 0.56}{0.288} = 0.178$$

d) $P(a|\neg c, d) = \alpha P(a, \neg c, d) = \alpha \sum_b P(a, b, \neg c, d) = \alpha \sum_b P(a) \times P(b|a) \times P(\neg c|b) \times P(d|b) = \alpha \times P(a) \times \sum_b P(b|a) \times P(\neg c|b) \times P(d|b) = \alpha(0.8(0.5 \times 0.9 \times 0.6 + 0.5 \times 0.7 \times 0.8)) = \alpha(0.44)$. Similar calculations for $P(\neg a|\neg c, d) = \alpha(0.1112)$. Hence $P(a|\neg c, d) = 0.798$

Problem 4

Code for Problem 4 is implemented in ***assignment_1.py***.

a) Kahn's algorithm was chosen for the topological ordering of the node. The implementation is based on the pseudocode described in Wikipedia's page about *Topological sorting* [1].

b) The following screenshot in [Figure 3](#) presents the output from running the implementation of the inference by enumeration algorithm found in *Artificial Intelligence: A Modern Approach* on the query from Problem 3c [2].

c) A model of the Monty Hall problem is visualized in [Figure 4](#) with the three random variables *ChosenByGuest*, *OpenedByHost* and *Prize*. Each of these variables has three states, namely door 0, 1 and 2 for convenience. The results from the implementation of the inference by enumeration algorithm on the Monty Hall problem query is shown in [Figure 5](#). From the results one can conclude that it is to ones advantage to switch the choice after the host opens the door.

```

Probability distribution, P(A)
+-----+-----+
|  A(0)  |  0.8000  |
+-----+-----+
|  A(1)  |  0.2000  |
+-----+-----+

Probability distribution, P(B | A)
+-----+-----+-----+
|    A    |  A(0)  |  A(1)  |
+-----+-----+-----+
|  B(0)  |  0.5000 |  0.2000 |
+-----+-----+-----+
|  B(1)  |  0.5000 |  0.8000 |
+-----+-----+-----+

Probability distribution, P(C | B)
+-----+-----+-----+
|    B    |  B(0)  |  B(1)  |
+-----+-----+-----+
|  C(0)  |  0.1000 |  0.3000 |
+-----+-----+-----+
|  C(1)  |  0.9000 |  0.7000 |
+-----+-----+-----+

Probability distribution, P(D | B)
+-----+-----+-----+
|    B    |  B(0)  |  B(1)  |
+-----+-----+-----+
|  D(0)  |  0.6000 |  0.8000 |
+-----+-----+-----+
|  D(1)  |  0.4000 |  0.2000 |
+-----+-----+-----+

Probability distribution, P(C | !D)
+-----+-----+
|  C(0)  |  0.1778 |
+-----+-----+
|  C(1)  |  0.8222 |
+-----+-----+

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Figure 3: Results from Problem 3c.

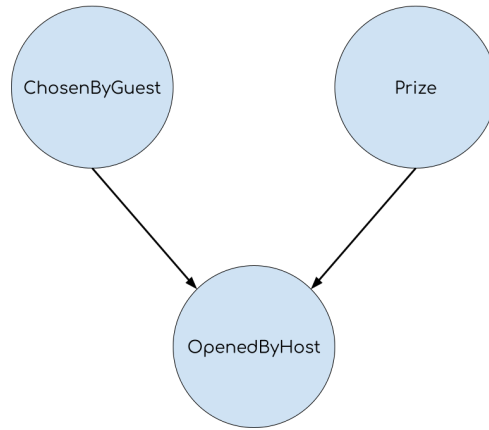


Figure 4: Visual model of the Monty Hall problem as a Bayesian Network.

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Probability distribution, P(P)
+-----+
| P(0) | 0.3333 |
+-----+
| P(1) | 0.3333 |
+-----+
| P(2) | 0.3333 |
+-----+

Probability distribution, P(C)
+-----+
| C(0) | 0.3333 |
+-----+
| C(1) | 0.3333 |
+-----+
| C(2) | 0.3333 |
+-----+

Probability distribution, P(O | P,C)
+-----+
| P | P(0) | P(1) | P(2) | P(0) | P(1) | P(2) | P(0) | P(1) | P(2) |
+-----+
| C | C(0) | C(0) | C(0) | C(1) | C(1) | C(1) | C(2) | C(2) | C(2) |
+-----+
| O(0) | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.5000 | 1.0000 | 0.0000 | 1.0000 | 0.5000 |
+-----+
| O(1) | 0.5000 | 0.0000 | 1.0000 | 0.0000 | 0.0000 | 0.0000 | 1.0000 | 0.0000 | 0.5000 |
+-----+
| O(2) | 0.5000 | 1.0000 | 0.0000 | 1.0000 | 0.5000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 |
+-----+

Probability distribution, P(P | C=0,O=2)
+-----+
| P(0) | 0.3333 |
+-----+
| P(1) | 0.6667 |
+-----+
| P(2) | 0.0000 |
+-----+

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Figure 5: Implementation of inference by enumeration algorithm on query from Monty Hall problem.

Bibliography

- [1] Wikipedia, “Topological sorting — Wikipedia, the free encyclopedia,” <http://en.wikipedia.org/w/index.php?title=Topological%20sorting&oldid=1001259852>, 2021, [Online; accessed 22-Jebruary-2021].
- [2] P. N. Stuart Russell, *Artificial Intelligence: A Modern Approach (4th Edition) (Pearson Series in Artifical Intelligence)*, 4th ed. Language: English, 2020.