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# CMSI 485 – Classwork 3

## Solution

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**Instructions:**

This worksheet will not only provide you with practice problems for your upcoming exam, but will add to your deep understanding of the mechanics of many probabilistic reasoning systems.

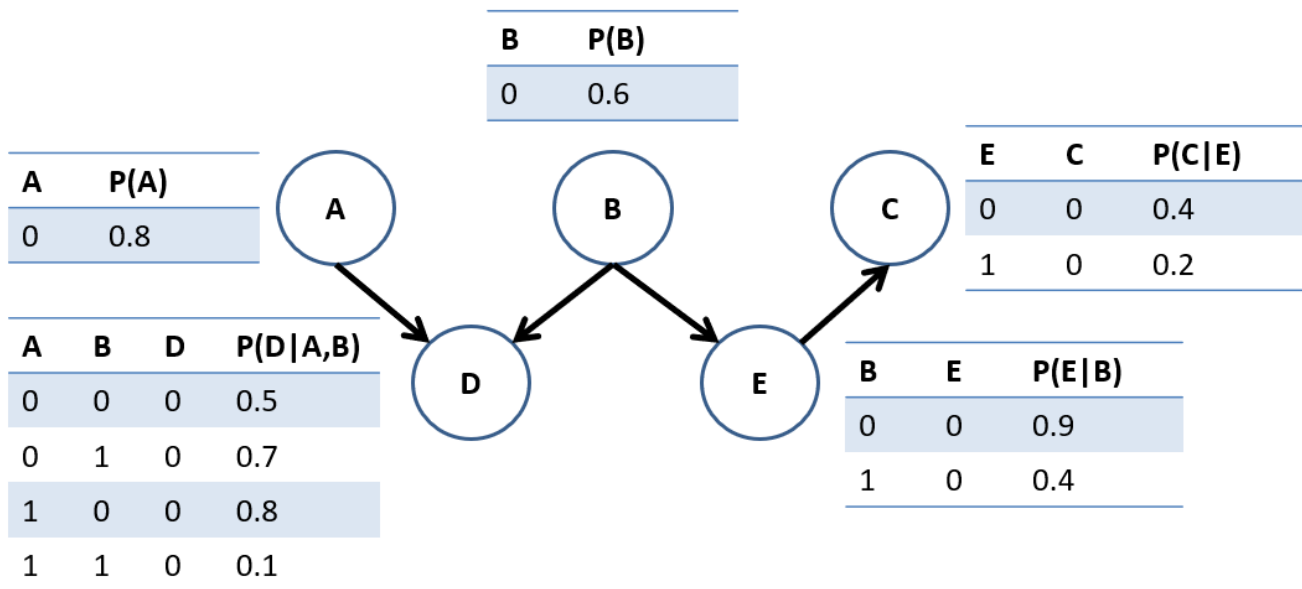
- Provide answers to each of the following questions and write your responses in the blanks. If you are expected to show your work in arriving at a particular solution, space will be provided for you.
- Place the names of your group members below:

**Group Members:**

1. \_\_\_\_\_
2. \_\_\_\_\_
3. \_\_\_\_\_
4. \_\_\_\_\_

## Problem 1 – Bayesian Network Exact Inference

Consider the following Bayesian Network and use it to answer the questions that follow.



While examining Exact Inference in Bayesian Networks, we saw some methods for *simplifying* queries and the resulting computations that can yield large performance improvements when implemented. E.g., variables whose CPTs never affect the query outcome can be ignored.

- 1.1.** For each of the following queries, determine *which variables' CPTs will at all affect* the answer to the query. Justify your responses in the boxes that follow. Hint: in your justification, express the query in terms of:  $P(A|B) = \alpha \sum P(\dots)$  with  $\alpha = \frac{1}{P(e)}$

- 1.1.1.  $P(A|B = b, D = d)$

CPTs Used: ☒ A ☒ B ☐ C ☒ D ☐ E

Justification:

$$\begin{aligned}
 P(A|b, d) &= \alpha \sum_e \sum_c P(A, b, c, d, e) = \alpha \sum_e \sum_c P(A)P(d|A, b)P(b)P(e|b)P(c|e) \\
 &= \alpha P(A)P(d|A, b)P(b) \sum_e P(e|b) \sum_c P(c|e) = \alpha P(A)P(d|A, b)P(b)
 \end{aligned}$$

- 1.1.2.  $P(E|D = d)$

CPTs Used: ☒ A ☒ B ☐ C ☒ D ☒ E

Justification:

$$\begin{aligned}
 P(E|d) &= \alpha \sum_a \sum_b \sum_c P(a, b, c, d, E) = \alpha \sum_a \sum_b \sum_c P(a)P(d|a, b)P(b)P(E|b)P(c|E) \\
 &= \alpha \sum_b P(b)P(E|b) \sum_a P(a)P(d|a, b) \sum_c P(c|E) = \alpha \sum_b P(b)P(E|b) \sum_a P(a)P(d|a, b)
 \end{aligned}$$

1.2. Using the Bayesian Network on the previous page, find the solutions to the following.

1.2.1.  $P(A = 0|B = 1, D = 1)$

(Box your answer once finished)

Step 1 – Label Vars:  $Q$  = query,  $e$  = evidence,  $Y$  = hidden

$$Q = \{A\}, e = \{B = 1, D = 1\}, Y = \{C, D\}$$

Step 2 – Find  $P(Q, e) = \sum_y P(Q, e, y)$  (taken from 1.1.1 above):

$$P(A|B = 1, D = 1) = \alpha \sum_e \sum_c P(A, B = 1, D = 1, C = c, E = e) = \alpha P(A)P(d|A, b)P(b)$$

$$A = 0 \Rightarrow P(A, b, d) = P(A = 0)P(D = 1|A = 0, B = 1)P(B = 1) = 0.8 * 0.3 * 0.4 = 0.096$$

$$A = 1 \Rightarrow P(A, b, d) = P(A = 1)P(D = 1|A = 1, B = 1)P(B = 1) = 0.2 * 0.9 * 0.4 = 0.072$$

Step 3 – Find  $P(e) = \sum_q P(Q = q, e)$ :

$$P(e) = P(B = 1, D = 1)$$

$$= \sum_a P(a, b, d) = P(A = 0, B = 1, D = 1) + P(A = 1, B = 1, D = 1)$$

$$= 0.096 + 0.072 = 0.168$$

Step 4 – Solve query:  $P(Q|e) = \frac{P(Q, e)}{P(e)} = \frac{\{\text{Step 2}\}}{\{\text{Step 3}\}}$

$$\therefore P(A = 0|B = 1, D = 1) = \frac{P(A = 0, B = 1, D = 1)}{P(B = 1, D = 1)} = \frac{0.096}{0.168} \approx 0.571$$

1.2.2  $P(B = 1|A = 0, D = 1, E = 0, C = 1)$

(Box your answer once finished)

Step 1 – Label Vars:

$$Q = \{B\}, e = \{A = 0, C = 1, D = 1, E = 0\}, Y = \emptyset$$

Step 2 – Find  $P(Q, e) = \sum_y P(Q, e, y)$ :

$$\begin{aligned} P(B|A = 0, D = 1, E = 0, C = 1) \\ = \alpha P(A = 0)P(B)P(D = 1|A = 0, B)P(E = 0|B)P(C = 1|E = 0) \end{aligned}$$

$$B = 0 \Rightarrow P(B, a, c, d, e)$$

$$= P(A = 0)P(B = 0)P(D = 1|A = 0, B = 0)P(E = 0|B = 0)P(C = 1|E = 0)$$

$$= 0.8 * 0.6 * 0.5 * 0.9 * 0.6 = 0.1296$$

$$B = 1 \Rightarrow P(B, a, c, d, e)$$

$$= P(A = 0)P(B = 1)P(D = 1|A = 0, B = 1)P(E = 0|B = 1)P(C = 1|E = 0)$$

$$= 0.8 * 0.4 * 0.3 * 0.4 * 0.6 = 0.02304$$

Step 3 – Find  $P(e) = \sum_q P(Q = q, e)$ :

$$P(e) = P(A = 0, D = 1, E = 0, C = 1) = \sum_b P(B, a, c, d, e)$$

$$= P(B = 0, A = 0, D = 1, E = 0, C = 1) + P(B = 1, A = 0, D = 1, E = 0, C = 1)$$

$$= 0.1296 + 0.02304 = 0.15264$$

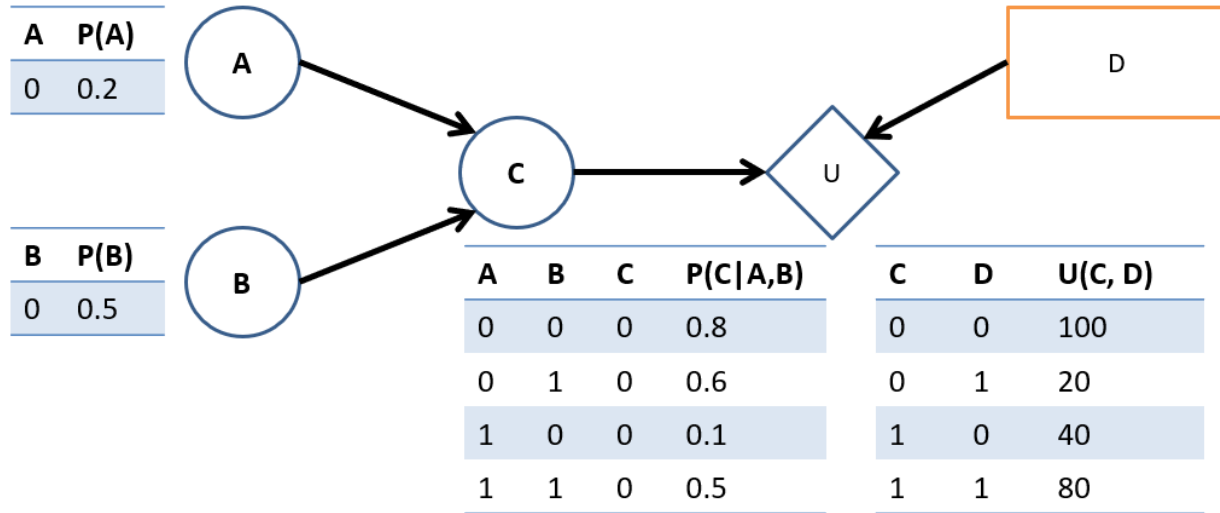
Step 4 – Solve query:  $P(Q|e) = \frac{P(Q, e)}{P(e)} = \frac{\{\text{Step 2}\}}{\{\text{Step 3}\}}$ .

$$\begin{aligned} \therefore P(B = 1|A = 0, D = 1, E = 0, C = 1) &= \frac{P(B = 1, A = 0, D = 1, E = 0, C = 1)}{P(A = 0, D = 1, E = 0, C = 1)} \\ &= \frac{0.02304}{0.15264} \approx 0.151 \end{aligned}$$

(sorry for the messy probability values there, oh well... life isn't always nice and neat!)

## Problem 2 – Decision Networks & MEU

Use the following Decision Network with chance nodes  $A, B, C$ , decision node  $D \in \{0,1\}$ , and utility node  $U$  to answer the questions that follow.



2.1. Find the  $MEU(B = 0)$  (i.e., the Maximum Expected Utility with evidence  $B = 0$ ).

*Reminder: (for evidence  $e$ , actions  $a$ , and chance nodes that are parents of utility  $s$ ):*

$$MEU(e) = \max_a EU(a|e) = \sum_s P(s|a, e)U(s, a)$$

$$MEU(B = 0) = \max_d EU(d|B = 0) = \max_d \sum_c P(c|d, B = 0)U(c, d)$$

Since  $P(c|d, B = 0)$  is not in terms of our network parameters (the CPTs) we'll transform it:

$$\sum_c P(c|d, B = 0)U(c, d) = \sum_c P(c|B = 0)U(c, d) = \sum_c \sum_a P(c|a, B = 0)P(a)U(c, d)$$

Now, computed for each available action:

$$\begin{aligned} EU(D = 0|B = 0) &= \sum_c \sum_a P(c|a, B = 0)P(a)U(c, D = 0) \\ &= P(C = 0|A = 0, B = 0)P(A = 0)U(C = 0, D = 0) \\ &\quad + P(C = 0|A = 1, B = 0)P(A = 1)U(C = 0, D = 0) \\ &\quad + P(C = 1|A = 0, B = 0)P(A = 0)U(C = 1, D = 0) \\ &\quad + P(C = 1|A = 1, B = 0)P(A = 1)U(C = 1, D = 0) \\ &= 0.8 * 0.2 * 100 + 0.1 * 0.8 * 100 + 0.2 * 0.2 * 40 + 0.9 * 0.8 * 40 = 54.4 \end{aligned}$$

$$\begin{aligned} EU(D = 1|B = 0) &= \sum_c \sum_a P(c|a, B = 0)P(a)U(c, D = 1) \\ &= P(C = 0|A = 0, B = 0)P(A = 0)U(C = 0, D = 1) \\ &\quad + P(C = 0|A = 1, B = 0)P(A = 1)U(C = 0, D = 1) \\ &\quad + P(C = 1|A = 0, B = 0)P(A = 0)U(C = 1, D = 1) \\ &\quad + P(C = 1|A = 1, B = 0)P(A = 1)U(C = 1, D = 1) \\ &= 0.8 * 0.2 * 20 + 0.1 * 0.8 * 20 + 0.2 * 0.2 * 80 + 0.9 * 0.8 * 80 = 65.6 \end{aligned}$$

$$\therefore MEU(B = 0) = 65.6$$

[!] Note: The above derivation solved for the computable query  $P(C|B = 0)$  using the insight that summing over A would yield expressions in terms of the CPTs – it did NOT use enumeration inference, which you can always do, like in the following derivation:

*Step 1 – Label Vars:*

$$Q = \{C\}, e = \{B = 0\}, Y = \{A\}$$

*Step 2 – Find  $P(Q, e) = \sum_y P(Q, e, y)$ :*

$$P(C, B = 0) = \sum_a P(C, B = 0, A = a) = P(B = 0) \sum_a P(A = a)P(C|B = 0, A = a)$$

$$C = 0 \Rightarrow P(C = 0, B = 0) = 0.12$$

$$C = 1 \Rightarrow P(C = 1, B = 0) = 0.38$$

*Step 3 – Find  $P(e) = \sum_q P(Q = q, e)$ :*

$$\begin{aligned} P(e) &= P(B = 0) = \sum_c P(C = c, B = 0) \\ &= 0.12 + 0.38 = 0.5 \end{aligned}$$

*(which you can double check from the CPT, since the  $P(e)$  happened to be immediately answerable in this particular problem).*

*Step 4 – Solve query:  $P(Q|e) = \frac{P(Q,e)}{P(e)} = \frac{\{\text{Step 2}\}}{\{\text{Step 3}\}}$ .*

$$P(C = 0|B = 0) = \frac{P(C = 0, B = 0)}{P(B = 0)} = \frac{0.12}{0.5} = 0.24$$

$$P(C = 1|B = 0) = \frac{P(C = 1, B = 0)}{P(B = 0)} = \frac{0.38}{0.5} = 0.76$$

*Back to the MEU Computation...*

$$MEU(B = 0) = \max_d EU(d|B = 0) = \max_d \sum_c P(c|B = 0)U(c, d)$$

$$\begin{aligned} EU(D = 0|B = 0) &= P(C = 0|B = 0)U(C = 0, D = 0) + P(C = 1|B = 0)U(C = 1, D = 0) \\ &= 0.24 * 100 + 0.76 * 40 = 54.4 \end{aligned}$$

$$\begin{aligned} EU(D = 1|B = 0) &= P(C = 0|B = 0)U(C = 0, D = 1) + P(C = 1|B = 0)U(C = 1, D = 1) \\ &= 0.24 * 20 + 0.76 * 80 = 65.6 \end{aligned}$$

$$\therefore MEU(B = 0) = 65.6$$

2.2. Given your computations above, what decision should your agent make by *MEU*?

Since  $d^* = \operatorname{argmax}_d EU(d|B = 0) = 1$ , our agent should choose action  $D = 1$ .

### Problem 3 – Value of Perfect Information

Using the network and your answer from the previous problem, we're going to compute the Value of Perfect Information (VPI) of knowing the state of variable  $A$  when  $B = 0$  is given. Let's do so step-by-step:

3.1. Find  $MEU(A = 0, B = 0)$ .

$$MEU(A = 0, B = 0) = \max_d EU(d|A = 0, B = 0) = \max_d \sum_c P(c|A = 0, B = 0)U(c, d)$$

$$EU(D = 0|A = 0, B = 0) = \sum_c P(c|A = 0, B = 0)U(c, D = 0)$$

$$= P(C = 0|A = 0, B = 0)U(C = 0, D = 0) + P(C = 1|A = 0, B = 0)U(C = 1, D = 0) \\ = 0.8 * 100 + 0.2 * 40 = 88$$

$$EU(D = 1|A = 0, B = 0) = \sum_c P(c|A = 0, B = 0)U(c, D = 1)$$

$$= P(C = 0|A = 0, B = 0)U(C = 0, D = 1) + P(C = 1|A = 0, B = 0)U(C = 1, D = 1) \\ = 0.8 * 20 + 0.2 * 80 = 32$$

$$\therefore MEU(A = 0, B = 0) = 88$$

3.2. Using your answer to 3.1 and knowledge that  $MEU(A = 1, B = 0) = 74$  (freebee!) find  $MEU(A, B = 0)$ .

We have to weight the  $MEU(A = a, B = 0)$  by the likelihood of  $A = a|B = 0$ :

$$MEU(A, B = 0) = \sum_a P(a|B = 0)MEU(a, B = 0) = \sum_a P(a)MEU(a, B = 0)$$

$$= P(A = 0)MEU(A = 0, B = 0) + P(A = 1)MEU(A = 1, B = 0) \\ = 0.2 * 88 + 0.8 * 74 = 76.8$$

3.3. Compute the  $VPI(A|B = 0)$ .

$$VPI(A|B = 0) = MEU(A, B = 0) - MEU(B = 0) = 76.8 - 65.6 = 11.2$$

3.4. If the utility scores represent dollar amounts, what would be a fair price for  $A$  when  $B = 0$ ? The fair price is given by the VPI! That's its definition. So, the fair cost for info on  $A$  is: \$11.20. If we bought it for less, it would be a good deal, and for more, a ripoff.