

CS3243 Tutorial 9

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April 7, 2022

Announcements

1. Assignment 7 scores are now on Gradebook.
2. Assignment 8 scores are being marked, will appear on tunitin, please check.
3. Highly recommended to send in your teaching feedback - I appreciate it greatly!
4. Next week is a recap tutorial; But will be rescheduled.

Student Feedback on Teaching (SFT)

Feedback is *optional* but *highly encouraged*, access here: <https://es.nus.edu.sg/blue/>

- **[Tutorial Feedback]** Your feedback is important to me, and will be used to improve my teaching.
 - If I have helped your learning in any way, your positive feedback will be an encouragement to me.
 - If you find your learning can be enhanced by some action on my part, that feedback will be used to improve my teaching.
- **[Module Feedback]** Your feedback will be used to improve the module.
- Feedback is confidential to the university and anonymous to us.
- Avoid mixing the feedback; ie. project feedback to tutorial feedback.

Past student feedback had been used to improve teaching; ie. Telegram access to provide faster feedback. I would greatly appreciate your feedback, especially this is my first time teaching AI.

Recap tutorial

- Tutorial will be rescheduled due to college well-being day;
 - I will consolidate the 3 tutorials into 1 or 2.
 - Select all that you are available: **Apr 13, 3-4pm or 4-5pm or 5-6pm.**
- Prof Daren will upload a review tutorial: No assignment questions & 4 past yr qns.
- I have decided to help all of you with a special recap; two options
 1. Recap one topic + Quickly go through the 4 past year questions
 2. Go through 4 past year questions + Time for Q&A at the end



Figure 1: Vote Now - <https://forms.gle/xCiVFZKmyE93NTPFA>

Previously from T08, Q1

Consider below, a Vertex Cover where it is a set of vertices that covers all edges.

- i. Write down the constraints as logical statements for a vertex cover of size 1.
- ii. Apply the resolution algorithm in order to prove that the vertex 1 must be part of the vertex cover.

Recap

- What is a Vertex Cover of size k ?
- How to formulate a *KB* problem?

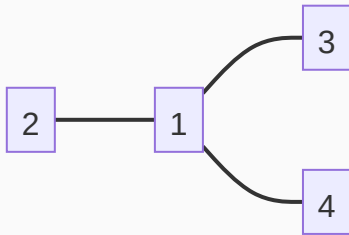


Figure 2: Graph for Vertex Cover CSP

Answer T8.Q1.i - Formulating KB

Variables:

- x_i represents i node on the graph that is in the vertex cover.
- $\neg x_i$ represents not in the vertex cover.

Constraints:

- Edge cover constraints: $x_1 \vee x_2$; $x_1 \vee x_3$; $x_1 \vee x_4$
- Size $k = 1$ constraints: (We ignore contrapositive)
 - x_1 set then... $x_1 \implies \neg x_2$; $x_1 \implies \neg x_3$; $x_1 \implies \neg x_4$
 - x_2 set then... $x_2 \implies \neg x_3$; $x_2 \implies \neg x_4$;
 - x_3 set then... $x_3 \implies \neg x_4$

Then convert to CNF!

Answer T8.Q1.ii

Show $KB \models \alpha = x_1$; we resolve $KB \wedge \neg x_1$

1. $\neg x_1 \oplus x_1 \vee x_2 \implies x_2$

2. $x_2 \oplus \neg x_2 \vee \neg x_3 \implies \neg x_3$

3. $\neg x_1 \oplus x_1 \vee x_3 \implies x_3$

4. $x_3 \oplus \neg x_3 \implies \square$

Question 1

Having both good grades (G) and good communication skills (C) will increase your chances of performing well in your interview (I).

Table 1: Probability of I ; $Pr[G = 1] = 0.7, Pr[C = 1] = 0.2$

G	C	$Pr[I = 1 G, C]$
1	1	0.9
1	0	0.5
0	1	0.5
0	0	0.1

Recap

- How to read a Bayes Network?
- What is Marginalisation?

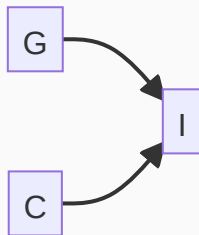


Figure 3: Belief Network

Answer

What is the probability that

- a. Alice, who has poor grades and communication skills, performs well on her interview? - $Pr[I = 1|G = 0, C = 0] = 0.1$
- b. Bob is a student with great communication skills, assuming we do not know anything about him? - $Pr[C = 1] = 0.2$
- c. A student has good communication skills, given that he or she has performed well in an interview? - $Pr[C = 1|I = 1] = \frac{Pr[C=1,I=1]}{Pr[I=1]} = \frac{\sum_g Pr[G=g,C=1,I=1]}{\sum_{g,c} Pr[G=g,C=c,I=1]} = 0.339$
 - a. Are good communication skills independent of good performance in an interview? - Remember independence; $Pr[C = 1|I = 1] \neq Pr[C = 1]$.

Question 2

Assume that 2% of the population in a country carry a particular virus (Y is a carrier).
A test kit developed to detect the presence (X is positive test).

Y	$Pr[X = 1 Y]$
1	0.998
0	$1 - 0.996 = 0.004$

Recap

- What is Conditional Probability?
- What is Conditional Independence?

Question 2

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Recap

- What is Conditional Probability?
- What is Conditional Independence?

Careful with the conditional independence here...

Answer 2a

Given that a patient is tested to be positive using this kit, what is the posterior belief that he is not a carrier?

$$Pr[Y = 0|X = 1] = \frac{Pr[Y=0,X=1]}{Pr[X=1]} = \frac{Pr[Y=0,X=1]}{\sum_y Pr[Y=y,X=1]} = \frac{Pr[X=1|Y=0] \times Pr[Y=0]}{\sum_y Pr[Y=y,X=1]} = 0.164$$

Answer 2b

Patient tested positive again using the second kit (X_2 is the second test, X_1 is the first test). Assume conditional independence between results of different test kits given the patient's state of virus contraction.

$$Pr[Y = 0|X_1 = 1, X_2 = 1] = 0.0008$$

- $= \frac{Pr[Y=0,X_2=1,X_1=1]}{Pr[X_1=1,X_2=1]} = \frac{Pr[X_2=1,X_1=1|Y=0]Pr[Y=0]}{Pr[X_1=1,X_2=1]}$ (Conditional Probability)
- $= \frac{Pr[X_2=1|Y=0]Pr[X_1=1|Y=0]Pr[Y=0]}{Pr[X_1=1,X_2=1]}$ (Conditional Independence)
- $= \frac{Pr[X_2=1|Y=0]Pr[X_1=1|Y=0]Pr[Y=0]}{\sum_y Pr[X_1=1,X_2=1,Y=y]}$ (Marginalisation)
- $= \frac{Pr[X_2=1|Y=0]Pr[X_1=1|Y=0]Pr[Y=0]}{\sum_y Pr[X_1=1|Y=y]Pr[X_2=1|Y=y]Pr[Y=y]}$ (Conditional Probability, Independence)

Question 3

Construct a Bayesian network and determine the probability

$$Pr[WG = 1, RS = 1, R = 0, S = 0]$$

Recap

- Any useless variables from the table?

Answer

Lemma 1. Given two random boolean variables A and B, if $Pr[A|B] = 0$ and $Pr[A|\neg B] = 1$ then $Pr[A] = 1 - Pr[B]$; in fact, $A \equiv \neg B$.

Not difficult to proof (*Note the notation change.*):

$$Pr[A|B] = 0 \wedge Pr[A|\neg B] = 1 \implies Pr[A] = 1 - Pr[B]$$

$Pr[A]$

- $= Pr[A \wedge B] + Pr[A \wedge \neg B]$ (Marginalisation)
- $= Pr[A|B] \times Pr[B] + Pr[A|\neg B] \times Pr[\neg B]$ (Conditional Probability)
- $= Pr[\neg B] = 1 - Pr[B]$ (Subst given)

So, $Pr[A \wedge \neg B \wedge \dots] = Pr[\dots | A \wedge \neg B] \times Pr[A \wedge \neg B] = Pr[\dots | A] \times Pr[A]$,

- Since, $Pr[A \wedge \neg B] = Pr[A|\neg B] \times Pr[\neg B] = Pr[A]$;
- And, $Pr[\dots | A \wedge \neg B] = Pr[\dots \wedge \neg B | A] / Pr[\neg B | A] = Pr[\dots | A]$

Since $S \equiv \neg RS$, so:

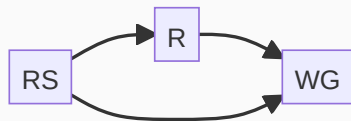


Figure 4: RS, WG, R network.

$$Pr[RS = 1] = 0.7$$

RS	$Pr[R = 1 RS]$
1	0.9
0	0.1

RS	R	$Pr[WG = 1 RS, R]$
1	1	0.8
1	0	0.1
0	1	0.95
0	0	0.9

$$Pr[WG = 1, RS = 1, R = 0, S = 0] = Pr[WG = 1, RS = 1, R = 0]$$

- $= Pr[WG = 1, R = 0|RS = 1] \times Pr[RS = 1]$ (Conditional Probability)
- $= Pr[WG = 1|RS = 1, R = 0] \times Pr[R = 0|RS = 1] \times Pr[RS = 1]$ (Cond. Prob.)
- $= 0.1 \times (1 - 0.9) \times 0.7 = 0.007$