INF5390 Obligatorisk oppgave 2

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1 First-order logic

We are going to represent the following sentences using first-order predicate logic:

- a. Not all students are enrolled in both History and Biology.
- b. Only one student failed History.
- c. Only one student failed both History and Biology.
- d. The best grade in History is better than the best grade in Biology.

In order to do this, we will use the following consistent vocabulary, consisting of constants and predicates:

- Constants:
 - History: a course in which students enroll
 - Biology: a course in which students enroll
- Predicates:
 - Student(s): True if s is a student
 - Enrolled (s, c): True if s is enrolled in course c
 - Failed(s, c): True if s failed course c
 - BestGrade(g, c): True if g is the best grade in course c
 - $Better(g_1, g_2)$: True if grade g_1 is better than grade g_2

Using the above vocabulary, we can represent a.-d. in first-order logic:

- a. "Not all Xs do Y" is the same as saying "At least one X does not do Y". Therefore: $\exists s(Student(s) \land \neg(Enrolled(s, History) \land Enrolled(s, Biology)))$
- b. $\exists !s(Student(s) \land Failed(s, History))$
- c. $\exists !s(Student(s) \land Failed(s, History) \land Failed(s, Biology))$
- d. $\exists g_1(BestGrade(g_1, History) \land \exists g_2(BestGrade(g_2, Biology) \land Better(g_1, g_2)))$

where $\exists!$ is the uniqueness quantifier (i.e. "there exists one and only one").

2 Agents that plan

Here our task is to define action schemas for the problem of putting on socks and shoes, hat and coat. We shall then define and sketch a partially ordered plan that solves the problem, and consequently evaluate the number of different linearizations of the solution.

The action schemas are:

Action(PutOnSocks), PRECOND: N/A EFFECT: SocksOn $\begin{aligned} \textbf{Action} & (PutOn \text{Shoes}), \\ & \text{PRECOND: } SocksOn \\ & \text{EFFECT: } ShoesOn \end{aligned}$

 $egin{aligned} Action(PutOnHat), \ & ext{PRECOND: N/A} \ & ext{EFFECT: } HatOn \end{aligned}$

 $\begin{array}{c} \boldsymbol{Action}(PutOnCoat), \\ \text{PRECOND: N/A} \\ \text{EFFECT: } \boldsymbol{CoatOn} \end{array}$

Table 1: Action schemas

The partially ordered plan can be defined as follows:

Initial state:

• $\neg SocksOn \land \neg ShoesOn \neg HatOn \land \neg CoatOn (== Undressed)$

Goal state:

- $SocksOn \wedge ShoesOn \wedge HatOn \wedge CoatOn (== Dressed)$
- Actions:
 - PutOnSocks, PRECOND: N/A, EFFECT: SocksOn
 - PutOnShoes, PRECOND: SocksOn, EFFECT: ShoesOn
 - PutOnHat, PRECOND: N/A, EFFECT: HatOn
 - PutOnCoat, PRECOND: N/A, EFFECT: CoatOn

A sketch of a partial order plan is shown in Figure 1.

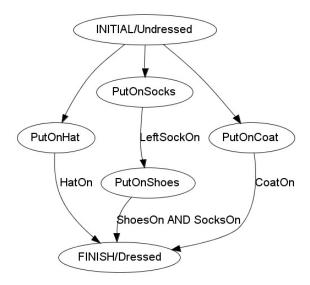


Figure 1: A partial order plan

If the order in which one equips the different clothes did not matter at all (socks/shoes included), the number of different permutations/orderings/linearizations would be 4! = 24 (4 initial positions; one for each piece of clothing). But since the socks need to be equipped *before* the shoes, this is number would obviously be erroneous. Specifically, this will slice the previous number in half, to 12 linearizations:

- Starting with socks: 3 free (order unimportant) actions (hat/shoes/coat) to choose from, thus 3! = 6 permutations.
- Starting with hat: 2 free actions (coat/socks) to choose from, but have to add 1 since the shoes are "locked" to the socks; 2! + 1 = 3 permutations.
- Starting with coat: 2 free actions (hat/socks) to choose from, but have to add 1 as per the previous bullet point; 2! + 1 = 3 permutations.
- So in total: 6+3+3=12 permutations/linearizations.

(The number of action orderings would be vastly greater if we specified the problem with left/right socks and shoes, but similar reasoning could be done).

3 Agents that reason under uncertainty

A factory alarm goes off if a thermometer reading exceeds a certain threshold value. We define the Boolean variables A (alarm produces sound), F_A (alarm fails) and F_M (thermometer fails), along with the multivalued variables M (thermometer) and T (actual temperature).

a) Bayesian network

The thermometer has a higher probability of failing when the temperature gets too high. With this in mind, we are going to sketch a Bayesian network of using the domain variables specified above. The dependencies needed are:

- As per the above, the probability of a faulty thermometer increases with actual temperature. Thus $T \to F_M$.
- The thermometer reading is obviously dependent upon the actual temperature. Thus $T \to M$.
- The event in which the thermometer fails obviously has an impact on the thermometer reading. Thus $F_M \to M$.
- Since the alarm goes off if the thermometer reading exceeds some (unspecified) limit, the alarm is obviously dependent upon the thermometer reading. Thus $M \to A$.
- Finally, the event in which the alarm goes off is dependent upon whether the alarm fails. Thus $F_A \to A$.

The network is illustrated in Figure 2.

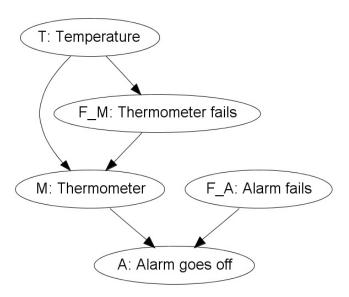


Figure 2: Bayesian network

b) Conditional Probability Table (CPT) for the alarm

We see from Figure 2 that the alarm, A, is dependent upon the thermometer reading, M, and the event in which the alarm fails, F_A .

We assume that there are only two possible thermometer readings: Normal and High. That is, M can now be thought of as a Boolean variable, with e.g. False/True indicating Normal/High temperature, respectively. We also assume that the alarm is always in working order unless it has failed (== not producing sound). In other words: if we assume that the alarm is supposed to go off whenever the thermometer reading is High (== exceeded threshold value), it will always go off when the temperature is High if it is not faulty $(\neg F_A)$.

Based on these assumptions, we can define the conditional probability table (CPT) for A as shown in Table 2.

M: Temperature reading	F_A : Alarm fails	P(A)	$P(\neg A)$
False	False	0	1
False	True	0	1
True	False	1	0
True	True	0	1

Table 2: CPT for the alarm