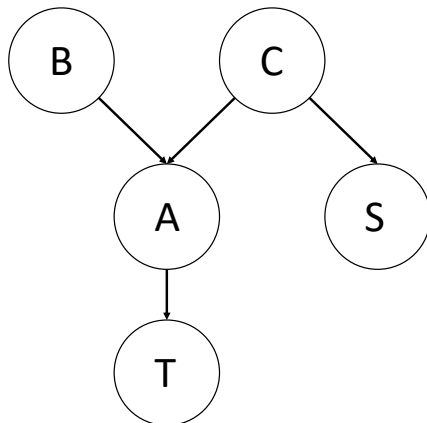


Updated at 10:00 AM on 4/6 to fix typos in question 4.

Our Bayes net questions will use the following network. Note that all variables are binary:



We motivate this structure with the following scenario: You have a crude burglar alarm installed at your house (A). Two things can cause the alarm to go off, a burglar (B) or vibration from construction in your neighborhood (C). If your alarm is ringing, you may get a text from your neighbor (T). If there is construction in your neighborhood, there might be a post about the construction on social media (S).

This network has the following CPTs. Note that the CPTs have probabilities that are convenient, not necessarily realistic:

$$P(b) = 1/8$$

$$P(c) = 1/4$$

B	C	$P(a BC)$
\bar{b}	\bar{c}	1/8
\bar{b}	c	3/8
b	\bar{c}	7/8
b	c	1

C	$P(s C)$
\bar{c}	$1/8$
c	$3/4$

A	$P(t A)$
\bar{a}	$1/8$
a	$3/4$

For all of the following questions, use variable elimination to compute the desired quantities. Show your work and any intermediate results computed explicitly.

1 Bayes Nets I (10 points)

Compute the marginal probability: $P(t)$.

2 Bayes Nets II (10 points)

Compute $P(b|t)$, the probability of burglary given that you got a text. (You can do this by computing $P(bt)$). Note that you should expect this to be higher than the prior probability of burglary, $P(b) = 1/8$.

3 Bayes Nets III (10 points)

Compute $P(st)$, the probability of a social media report about construction, and a text from your neighbor about your alarm.

4 Bayes Nets IV (10 points)

Compute $P(b|st)$. You can do this by computing $P(bst)$. This is the probability of burglary given *both* a social media report and a text from your neighbor. Comparing this with $P(b|t)$, you should expect a lower probability for this question. Why? The report about construction partly *explains away* the burglary as a cause for the text.

5 HMMs (30 points)

Consider a hidden Markov model with two states, x and y , two observations a and b , and the following somewhat unusual parameters. We use the random variable S_t for the state at time t :

- $P(S_{t+1} = x|S_t = x) = P(S_{t+1} = y|S_t = y) = 1$
- $P(a|S_t = x) = P(b|S_t = x) = 0.5$

- $P(b|S_t = y) = 1.0$

These parameters were chosen to help you realize some things about HMMs as you work through the basic algorithms. Assume that at time 0, the distribution over states is: $P(S_0 = x) = 0.6$, and that there are no observations at time 0. At time 1, the observation is b . At time 2, the observation is a .

a) Show the work to run the forward algorithm (AKA monitoring or tracking) and compute the distribution over states at time steps 1 and 2, i.e., $P(S_1|O_1 = b)$ and $P(S_2|O_1 = b, O_2 = a)$.

b) Show the work to compute the smoothed distribution over states at time step 1, i.e., $P(S_1|O_1 = b, O_2 = a)$, where O_t is the observation at time t . (Hint: You should see a big difference between the forward probabilities at time 1 and the smoothed probabilities.)

c) Show the work to compute the Viterbi path given the same initial distribution at time 0, and same observations, $O_1 = b$, then extend your calculation to include $O_2 = a$. Note that final your answers should be a sequence of 2 states up to $O_1 = b$, and then a sequence of 3 states when you extend your answer. You should observe something interesting about how the optimal path changes when you get the observation at time 2.