## COMPSCI 370

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## Homework 5

Due: Wednesday, April 21, 2021

## 1 Bayes Nets I (5 points)

Consider a Bayesian network with a simple, chain-like strucure and binary variables, $A \ldots E$, where $A$ has no parents, $B$ has $A$ as its only parent, $C$ has $B$ as its only parent, and so on. (We'll use the letters $A$ through $E$ for conciseness, but if you'd like a simple example of how such a network might arise, $A$ could be discipline, $B$ could be studying, $C$ could be knowledge, $D$ could be grades and $E$ could be Internship quality).

Suppose we wish to compute $P(E=e)$, by eliminating $A$, then $B$, then, $C$ and so on. How many multiplications and additions are required to do this? You should explain your reasoning and give an exact number, not a $O()$ bound in this case.

## 2 Bayes Nets II (10 points)

Consider the AFSHN network from class, and the following conditional probabilities:

- $P(a)=P(f)=\frac{1}{4}$
- $P(s \mid a f)=1, P(s \mid a \bar{f})=\frac{3}{4}, P(s \mid \bar{a} f)=\frac{3}{4}, P(s \mid \overline{a f})=\frac{1}{4}$
- $P(n \mid s)=\frac{3}{4}, P(h \mid s)=\frac{1}{2}, P(n \mid \bar{s}) \frac{1}{8}, P(h \mid \bar{s})=\frac{1}{4}$

Compute the joint probability distribution $P(H N)$ by variable elimination. You should write out your answer in a way that makes the order in which you are doing summations explicit. You should also show all entries for every table that you compute as a result of eliminating a variable, but you don't need to show every calculation to compute these entries, i.e., you can use a calculator and just plug the numbers into your table if you want.

## 3 HMMs (15 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:

- $P\left(S_{t+1}=x \mid S_{t}=x\right)=P\left(S_{t+1}=y \mid S_{t}=y\right)=1$
- $P\left(a \mid S_{t}=x\right)=P\left(b \mid S_{t}=x\right)=0.5$
- $P\left(b \mid S_{t}=y\right)=1.0$

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: $P\left(S_{0}=x\right)=0.6$, 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., $P\left(S_{1} \mid O_{1}=b\right)$ and $P\left(S_{2} \mid O_{1}=b, O_{2}=a\right)$.
b) Show the work to compute the smoothed distribution over states at time step 1, i.e., $P\left(S_{1} \mid O_{1}=\right.$ $b, O_{2}=a$ ), 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. Note that your answer should be a sequence of 3 states.

