

**INSTRUCTIONS**

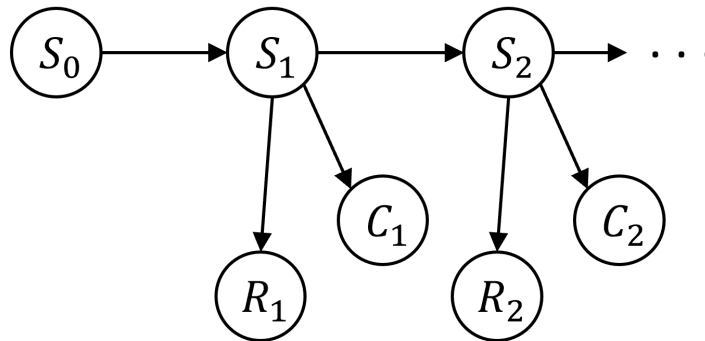
- **Due: Monday, November 24, 2025 at 10:00 PM EDT.** Remember that you may use up to 2 slip days for the written homework making the last day to submit **Wednesday, November 26, 2025 at 10:00 PM EDT.**
- **Format:** Write your answers in the `yoursolution.tex` file and compile a pdf (preferred) or you can type directly on the blank pdf. Make sure that your answers are within the dedicated regions for each question/part. If you do not follow this format, we may deduct points. Handwritten solutions are acceptable, but do not print out and scan the homework.
- **How to submit:** Submit a pdf with your answers on Gradescope. Log in and click on our class 15-281, click on the HW10 assignment, and upload your pdf containing your answers.
- **Policy:** See the course website for homework policies and academic integrity.

Name	
Andrew ID	
Hours to complete?	

# Q1. [62 pts] Dynamic Bayes Net and Hidden Markov Model

A professor wants to know if students are getting enough sleep. Each day, the professor observes whether the students sleep in class, and whether they have red eyes. Let  $S_t$  be the random variable of the student having enough sleep,  $R_t$  be the random variable for the student having red eyes, and  $C_t$  be the random variable of the student sleeping in class on day  $t$ . The professor has the following theory:

- The prior probability of getting enough sleep at time  $t$ , with no observations, is 0.6
- The probability of getting enough sleep on night  $t$  is 0.9 given that the student got enough sleep the previous night, and 0.2 if not
- The probability of having red eyes is 0.1 if the student got enough sleep, and 0.7 if not
- The probability of sleeping in class is 0.2 if the student got enough sleep, and 0.4 if not



$S_0$	$P(S_0)$	$S_{t+1}$	$S_t$	$P(S_{t+1}   S_t)$	$R_t$	$S_t$	$P(R_t   S_t)$	$C_t$	$S_t$	$P(C_t   S_t)$
$+s$	0.6	$+s_{t+1}$	$+s_t$	0.9	$+r$	$+s$	0.1	$+c$	$+s$	0.2
$-s$	0.4	$-s_{t+1}$	$+s_t$	0.1	$-r$	$+s$	0.9	$-c$	$+s$	0.8
		$+s_{t+1}$	$-s_t$	0.2	$+r$	$-s$	0.7	$+c$	$-s$	0.4
		$-s_{t+1}$	$-s_t$	0.8	$-r$	$-s$	0.3	$-c$	$-s$	0.6

Using the DBN above and these evidence values

- $[-r_1, -c_1]$  = not red eyes, not sleeping in class
- $[+r_2, -c_2]$  = red eyes, not sleeping in class
- $[+r_3, +c_3]$  = red eyes, sleeping in class

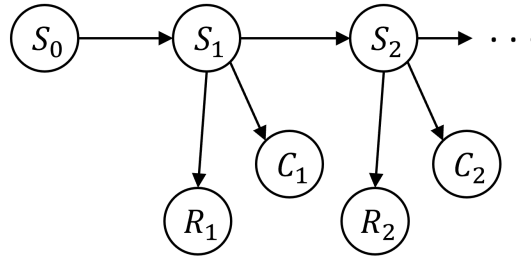
we want to compute  $P(S_t | r_{1:t}, c_{1:t})$  for each of  $t = 1, 2, 3$  as well as perform smoothing to get  $P(S_2 | r_{1:3}, c_{1:3})$ .

In order to do so, we will compute intermediate values which will correspond to the predict and update steps of our forward algorithm as well as finding the value of  $\alpha$  (**the normalization constant**) in each case.

**Note:** Please round your answers to 3 decimal places at the *end* of each calculation. That is, if you need to compute multiple intermediate values to get your answer, do not round until you get your final answer. Please also note we will only be able to award partial credit if work is shown.

**Also note:** You can (and should) check all your solutions via the Gradescope 'Online 10' assignment. This can serve as a kind of autograder. :)

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$S_0$	$P(S_0)$	$S_{t+1}$	$S_t$	$P(S_{t+1}   S_t)$	$R_t$	$S_t$	$P(R_t   S_t)$	$C_t$	$S_t$	$P(C_t   S_t)$
$+s$	0.6	$+s_{t+1}$	$+s_t$	0.9	$+r$	$+s$	0.1	$+c$	$+s$	0.2
$-s$	0.4	$-s_{t+1}$	$+s_t$	0.1	$-r$	$+s$	0.9	$-c$	$+s$	0.8
		$+s_{t+1}$	$-s_t$	0.2	$+r$	$-s$	0.7	$+c$	$-s$	0.4
		$-s_{t+1}$	$-s_t$	0.8	$-r$	$-s$	0.3	$-c$	$-s$	0.6

Round all numerical answers to 3 decimal places. Please also note we will only be able to award partial credit if work is shown. Evidence values:  $[-r_1, -c_1]$ ,  $[+r_2, -c_2]$ ,  $[+r_3, +c_3]$ . You have access to  $P(S_0)$ .

(a) [16 pts] State Estimation:  $t = 1$

(i) [6 pts] Predict:

Use the Bayes' Net's CPTs to write an equivalent expression for  $P(+s_1)$  :

$P(+s_1) =$

$P(-s_1) =$

(ii) [6 pts] Update:

Use the Bayes' Net's CPTs and  $\alpha$  to find an equivalent expression for  $P(+s_1 | -r_1, -c_1)$  :

When solving for probabilities, **don't** leave  $\alpha$  in your answer.

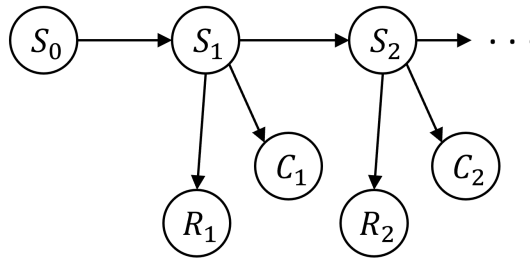
$P(+s_1 | -r_1, -c_1) =$

$P(-s_1 | -r_1, -c_1) =$

(iii) [4 pts] What was the value for  $\alpha$ ?

$\alpha =$

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$S_0$	$P(S_0)$	$S_{t+1}$	$S_t$	$P(S_{t+1}   S_t)$	$R_t$	$S_t$	$P(R_t   S_t)$	$C_t$	$S_t$	$P(C_t   S_t)$
$+s$	0.6	$+s_{t+1}$	$+s_t$	0.9	$+r$	$+s$	0.1	$+c$	$+s$	0.2
$-s$	0.4	$-s_{t+1}$	$+s_t$	0.1	$-r$	$+s$	0.9	$-c$	$+s$	0.8
		$+s_{t+1}$	$-s_t$	0.2	$+r$	$-s$	0.7	$+c$	$-s$	0.4
		$-s_{t+1}$	$-s_t$	0.8	$-r$	$-s$	0.3	$-c$	$-s$	0.6

Round all numerical answers to 3 decimal places. Please also note we will only be able to award partial credit if work is shown. Evidence values:  $[-r_1, -c_1]$ ,  $[+r_2, -c_2]$ ,  $[+r_3, +c_3]$

(b) [16 pts] State Estimation:  $t = 2$

(i) [6 pts] Predict:

$$P(+s_2 \mid -r_1, -c_1) =$$

$$P(-s_2 \mid -r_1, -c_1) =$$

(ii) [6 pts] Update:

Use the Bayes' Net's CPTs, previous probabilities, and  $\alpha$  to find an equivalent expression for  $P(+s_2 \mid r_{1:2}, c_{1:2})$ :

When solving for probabilities, **don't** leave  $\alpha$  in your answer.

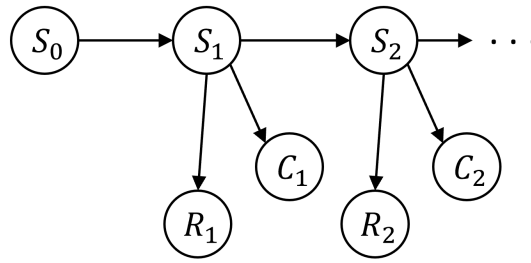
$$P(+s_2 \mid r_{1:2}, c_{1:2}) =$$

$$P(-s_2 \mid r_{1:2}, c_{1:2}) =$$

(iii) [4 pts] What was the value for  $\alpha$ ?

$$\alpha =$$

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$S_0$	$P(S_0)$	$S_{t+1}$	$S_t$	$P(S_{t+1}   S_t)$	$R_t$	$S_t$	$P(R_t   S_t)$	$C_t$	$S_t$	$P(C_t   S_t)$
$+s$	0.6	$+s_{t+1}$	$+s_t$	0.9	$+r$	$+s$	0.1	$+c$	$+s$	0.2
$-s$	0.4	$-s_{t+1}$	$+s_t$	0.1	$-r$	$+s$	0.9	$-c$	$+s$	0.8
		$+s_{t+1}$	$-s_t$	0.2	$+r$	$-s$	0.7	$+c$	$-s$	0.4
		$-s_{t+1}$	$-s_t$	0.8	$-r$	$-s$	0.3	$-c$	$-s$	0.6

Round all numerical answers to 3 decimal places. Please also note we will only be able to award partial credit if work is shown. Evidence values:  $[-r_1, -c_1]$ ,  $[+r_2, -c_2]$ ,  $[+r_3, +c_3]$

(c) [20 pts] State Estimation:  $t = 3$

(i) [8 pts] Predict:

$$P(+s_3 | r_{1:2}, c_{1:2}) =$$

$$P(-s_3 | r_{1:2}, c_{1:2}) =$$

(ii) [8 pts] Update:

Use the Bayes' Net's CPTs, previous probabilities, and  $\alpha$  to find an equivalent expression for  $P(+s_3 | r_{1:3}, c_{1:3})$ :

When solving for probabilities, **don't** leave  $\alpha$  in your answer.

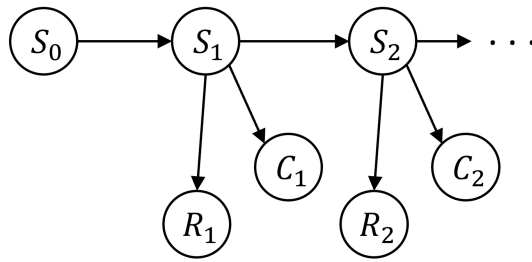
$$P(+s_3 | r_{1:3}, c_{1:3}) =$$

$$P(-s_3 | r_{1:3}, c_{1:3}) =$$

(iii) [4 pts] What was the value for  $\alpha$ ?

$$\alpha =$$

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$S_0$	$P(S_0)$	$S_{t+1}$	$S_t$	$P(S_{t+1}   S_t)$	$R_t$	$S_t$	$P(R_t   S_t)$	$C_t$	$S_t$	$P(C_t   S_t)$
$+s$	0.6	$+s_{t+1}$	$+s_t$	0.9	$+r$	$+s$	0.1	$+c$	$+s$	0.2
$-s$	0.4	$-s_{t+1}$	$+s_t$	0.1	$-r$	$+s$	0.9	$-c$	$+s$	0.8
		$+s_{t+1}$	$-s_t$	0.2	$+r$	$-s$	0.7	$+c$	$-s$	0.4
		$-s_{t+1}$	$-s_t$	0.8	$-r$	$-s$	0.3	$-c$	$-s$	0.6

Round all numerical answers to 3 decimal places. Please also note we will only be able to award partial credit if work is shown. Evidence values:  $[-r_1, -c_1]$ ,  $[+r_2, -c_2]$ ,  $[+r_3, +c_3]$

(d) [10 pts] We can build upon the previous three parts and use smoothing to compute  $P(S_2 | r_{1:3}, c_{1:3})$ .

(i) [4 pts] **Backward message:**  $P(+r_3, +c_3 | S_2) = \sum_{s_3} P(+r_3, +c_3 | s_3) P(s_3 | S_2)$

$$P(+r_3, +c_3 | +s_2) =$$

$$P(+r_3, +c_3 | -s_2) =$$

(ii) [4 pts] **Smoothing:**  $P(S_2 | r_{1:3}, c_{1:3}) = \alpha P(S_2 | r_{1:2}, c_{1:2}) P(+r_3, +c_3 | S_2)$   
When solving for probabilities, **don't** leave  $\alpha$  in your answer.

$$P(+s_2 | r_{1:3}, c_{1:3}) =$$

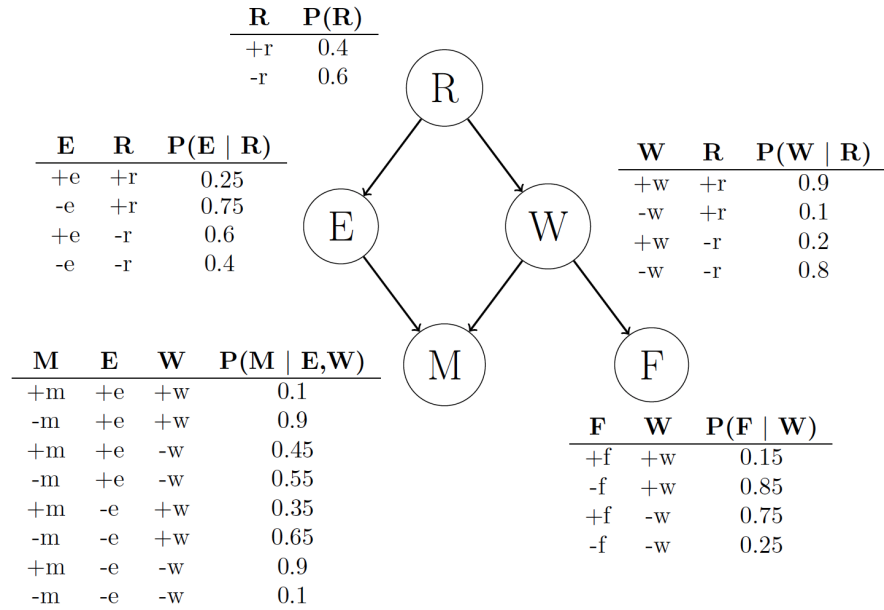
$$P(-s_2 | r_{1:3}, c_{1:3}) =$$

(iii) [2 pts] What was the value for  $\alpha$ ?

$$\alpha =$$

## Q2. [38 pts] Sampling

Consider the following Bayes Net and corresponding probability tables.



Consider the case where we are sampling to approximate the query  $P(R | +f, +m)$ .

- (a) [24 pts] Fill in the following table with the probabilities of *drawing* each respective sample given that we are using each of the following sampling techniques. Pay close attention to the domain of samples that can be drawn using the following methods. *Hint*:  $P(+f, +m) = 0.2682$ . *Hint 2*: probabilities can be 0.

Method	$\langle +r, +e, -w, +m, +f \rangle$	$\langle +r, -e, +w, -m, +f \rangle$
Prior sampling		
Rejection sampling		
Likelihood weighting		

- (b) [14 pts] We are going to use Gibbs sampling to estimate the probability of getting the sample  $\langle +r, +e, -w, +m, +f \rangle$ . We will start from the sample  $\langle -r, -e, -w, +m, +f \rangle$  and resample  $E$  first then  $R$ . What is the probability of drawing sample  $\langle +r, +e, -w, +m, +f \rangle$ ?

**Answer:**