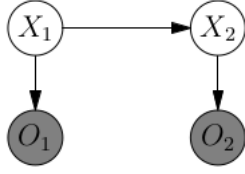


Consider the following Hidden Markov Model.



$X_1$   $\Pr(X_1)$	$X_t$   $X_{t+1}$   $\Pr(X_{t+1}   X_t)$	$X_t$   $O_t$   $\Pr(O_t   X_t)$
0   0.3	0   0   0.4	0   A   0.9
1   0.7	0   1   0.6	0   B   0.1
	1   0   0.8	1   A   0.5
	1   1   0.2	1   B   0.5

Use the Forward algorithm to compute the probability distribution  $\Pr(X_2, O_1 = A, O_2 = B)$ . Show your work. You do not need to evaluate arithmetic expressions involving only numbers.

**Solution.** We must compute  $\Pr(X_2, O_1 = A, O_2 = B)$  for each value of  $X_2$ . That is, we compute the two probabilities  $\Pr(X_2 = 0, O_1 = A, O_2 = B)$  and  $\Pr(X_2 = 1, O_1 = A, O_2 = B)$ . We do this by first computing  $\Pr(X_2 = 0, x_1, O_1 = A, O_2 = B)$  for each  $x_1 \in \{0, 1\}$  (the possible values for the random variable  $X_1$ ), and then we sum over  $x_1$  to get the marginal probability  $\Pr(X_2 = 0, O_1 = A, O_2 = B)$ .

$$\Pr(X_2, O_1 = A, O_2 = B) = \sum_{x_1} \Pr(X_2, x_1, O_1 = A, O_2 = B).$$

Recall, by the definition of conditional probability (or the chain rule), we have

$$\begin{aligned} \Pr(X_2, x_1, O_1 = A, O_2 = B) &= \Pr(x_1) \Pr(X_2, O_1 = A, O_2 = B | x_1) \\ &= \Pr(x_1) \Pr(O_1 = A | x_1) \Pr(X_2, O_2 = B | x_1, O_1 = A) \\ &= \Pr(x_1) \Pr(O_1 = A | x_1) \Pr(X_2 | x_1, O_1 = A) \Pr(O_2 = B | X_2, x_1, O_1 = A) \end{aligned}$$

Now we use the structure of the Bayes net to infer some conditional independences and simplify the above express. Specifically,  $O_2$  is independent of  $O_1$  given  $X_2$  and  $O_2$  is independent of  $X_1$  given  $X_2$ . Also,  $X_2$  is independent of  $O_1$  given  $X_1$ . Therefore,  $\Pr(X_2, x_1, O_1 = A, O_2 = B) = \Pr(x_1) \Pr(O_1 = A | x_1) \Pr(X_2 | x_1) \Pr(O_2 = B | X_2)$ . We now use the data in the tables above to compute

$$\begin{aligned} \Pr(X_2 = 0, X_1 = 0, O_1 = A, O_2 = B) &= \Pr(X_1 = 0) \Pr(O_1 = A | X_1 = 0) \Pr(X_2 = 0 | X_1 = 0) \Pr(O_2 = B | X_2 = 0) \\ &= 0.3 \cdot 0.9 \cdot 0.4 \cdot 0.1, \text{ and} \end{aligned}$$

$$\begin{aligned} \Pr(X_2 = 0, X_1 = 1, O_1 = A, O_2 = B) &= \Pr(X_1 = 1) \Pr(O_1 = A | X_1 = 1) \Pr(X_2 = 0 | X_1 = 1) \Pr(O_2 = B | X_2 = 0) \\ &= 0.7 \cdot 0.5 \cdot 0.8 \cdot 0.1 \end{aligned}$$

Therefore,  $\Pr(X_2 = 0, O_1 = A, O_2 = B) = 0.3 \cdot 0.9 \cdot 0.4 \cdot 0.1 + 0.7 \cdot 0.5 \cdot 0.8 \cdot 0.1 = 0.1 (0.3 \cdot 0.9 \cdot 0.4 + 0.7 \cdot 0.5 \cdot 0.8) = 0.1 (0.108 + 0.280) = 0.0388$ , Similarly,

$$\begin{aligned} \Pr(X_2 = 1, X_1 = 0, O_1 = A, O_2 = B) &= \Pr(X_1 = 0) \Pr(O_1 = A | X_1 = 0) \Pr(X_2 = 1 | X_1 = 0) \Pr(O_2 = B | X_2 = 1) \\ &= 0.3 \cdot 0.9 \cdot 0.6 \cdot 0.5, \text{ and} \end{aligned}$$

$$\begin{aligned} \Pr(X_2 = 1, X_1 = 1, O_1 = A, O_2 = B) &= \Pr(X_1 = 1) \Pr(O_1 = A | X_1 = 1) \Pr(X_2 = 1 | X_1 = 1) \Pr(O_2 = B | X_2 = 1) \\ &= 0.7 \cdot 0.5 \cdot 0.2 \cdot 0.5 \end{aligned}$$

Therefore,  $\Pr(X_2 = 1, O_1 = A, O_2 = B) = 0.3 \cdot 0.9 \cdot 0.6 \cdot 0.5 + 0.7 \cdot 0.5 \cdot 0.2 \cdot 0.5 = 0.5 (0.3 \cdot 0.9 \cdot 0.6 + 0.7 \cdot 0.5 \cdot 0.2) = 0.5 (0.162 + 0.07) = 0.1160$ , So the distribution  $\Pr(X_2, O_1 = A, O_2 = B)$  is given by the following table:

$X_2$   $\Pr(X_2, O_1 = A, O_2 = B)$
0   0.0388
1   0.1160