## Learning Objectives

- To practice computing HMM probabilities
- To run Predict and Observe steps in the forward algorithm

## Q1. Hidden Markov Models

Below is the Weather HMM from class. We want to predict the weather given both a distribution for weather today given weather yesterday (transition model) and weather today given people using umbrella today (sensor model). We want to compute the weather on day  $4 P(X_4|e_1, e_2, e_3, e_4)$  assuming that we've witnessed umbrellas each day (all the evidence is True that there is an umbrella). Note that the states are the weather  $X_i = W_i$  and the evidence is the umbrella  $e_i = U_i$ .

## An HMM is defined by:

■ Initial distribution:  $P(X_1)$ 

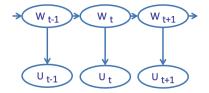
■ Transition model:  $P(X_t \mid X_{t-1})$ 

■ Sensor model:  $P(E_t \mid X_t)$ 

Given  $P(X_1) = \{sun: 0.5, rain: 0.5\}$ Compute  $P(X_4 = sun \mid e_4 = e_3 = e_2 = e_1 = True)$ 

| W <sub>t-1</sub> | P(W <sub>1</sub> | W <sub>t-1</sub> |
|------------------|------------------|------------------|
|                  | sun              | rain             |
| sun              | 0.9              | 0.1              |
| rain             | 0.3              | 0.7              |

| W <sub>t</sub> | P(U <sub>t</sub>  W <sub>t</sub> ) |       |
|----------------|------------------------------------|-------|
|                | true                               | false |
| sun            | 0.2                                | 0.8   |
| rain           | 0.9                                | 0.1   |



(a) OBSERVE! We are given the initial distribution  $P(X_1)$ . Now we need to Observe the evidence  $e_1 = True$  and compute  $P(X_1|e_1)$ . We can derive the equation for  $P(X_1|e_1)$  directly using Bayes Rule with the probabilities  $P(e_1|X_1)$  and  $P(X_1)$  or by computing the joint  $P(X_1,e_1)$  and normalizing  $Z = P(e_1)$ . Write the equation below and then compute the probability table  $P(X_1|e_1)$ .

$$P(X_1|e_1) = P(X_1, e_1)/P(e_1) = P(e_1|X_1)P(X_1)/\sum_x P(e_1|x)P(x)$$

Table 
$$P(X_1|e_1)$$
  
 $\frac{\sin | .1/(.1+.45) = .18}{ \text{rain } | .45/(.1+.45) = .82}$ 

(b) PREDICT! Now we have  $P(X_1|e_1)$ , and we want to Predict  $P(X_2|e_1)$ . We can do this by summing over  $X_1$ :  $P(X_2|e_1) = \sum_{x \in X_1} P(X_2|x)P(x|e_1)$ . Rewrite this equation below and then compute the probability table using your answer above and the HMM model tables.

$$P(X_2|e_1) = \sum_{x \in X_1} P(X_2|x)P(x|e_1)$$

Table 
$$P(X_2|e_1)$$
  
 $\sup | .9^*.18 + .3^*.82 = .41$   
 $rain | .1^*.18 + .7^*.82 = .59$ 

## An HMM is defined by:

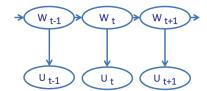
Initial distribution: P(X₁)
 Transition model: P(X₁ | X₁-1)

• Sensor model:  $P(E_t \mid X_t)$ 

Given  $P(X_1) = \{sun: 0.5, rain: 0.5\}$ Compute  $P(X_4=sun \mid e_4=e_3=e_2=e_1=True)$ 

| W <sub>t-1</sub> | P(W <sub>t</sub>  W <sub>t-1</sub> ) |      |
|------------------|--------------------------------------|------|
|                  | sun                                  | rain |
| sun              | 0.9                                  | 0.1  |
| rain             | 0.3                                  | 0.7  |

| W <sub>t</sub> | P(U <sub>t</sub>  W <sub>t</sub> ) |       |
|----------------|------------------------------------|-------|
|                | true                               | false |
| sun            | 0.2                                | 0.8   |
| rain           | 0.9                                | 0.1   |



(c) OBSERVE! Now that we've predicted  $X_2$ , we can update the probability given new evidence  $e_2 = True$ . Use the Observation update to write the formula for  $P(X_2|e_1, e_2)$  using  $P(X_2|e_1)$  above and then solve.

$$\begin{vmatrix} P(X_2|e_1, e_2) = \\ \alpha P(X_2, e_2|e_1) = P(e_2|X_2)P(X_2|e_1) / \sum_{x \in X_2} P(e_2|x)P(x|e_1) \end{vmatrix}$$

Table 
$$P(X_2|e_1, e_2)$$
  
 $\sup | .2^*.41/.613 = .13$   
 $rain | .9^*.59/.613 = .87$ 

(d) PREDICT! Compute  $P(X_3|e_1,e_2)$  using the transition probabilities and  $P(X_2|e_1,e_2)$  above. Write this equation below and then compute the probability table.

$$P(X_3|e_1, e_2) = \sum_{x \in X_2} P(X_3|x) P(x|e_1, e_2)$$

Table 
$$P(X_3|e_1, e_2)$$
  
 $\begin{array}{c|c} \text{sun} & .9^*.13 + .3^*.87 = .38 \\ \hline \text{rain} & .1^*.13 + .7^*.87 = .62 \end{array}$ 

(e) OBSERVE! Now that we've predicted  $X_3$ , we can update the probability given new evidence  $e_3 = True$ . Use the Observation update to write the formula for  $P(X_3|e_1,e_2,e_3)$  using  $P(X_3|e_1,e_2)$  above and then solve.

$$P(X_3|e_1, e_2, e_3) = \alpha P(X_3, e_3|e_1, e_2) = \alpha P(e_3|X_3) P(X_3|e_1, e_2)$$

$$\alpha = 1/\sum_{x \in X_3} P(e_3|x) P(x|e_1, e_2)$$

| <b>Table</b> $P(X_3 e_1, e_2, e_3)$ |                   |
|-------------------------------------|-------------------|
| sun                                 | .2*.38/.634 = .12 |
| rain                                | .9*.62/.634 = .88 |
|                                     |                   |

(f) PREDICT! Compute  $P(X_4|e_1, e_2, e_3)$  using the transition probabilities and  $P(X_3|e_1, e_2, e_3)$  above. Write the equation below and then compute the probability table.

$$P(X_4|e_1, e_2, e_3) = \sum_{x \in X_3} P(X_4|x)P(x|e_1, e_2, e_3)$$

Table 
$$P(X_4|e_1, e_2, e_3)$$
  
 $\sup | .9^*.12 + .3^*.88 = .37$   
 $rain | .1^*.12 + .7^*.88 = .63$ 

(g) OBSERVE! Finally, we can update the probability of  $X_4$  given new evidence  $e_4 = True$  (and the rest of the evidence). Use the Observation update rule to write the formula for  $P(X_4|e_1, e_2, e_3, e_4)$  using  $P(X_4|e_1, e_2, e_3)$  above and then solve for the new probability table.

$$|P(X_4|e_1, e_2, e_3, e_4) = \alpha P(X_4, e_4|e_1, e_2, e_3) = \alpha P(e_4|X_4) P(X_4|e_1, e_2, e_3)$$
  

$$\alpha = 1/\sum_{x \in X_4} P(e_4|x) P(x|e_1, e_2, e_3)$$

Table 
$$P(X_4|e_1, e_2, e_3, e_4)$$
  
 $\sup | .2^*.37/.641 = .115$   
 $rain | .9^*.63/.641 = .885$