CS 440: Introduction to Artificial Intelligence

Homework #3 Part 2 Probabilistic Reasoning

May 2, 2023

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Problem 1

Before we begin to tackle these problems, let's define the transition and observation matrices to make the calculations easier.

$$T = \begin{bmatrix} 0.2 & 0.8 & 0 & 0 & 0 & 0 \\ 0 & 0.2 & 0.8 & 0 & 0 & 0 \\ 0 & 0 & 0.2 & 0.8 & 0 & 0 \\ 0 & 0 & 0 & 0.2 & 0.8 & 0 \\ 0 & 0 & 0 & 0 & 0.2 & 0.8 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

1. Filtering

In order to find $P(X_3|hot_1, cold_2, cold_3)$, we can use the forward model:

$$P(X_3|hot_1, cold_2, cold_3) = O_{hot} * T^T * P(X_2|hot_1, cold_2)$$

= $O_{hot} * T^T * O_{cold} * T^T * P(X_1|hot_1)$

where $P(X_1|hot_1)$ is simply [1,0,0,0,0,0] as stated in the problem (because the rover is guaranteed to be on location A on the first day). So then,

$$O_{hot} * T^T * O_{cold} * T^T * [1, 0, 0, 0, 0, 0]^T = [0, 0.16, 0.64, 0, 0, 0]$$

which, after normalization, becomes

2. Smoothing

To find $P(X_2|hot_1, cold_2, cold_3)$, we can split up the calculation according into a backward model and a forward model to get the smoothed probability.

$$P(X_2|hot_1, cold_2, cold_3) = b_3 * f_{1:2}$$

where

$$b_3 = T * O_{cold} * b_4$$

= $T * O_{cold} * [1, 1, 1, 1, 1, 1]^T$
= $[0.8, 1, 0.2, 0.8, 1, 0]$

and

$$f_{1:2} = O_{cold} * T^T * P(X_1|hot_1)$$

= [0, 0.8, 0, 0, 0, 0]

which makes

$$P(X_2|hot_1, cold_2, cold_3) = b_3 * f_{1:2}$$

$$= [0.8, 1, 0.2, 0.8, 1, 0] * [0, 0.8, 0, 0, 0, 0]$$

$$= [0, 0.8, 0, 0, 0, 0].$$

After normalization, this becomes

$$[0, 1, 0, 0, 0, 0]$$
.

4. Prediction

In order to find the distribution $P(X_4|hot_1, cold_2, cold_3)$, we can simply split this up into:

$$P(X_4|hot_1, cold_2, cold_3) = T^T * P(X_3|hot_1, cold_2, cold_3)$$
$$= T^T * [0, 0.2, 0.8, 0, 0, 0]^T$$
$$= [0, 0.04, 0.32, 0.64, 0, 0].$$

We just used the normalized $P(X_3|hot_1, cold_2, cold_3)$ so as to bypass normalizing this new distribution again.

Problem 2

1. Bellman Equation

The Bellman equation for this particular MDP would be:

$$V^{\pi}(s) = R(s, \pi(s)) + \sum_{s' \in \{A, B\}} T(s, \pi(s), s') V^{\pi}(s')$$

2. Policy Evaluation

Using the initial policy rules $\pi(A) = 1$ and $\pi(B) = 1$ with initial policy values $V_0^{\pi}(A) = V_0^{\pi}(A) = 0$, two iterations of policy evaluation would look like this:

1. Iteration 1:

$$A: V_1^{\pi}(A) = R(A, \pi_0(A)) + [T(A, \pi_0(A), A)V_0^{\pi}(A) + T(A, \pi_0(A), B)V_0^{\pi}(B)]$$

$$= 0 + [0 + 0]$$

$$= 0$$

$$B: V_1^{\pi}(B) = R(B, \pi_0(B)) + [T(B, \pi_0(B), A)V_0^{\pi}(A) + T(B, \pi_0(B), B)V_0^{\pi}(B)]$$

$$= 5 + [0 + 0]$$

$$= 5$$

2. Iteration 2:

$$\begin{split} A: V_2^\pi(A) &= R(A, \pi_0(A)) + [T(A, \pi_0(A), A)V_1^\pi(A) + T(A, \pi_0(A), B)V_1^\pi(B)] \\ &= 0 + [0+0] \\ &= 0 \\ B: V_2^\pi(B) &= R(B, \pi_0(B)) + [T(B, \pi_0(B), A)V_1^\pi(A) + T(B, \pi_0(B), B)V_1^\pi(B)] \\ &= 5 + [0+5] \\ &= 10 \end{split}$$

So the updated values are $V^{\pi}(A) = 0$ and $V^{\pi}(A) = 10$.

3. Improved Policy

We can find π_{new} using the policy improvement equation:

$$\begin{split} \pi_{new}(A) &= \underset{a \in \{1,2\}}{\operatorname{argmax}} [R(A,a) + (T(A,a,A)V_2(A) + T(A,a,B)V_2(B))] \\ &= 2 \\ \pi_{new}(B) &= \underset{a \in \{1,2\}}{\operatorname{argmax}} [R(B,a) + (T(B,a,A)V_2(A) + T(B,a,B)V_2(B))] \\ &= 1 \end{split}$$

The work is shown below.

•
$$\pi_{new}(A)$$

- a=1: 0 + 1 * 0 + 0 * 10 = 0
- a=2: -1 + 0.5 * 0 + 0.5 * 10 = 4

•
$$\pi_{new}(B)$$

- a=1: 5 + 0 + 1 * 10 = 15
- a=2: 0 + 0 + 1 * 10 = 10