

# Car

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Environment: Mac OS

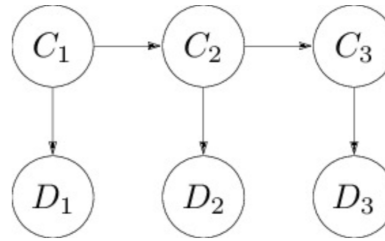
## Problem 1: Warmup

### 1a. Question

Suppose we have a sensor reading for the second timestep,  $D_2 = 0$ . Compute the posterior distribution  $P(C_2 = 1 \mid D_2 = 0)$ .

### 1a. Answer

Here's what the Bayesian network (it's an HMM, in fact) looks like:



So the posterior distribution can be computed like following:

$$\begin{aligned} P(C_2 = 1 \mid D_2 = 0) &= \frac{P(C_2 = 1, D_2 = 0)}{P(D_2 = 0)} \\ &= \frac{P(D_2 = 0 \mid C_2 = 1) * P(C_2 = 1)}{P(D_2 = 0)} \\ &= \frac{P(D_2 = 0 \mid C_2 = 1) * P(C_2 = 1)}{\sum_{C_2} P(D_2 = 0 \mid C_2) P(C_2)} \\ &= \frac{P(D_2 = 0 \mid C_2 = 1) * [\sum_{C_1} P(C_2 = 1 \mid C_1) P(C_1)]}{\sum_{C_2} P(D_2 = 0 \mid C_2) P(C_2)} \\ &= \frac{\eta * [\epsilon * 0.5 + (1 - \epsilon) * 0.5]}{(1 - \eta) * [\epsilon * 0.5 + (1 - \epsilon) * 0.5] + \eta * [\epsilon * 0.5 + (1 - \epsilon) * 0.5]} \\ &= \frac{\eta * 0.5}{0.5} \\ &= \eta \end{aligned}$$

### 1b. Question

Suppose a time step has elapsed and we got another sensor reading  $D_3 = 1$ , but we are still interested in  $C_2$ . Compute the posterior distribution  $P(C_2 = 1 \mid D_2 = 0, D_3 = 1)$ .


### 1b. Answer

$$\begin{aligned}
P(C_2 \mid D_2 = 0, D_3 = 1) &\propto \sum_{C_1, C_3} P(C_1)P(C_2 \mid C_1)P(D_2 = 0 \mid C_2)P(C_3 \mid C_2)P(D_3 = 1 \mid C_3) \\
&\propto P(D_2 = 0 \mid C_2) \left[ \sum_{C_1} P(C_1)P(C_2 \mid C_1) \right] \left[ \sum_{C_3} P(C_3 \mid C_2)P(D_3 = 1 \mid C_3) \right]
\end{aligned}$$

$$\begin{aligned}
P(C_2 = 1 \mid D_2 = 0, D_3 = 1) &\propto \eta [0.5(\epsilon + 1 - \epsilon)] [(1 - \epsilon)(1 - \eta) + \epsilon\eta] \\
P(C_2 = 0 \mid D_2 = 0, D_3 = 1) &\propto (1 - \eta) [0.5(\epsilon + 1 - \epsilon)] [(1 - \epsilon)\eta + \epsilon(1 - \eta)]
\end{aligned}$$

$$\begin{aligned}
P(C_2 = 1 \mid D_2 = 0, D_3 = 1) &= \frac{P(C_2 = 1 \mid D_2 = 0, D_3 = 1)}{P(C_2 = 1 \mid D_2 = 0, D_3 = 1) + P(C_2 = 0 \mid D_2 = 0, D_3 = 1)} \\
&= \frac{\eta [(1 - \epsilon)(1 - \eta) + \epsilon\eta]}{\eta [(1 - \epsilon)(1 - \eta) + \epsilon\eta] + (1 - \eta) [(1 - \epsilon)\eta + \epsilon(1 - \eta)]}
\end{aligned}$$

## 1c. Question

c.  [3 points] Suppose  $\epsilon = 0.1$  and  $\eta = 0.2$ .

i. Compute and compare the probabilities  $\mathbb{P}(C_2 = 1 \mid D_2 = 0)$  and  $\mathbb{P}(C_2 = 1 \mid D_2 = 0, D_3 = 1)$ . Give numbers, round your answer to 4 significant digits.

ii. How did adding the second sensor reading  $D_3 = 1$  change the result? Explain your intuition in terms of the car positions with respect to the observations.

iii. What would you have to set  $\epsilon$  while keeping  $\eta = 0.2$  so that  $\mathbb{P}(C_2 = 1 \mid D_2 = 0) = \mathbb{P}(C_2 = 1 \mid D_2 = 0, D_3 = 1)$ ? Explain your intuition in terms of the car positions with respect to the observations.

## 1c. Answer

i.

$$\begin{aligned}
P(C_2 = 1 \mid D_2 = 0) &= \eta = 0.2000 \\
P(C_2 = 1 \mid D_2 = 0, D_3 = 1) &= 0.4157
\end{aligned}$$

ii.

From our results, we know  $P(C_2 = 1 \mid D_2 = 0) < P(C_2 = 1 \mid D_2 = 0, D_3 = 1)$ . We can draw the conclusion that adding the second sensor  $D_3 = 1$  increased the probability of  $P(C_2) = 1$ . The position of a car observed at time step  $t$  is also related to the position of the car at time step  $t + 1$ . So the observation of  $D_3 = 1$  increased the probability of  $C_2 = 1$ .

iii.

Set  $\epsilon = 0$ . Hence  $P(C_t \mid C_{t-1}) = 1$ , when  $C_t = C_{t-1}$ . Thus the value of  $D_3$  is only related to  $D_3$  with parameter  $\eta$ , and we don't need to consider the transition probability of  $P(C_3 \mid C_2)$ .

You can also derive the above conclusion from the formula.

## PROBLEM 2: Emission probabilities

### 2a. Check my Code.

## **PROBLEM 3: Transition probabilities**

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**3a. Check my Code.**

## **PROBLEM 4: LEARNING TO PLAY BLACKJACK**

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**4a. Check my Code.**