

Project 4 Car Tracking

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Problem 1: Bayesian Network Basics

(a) We have:

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

(b) Since

$$\begin{aligned} P(C_2|D_2 = 0, D_3 = 1) &\propto P(C_2, D_2 = 0, D_3 = 1) \\ &\propto \sum_{C_1} P(C_2|D_2 = 0) P(C_3 = c_3|C_2) P(D_3 = 1|C_3 = c_3) \\ &= \alpha < 1 - \eta, \eta > < \epsilon + \eta - 2\epsilon\eta, 1 - \epsilon - \eta + 2\epsilon\eta > \\ &= \alpha < \epsilon + \eta - 3\epsilon\eta - \eta^2 + 2\epsilon\eta^2, \eta - \epsilon\eta - \eta^2 + 2\epsilon\eta^2 > \end{aligned}$$

After normalization, we can get

$$\alpha = \frac{1}{\epsilon + 2\eta - 4\epsilon\eta - 2\eta^2 + 4\epsilon\eta^2}$$

So,

$$P(C_2 = 1|D_2 = 0, D_3 = 1) = \frac{\eta - \epsilon\eta - \eta^2 + 2\epsilon\eta^2}{\epsilon + 2\eta - 4\epsilon\eta - 2\eta^2 + 4\epsilon\eta^2}$$

(c) Suppose $\epsilon = 0.1$ and $\eta = 0.2$

i. After put the value into the formula above, we can get

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

ii. After adding the second sensor reading $D_3 = 1$, the posterior probability of $P(C_2 = 1)$ increase. That's because the value of C_2 is related to the values of C_3 and D_3 , when $D_3 = 1$ is more likely to occur under $C_3 = 1$, the posterior probability of $C_3 = 1$ rises when $D_3 = 1$ is observed.

iii. While we set $\eta = 0.2$, in order to keep $P(C_2 = 1|D_2 = 0, D_3 = 1) = P(C_2 = 1|D_2 = 0, D_3 = 1)$, So,

$$\begin{aligned} \eta &= \frac{\eta - \epsilon\eta - \eta^2 + 2\epsilon\eta^2}{\epsilon + 2\eta - 4\epsilon\eta - 2\eta^2 + 4\epsilon\eta^2} \\ \epsilon &= 0.5 \end{aligned}$$