

Report of Car Project

Shun Zhang, 15300180012

Problem 1:

(a) To derive $P(C_2 = 1|D_2 = 0)$

According to Baye's formula, we have

$$P(C_2 = 1|D_2 = 0) = \alpha P(D_2 = 1|C_2 = 0)P(C_2 = 1)$$

where

$$P(C_2 = 1) = P(C_2 = 1|C_1 = 0)P(C_1 = 0) + P(C_2 = 1|C_1 = 1)P(C_1 = 1) = \frac{1}{2}\epsilon + \frac{1}{2}(1 - \epsilon) = \frac{1}{2}$$

so we have

$$P(C_2 = 1|D_2 = 0) = \alpha \frac{1}{2} P(D_2 = 0|C_2 = 1) = \alpha \frac{1}{2} \eta$$

Similarly, we have

$$P(C_2 = 0|D_2 = 0) = \alpha \frac{1}{2} P(D_2 = 0|C_2 = 0) = \alpha \frac{1}{2} (1 - \eta)$$

Which means $\alpha = 2$ and then

$$P(C_2 = 1|D_2 = 0) = \alpha \frac{1}{2} P(D_2 = 0|C_2 = 1) = \eta$$

(b) To derive $P(C_2 = 1|D_2 = 0, D_3 = 1)$

Again, according to Baye's formula and the conditional independence within this network, we have

$$\begin{aligned}
P(C_2 = 1|D_2 = 0, D_3 = 1) &= \alpha P(D_2 = 0, D_3 = 1|C_2 = 1)P(C_2 = 1) \\
&= \frac{\alpha}{2} P(D_2 = 0|C_2 = 1)P(D_3 = 1|C_2 = 1) \\
&= \frac{\alpha}{2} \eta [P(D_3 = 1, C_3 = 0|C_2 = 1) + P(D_3 = 1, C_3 = 1|C_2 = 1)] \\
&= \frac{\alpha}{2} \eta [\eta\epsilon + (1 - \eta)(1 - \epsilon)] \\
P(C_2 = 0|D_2 = 0, D_3 = 1) &= \alpha P(D_2 = 0, D_3 = 1|C_2 = 0)P(C_2 = 0) \\
&= \frac{\alpha}{2} P(D_2 = 0|C_2 = 0)P(D_3 = 1|C_2 = 0) \\
&= \frac{\alpha}{2} (1 - \eta) [P(D_3 = 1, C_3 = 0|C_2 = 0) + P(D_3 = 1, C_3 = 1|C_2 = 0)] \\
&= \frac{\alpha}{2} (1 - \eta) [\eta(1 - \epsilon) + (1 - \eta)\epsilon] \\
\Rightarrow \frac{\alpha}{2} &= \frac{1}{4\eta^2\epsilon - 2\eta^2 - 4\eta\epsilon + 2\eta + \epsilon} \\
\Rightarrow P(C_2 = 1|D_2 = 0, D_3 = 1) &= \frac{\eta[\eta\epsilon + (1 - \eta)(1 - \epsilon)]}{4\eta^2\epsilon - 2\eta^2 - 4\eta\epsilon + 2\eta + \epsilon}
\end{aligned}$$

(c)

- (i) With $\epsilon = 0.1$ and $\eta = 0.2$, we have

$$P(C_2 = 1|D_2 = 0) = \eta = 0.2$$

and

$$P(C_2 = 1|D_2 = 0, D_3 = 1) = \eta\epsilon + (1 - \eta)(1 - \epsilon) = 0.4157$$

- (ii) From my intuition, because of the high possibility of $P(C_t|D_t)$, $C_t = D_t$ and $P(C_t|C_{t-1})$, which means that the signal given by observation D_t are quite assuring. So, a positive observation like $D_3 = 1$ means that there is a higher possibility that $C_3 = 1$ and also a higher possibility that $C_2 = 1$. This intuition is further confirmed by the results above.
- (iii) Mathematically, we let the two items equal and we have

$$\frac{0.2[0.2\epsilon + 0.8(1 - \epsilon)]}{0.16\epsilon - 0.08 - 0.8\epsilon + 0.4 + \epsilon} = 0.2 \Rightarrow \epsilon = 0.5$$

which means that the the car will either change its position or stay for an equal probability, at every time step t . This means the information of C_3 is useless while inferring the information of C_2 and so is the information of observation D_3 . Consequently, $P(C_2 = 1|D_2 = 0) = P(C_2 = 1|D_2 = 0, D_3 = 1)$.

The other parts shall refer to [submission.py](#).