# CS221 Fall 2018 Homework [car]

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By turning in this assignment, I agree by the Stanford honor code and declare that all of this is my own work.

## Problem 1

(a) Step1: Remove variables that are not ancestors

Step2: Converting to factor graph

Step3: Conditioning on  $D_2 = 0$ 

Condition variable  $D_2$  on value  $D_2 = 0$ , replacing it with a factor  $\operatorname{cond}_{D_2=0}(C_2)$ , we get

$$\begin{array}{ccc} \operatorname{cond}_{D_2=0}(C_2) & C_2 \\ 1 - \eta & 0 \\ \eta & 1 \end{array}$$

Therefore, from the above table,  $p(C_2 = 1/D_2 = 0) = \eta$ 

(b) Step1: Remove variables that are not ancestors

Step2: Converting to factor graph

Step3: Conditioning on  $D_3 = 1$ 

Conditioning on variable  $D_3$ , and replacing it with a factor  $\operatorname{cond}_{D_3=1}(C_3)$ , we get

$$\begin{array}{ll}
\operatorname{cond}_{D_3=1}(C_3) & C_3 \\
\eta & 0 \\
1-\eta & 1
\end{array}$$

### Step4: Eliminating $C_3$

Defining function  $elim_{C_3}(C_2)$  in order to eliminate node  $C_3$  as

$$\operatorname{elim}_{C_3}(C_2) = \sum_{C_3} \operatorname{cond}_{D_3=1}(C_3) p(C_3/C_2)$$

The probability distribution  $p(C_3/C_2)$  is given by:

C2 C3 
$$p(C3/C2)$$
  
0 0 1  $-\epsilon$   
0 1  $\epsilon$   
1 0  $\epsilon$   
1 1  $1-\epsilon$ 

The probability distribution  $\operatorname{cond}_{D_3=1}(C_3)$  is defined in Step 3.

Combining both and substituting in equation 1, and doing summation over values of  $C_3$ , we will have probability distribution of  $\operatorname{elim}_{C_3}(C_2)$  is given by:

$$C_2$$
 elim $_{C_3}(C_2)$   
 $0$   $(1 - \epsilon)\eta + \epsilon(1 - \eta)$   
 $1$   $\epsilon \eta + (1 - \eta)(1 - \epsilon)$ 

#### Step5: Combining all factors of $C_2$

The other distribution which depends on is  $p(D_2 = 1/C_2)$ , which can be conditioned as  $\operatorname{cond}_{D_2=0}(C_2)$ , given by:

$$C_2 \quad \operatorname{cond}_{D_2=0}(C_2)$$

$$0 \quad 1-\eta$$

$$1 \quad \eta$$

Multiplying  $\operatorname{elim}_{C_3}(C_2)$  and  $\operatorname{cond}_{D_2=0}(C_2)$ :

$$C_2 \quad \text{elim}_{C_3}(C_2)$$

$$0 \quad ((1 - \epsilon)\eta + \eta(1 - \epsilon))(1 - \eta)$$

$$1 \quad (\epsilon \eta + (1 - \eta)(1 - \epsilon))\eta$$

Therefore,

$$P(C_2 = 1/D_2 = 0, D_3 = 1) = \frac{(\epsilon \eta + (1 - \eta)(1 - \epsilon))\eta}{(\epsilon \eta + (1 - \eta)(1 - \epsilon))\eta + ((1 - \epsilon)\eta + \epsilon(1 - \eta))(1 - \eta)}$$

(c) i.

$$P(C_2 = 1/D_2 = 0) = 0.2$$
  
 $P(C_2 = 1/D_2 = 0, D_3 = 1) = 0.4157$ 

- ii. Adding second sensor reading increased the probability from 0.2 to 0.4157. Since  $D_3$  is equal to 1, it means we observed the location to be 1 at location 3. This would increase the probability of  $C_3 = 1$  since the emission probability  $p(d_t/c_t)$  favours similar values with higher probability.  $C_3 = 1$  increases the probability of  $C_2 = 1$ , since the transition probability  $p(c_t/c_{t-1})$  favours same location with higher probability.
- iii. Both the probabilities would be same when the sensor reading at  $D_3$  doesn't matter. This won't matter when the transition probabilities  $p(c_t/c_{t-1})$  are equal meaning no matter what is the value of  $c_3$  out of all the possible values, we will get constant transition probability. This would happen when  $\epsilon = 1 \epsilon$ , therefore when  $\epsilon = 0.5$ .

# Problem 2

- (a) (your solution)
- (b) (your solution)