Machine Learning: Data to Models Assignment 1b: Bayesian Network

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2.3 Parameter Estimation [20 points] a. [5 points]

If X is binary, given its parents Pa(X), we can get the parameters $\theta_{X^0|Pa(X)}$ and $\theta_{X^1|Pa(X)}$. Since $\theta_{X^0|Pa(X)} + \theta_{X^1|Pa(X)} = 1$, so we only need to estimate one of them. In this case, the parameters of this network are: $\theta_{a^0}, \theta_{b^0|a^0}, \theta_{b^0|a^1}, \theta_{c^0|a^0,b^0}, \theta_{c^0|a^0,b^1}, \theta_{c^0|a^1,b^0}, \theta_{c^0|a^1,b^1}, \theta_{d^0|b^0}, \theta_{d^0|b^1}, \theta_{d^0|b^0,c^0}, \theta_{e^0|b^0,c^1}, \theta_{e^0|b^1,c^0}, \theta_{e^0|b^1,c^1}, \theta_{e^0|b^1,c^1}, 13$ parameters in total.

b. [15 points]

Algorithm 1 Initialization for EM algorithm

- 1: **procedure** INITIALIZE
- 2: //G: Bayesian network with nodes X_A, X_B, X_C, X_D, X_E
- 3: $//\theta^0$: randomly set parameters in (a)
- 4: //D: Partial observed data in the Table 1 of assignment 1b handout.
- 5: $//\epsilon$: convergence threshold
- 6: **for** $i \in \{A, B, C, D, E\}$ **do**
- 7: **for** each $x_i, u_i \in Val(X_i, Pa_{X_i}^{\mathcal{G}})$ **do**
- 8: $\bar{M}_{\theta^0}[x_i, u_i] = 0;$

Algorithm 2 EM Algorithm for learning parameters in Bayes Network

```
1: procedure EM-ALGORITHM
 2:
          INITIALIZE;
          isConverge = False;
 3:
 4:
          \epsilon = one small number;
          t=0;
 5:
          while isConverge == False do
 6:
               //E-Step (Compute the expected sufficient statistics)
 7:
               for m = 1...n do //n is sample size
 8:
 9:
                    Run inference on \langle \mathcal{G}, \theta \rangle using evidence o[m];
                    for i \in \{A, B, C, D, E\} do
10:
                        for x_i, u_i \in Val(X_i, Pa_{X_i}^{\mathcal{G}}) do
11:
                              \bar{M}_{\theta^t}[x_i, u_i] = \bar{M}_{\theta^t}[x_i, u_i] + P(x_i, u_i \mid o[m], \theta^t);
12:
13:
               //M-Step
               for i \in \{A, B, C, D, E\} do
14:
                    for x_i, u_i \in Val(X_i, Pa_{X_i}^{\mathcal{G}}) do
15:
                        \#Val(X_i, Pa_{X_i}^{\mathcal{G}}) finds all the parents of X_i and
16:
                        // return a set of paired values (X_i, \text{ parent of } X_i) \theta_{x_i|u_i}^{t+1} = \frac{\bar{M}_{\theta^t}[x_i,u_i]}{\bar{M}_{\theta^t}[u_i]}
17:
18:
               //Check convergence
19:
               if the difference between every parameter at t and t+1 < \epsilon then
20:
                    isConvergence = True;
21:
                    return \theta^{t+1}
22:
23:
               t = t + 1;
```