HW 8

Adam Karl

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1 Bayesian Belief Networks

a.
$$P(B = T, E = T) =$$

$$\sum_{a \in T, F} \sum_{d \in T, F, X} \sum_{c \in T, F} \sum_{f \in T, F} P(A = a) P(B = T) P(C = c) P(D = d | A = a, B = T, C = c) P(E = T | C = c) P(F = f | D = d)$$

There are 4 binary variables and 1 tertiary variable, so the number of parameters for the full joint is $2^4(3^1) = 48$. There are 23 addends and 120 multiplicands for this calculation. b. We can group together the variables that depend on each other.

$$\sum_{f \in T, F} P(F = f | D = d) \sum_{d \in T, F, X} P(D = d | A = a, B = T, C = c) P(B = T)$$

$$[\sum_{a \in T, F} P(A = a)] P(E = T | C = c) [\sum_{c \in T, F} P(C = c)]$$

This leaves us with 25 addends and 8 multiplicands, which is more efficient for addition but less efficient for multiplication.

2 Pneumonia Diagnosis

a.

- pneumonia = unknown
 - P(fever) = 0.606
 - P(paleness) = 0.504
 - P(cough) = 0.116
 - P(highWBcount) = 0.506
 - P(pneumonia) = 0.02
- $\bullet \ pneumonia = true$
 - P(fever) = 0.9
 - P(paleness) = 0.7
 - P(cough) = 0.9
 - P(highWBcount) = 0.8
- pneumonia = false
 - P(fever) = 0.6
 - P(paleness) = 0.5
 - P(cough) = 0.1
 - P(highWBcount) = 0.5

b.

$$\frac{P(fever = T|pneu = T)P(paleness = F|pneu = T)P(cough = T|pneu = T)P(highWBcount = F|pneu = T)P(pneu = T)P($$

$$\frac{0.9(0.3)(0.9)(0.2)(0.02)}{0.606(0.496)(0.116)(0.494)} = 0.0564$$

Therefore, there is a 5.64% chance the patient has pneumonia.

c.

$$\frac{P(fever = T|pneu = T)P(cough = T|pneu = T)P(pneu = t)}{P(fever = T)P(cough = T)}$$

$$\frac{0.9(0.9)(0.02)}{(0.606)(0.116)} = 0.2304$$

Therefore, there is a 23.04% chance the patient has pneumonia.