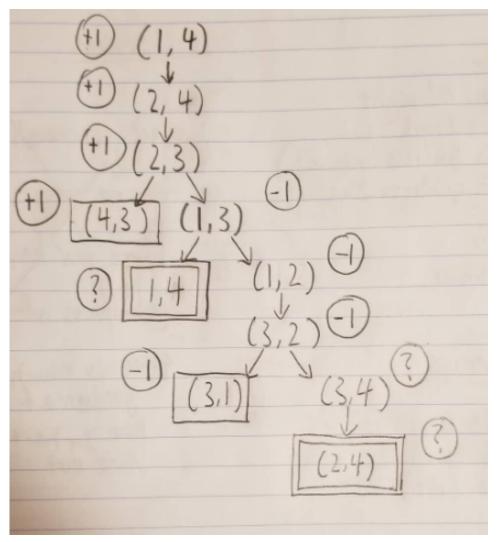
Adversarial Search -Bayesian Networks

Assignment 3

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Problem 1



"?" values are used to annotate loop states. When an agent has to choose between a +1 value and a "?" value (+1,?) the max player will always choose +1 while the min player will choose ?, and vice versa for (-1,?), with the min player choosing -1 and the max player choosing "?". Since the game value of a loop state is unknown, it is best for a max/min player to choose the corresponding max/min value if it is available, however if that value is not available the "?" state is a viable option to explore as it cannot be worse than an immediate loss. If all the successors of a state have a "?" value, the backed-up value is also "?".

Problem 2

(a) Computing the joint probability distribution:

$$P(a, b, c, d, e) = P(a) * P(b) * P(c) * P(d|a, b) * P(e|b, c)$$

$$P(a, b, c, d, e) = 0.2 * 0.5 * 0.8 * 0.1 * 0.3$$

$$P(a, b, c, d, e) = 0.0024$$

(b) Computing the joint probability distribution:

$$P(\neg a, \neg b, \neg c, \neg d, \neg e) = P(\neg a) * P(\neg b) * P(\neg c) * P(\neg d | \neg a, \neg b) * P(\neg e | \neg b, \neg c)$$

$$P(\neg a, \neg b, \neg c, \neg d, \neg e) = 0.8 * 0.5 * 0.2 * 0.1 * 0.8$$

$$P(\neg a, \neg b, \neg c, \neg d, \neg e) = 0.0064$$

(c) Using the conditional probability rule $P(A|B) = \frac{P(A \cap B)}{P(B)}$

$$P(\neg a|b,c,d,e) = \frac{P(\neg a,b,c,d,e)}{P(b,c,d,e)}$$

$$P(\neg a|b,c,d,e) = \frac{P(\neg a,b,c,d,e)}{P(a,b,c,d,e) + P(\neg a,b,c,d,e)}$$

$$P(\neg a|b,c,d,e) = \frac{P(\neg a) * P(b) * P(c) * P(d|\neg a,b) * P(e|b,c)}{0.0024 + P(\neg a) * P(b) * P(c) * P(d|\neg a,b) * P(e|b,c)}$$

$$P(\neg a|b,c,d,e) = \frac{0.8 * 0.5 * 0.8 * 0.6 * 0.3}{0.0024 + 0.8 * 0.5 * 0.8 * 0.6 * 0.3}$$

$$P(\neg a|b,c,d,e) = \frac{0.0576}{0.0024 + 0.0576}$$

$$P(\neg a|b,c,d,e) = 0.96$$

Problem 3

(a)

$$P(b|j,m) = \alpha * P(b,j,m)$$

$$\alpha = \frac{1}{P(\neg b, j, m) + P(b, j, m)}$$

$$P(b,j,m) = \sum_{e} \sum_{a} P(b,j,m,e,a)$$

$$P(b,j,m) = P(b) \sum_{e} P(e) \sum_{a} P(a|b,e) P(j|a) P(m|a)$$

$$P(b,j,m) = 0.001 * 0.002 * 0.95 * 0.9 * 0.7 + 0.001 * 0.002 * 0.05 * 0.05 * 0.01 + 0.001 * 0.998 * 0.94 * 0.9 * 0.7 + 0.001 * 0.998 * 0.06 * 0.05 * 0.01$$

$$P(b,j,m) = 0.00059224259$$

$$\begin{split} P(\neg b,j,m) &= \sum_e \sum_a P(\neg b,j,m,e,a) \\ P(\neg b,j,m) &= P(\neg b) \sum_e P(e) \sum_a P(a|\neg b,e) P(j|a) P(m|a) \\ P(\neg b,j,m) &= 0.999*0.002*0.29*0.9*0.7+0.999*0.002*0.71*0.05*0.01+\\ 0.999*0.998*0.001*0.9*0.7+0.999*0.998*0.999*0.05*0.01\\ P(\neg b,j,m) &= 0.00149185764 \end{split}$$

$$\alpha = \frac{1}{P(\neg b, j, m) + P(b, j, m)}$$

$$\alpha = \frac{1}{0.00149185764 + 0.00059224259}$$

$$\alpha = \frac{1}{0.00208410023}$$

$$P(b|j,m) = \alpha * P(b,j,m)$$

$$P(b|j,m) = \frac{0.00059224259}{0.00208410023}$$

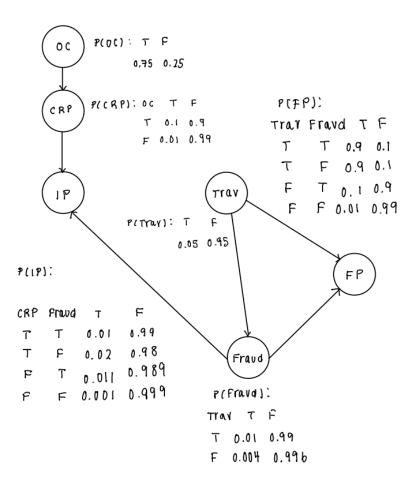
$$P(b|j,m) = 0.28417183659$$

Therefore, the probability of a burglary given that John and Mary both called, is 0.28417183659.

(b) In a Bayesian network in the form of a chain, using the enumeration method has a worst case runtime of $\mathcal{O}(2^n)$, as moving down each level in the chain generates another level in a binary tree.

Using the variable elimination method, moving down each level in the Bayesian network chain only adds one additional problem to compute (point-wise product of factors), as each node in the network only has one parent. Thus, the variable elimination method has a worst case runtime of $\mathcal{O}(n)$.

Problem 4



(a)

(b) Probability the current transanction is a fraud prior to searching for previous computer related purchases and before verifying if it is a foreign and/or an internet purchase:

$$P(Fraud) = \\ P(Fraud = True|Trav = True) * P(Trav = True) + \\ P(Fraud = True|Trav = False) * P(Trav = False) = \\ 0.01 * 0.05 + 0.004 * 0.95 = \\ 0.0043$$

Probability the current transaction is a fraud given that it is a foreign transaction, not an internet purchase, and card holder purchased computer related purchases in the past week:

$$P(Fraud = True | IP = False, CRP = True, OC = True) = \\ P(IP = False | CRP = True, Fraud = True) * P(CRP = True) + \\ P(IP = False | CRP = False, Fraud = True) * P(CRP = False) = \\ P(IP = False | CRP = True, Fraud = True) * (P(CRP = True | OC = True) * \\ P(OC = True) + \\ P(CRP = True | OC = False) * P(OC = False)) + \\ P(IP = False | CRP = False, Fraud = True) * (P(CRP = False | OC = True) * \\ P(OC = True) + \\ P(CRP = False | OC = False) * P(OC = False)) = \\ 0.99 * (0.1 * 0.75 + 0.01 * 0.25) + 0.89 * (0.9 * 0.75 + 0.99 * 0.25) = \\ 0.89775 \\ P(Fraud = True | FP = True) = \\ P(Fraud = True | Trav = True) * P(FP = True | Trav = True, Fraud = True) * \\ P(Trav = True) + \\ P(Fraud = True | Trav = False) * P(FP = True | Trav = False, Fraud = True) * \\ P(Trav = False) = \\ 0.01 * 0.90 * 0.05 + 0.004 * 0.10 * 0.95 = \\ 0.00083 \\ P(Fraud = True | IP = False, CRP = True, OC = True, FP = True) = \\ 0.89775 * 0.00083 = \\ 0.000745 \\ \hline$$