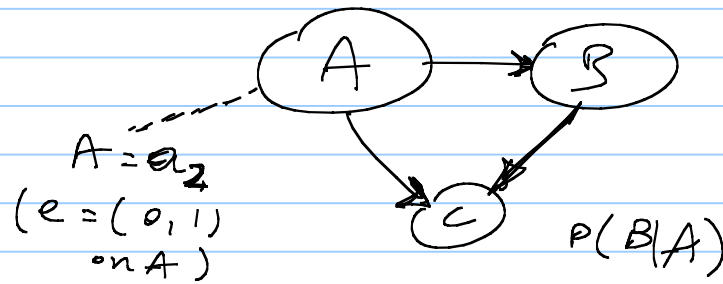


HW3 CSCI 582 due 2009-02-14 All exercises from [J07]

2.17



$$P(A) = (0.9, 0.1)$$

B \ A		
	a_1	a_2
b_1	0.3	0.6
b_2	0.7	0.4

$P(B|A)$

B \ A		
	a_1	a_2
b_1	$(0.1, 0.9)$	$(0.1, 0.9)$
b_2	$(0.2, 0.8)$	$(0.2, 0.8)$

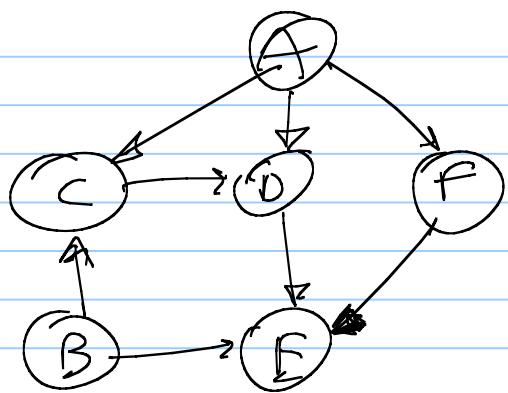
$P(C|A, B)$

$$P(B, C, e) = \sum_{A} \prod_{v \in A, B, C} P(v | \text{pa}(v)) \cdot e = P(A=a_2) \cdot P(B|A=a_2) \cdot P(C|A=a_2, B) =$$

$$0.1 \times \begin{pmatrix} 0.6 \\ 0.4 \end{pmatrix} \times \begin{array}{c|cc} & B & b_1 & b_2 \\ \hline C & & & \\ c_1 & 0.1 & 0.2 \\ c_2 & 0.9 & 0.8 \end{array}$$

C \ B		
	b_1	b_2
c_1	$0.1 \times 0.6 \times 0.1$ 0.006	$0.1 \times 0.4 \times 0.2$ 0.008
c_2	$0.1 \times 0.6 \times 0.9$ 0.054	$0.1 \times 0.4 \times 0.8$ 0.032

2.18



$P(A)$
 $P(B)$
 $P(C|A,B)$
 $P(D|A,C)$
 $P(E|B,D,F)$
 $P(F|A)$

Need: $P(F|C=c_1)$

$$\sum_{A,B,D,E} P(A)P(B)P(C_1|A,B)P(D|A,C_1)P(E|B,D,F)P(F|A) =$$

$$= \sum_{A,B,D} P(A)P(B)P(C_1|A,B)P(D|A,C_1)P(F|A) \sum_E P(E|B,D,F) =$$

$$= \sum_{A,B} P(A)P(B)P(C_1|A,B)P(F|A) \sum_D P(D|A,C_1) = \textcircled{*} =$$

$$= \sum_B P(B) \sum_A P(A)P(C_1|A,B)P(F|A) = \sum_B P(B) H_B(F) =$$

$= H_B(F)$. The answer is $\frac{H_B(F)}{H_B(F)}$, i.e., the result of normalizing $H_B(F)$.

The largest table has size 6000, and it is used to compute $H_A(B,F)$.

Note that we do not need to work with the 10^4 -sized table $P(E|D, B, F)$, because $\sum_E P(E|D, B, F)$ is one.

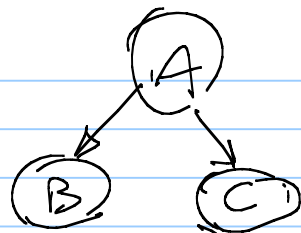
We could save computation time and space by summing over B before summing over A :

$$\begin{aligned} (*) &= \sum_A P(A) P(F|A) \sum_B P(B) P(c_i|A, B) = \\ &= \sum_A P(A) P(F|A) H_B(A) = H_A(F). \end{aligned}$$

The answer is $\frac{H_A(F)}{\sum_F H_A(F)}$, i.e., the result of normalizing $H_A(F)$.

Now, the largest tables have size 100. They are used to compute $H_B(A)$ and $H_A(F)$.

2.19



[See errata!!]

$P(B|A, c) = P(B|A)$ because A d-separates B and c and d-separation implies conditional independence.

Also, you can write $P(A, B, C) = (\text{chain rule for BNs}) = P(A)P(B|A)P(C|A)$
 So, $P(A, c) = \sum_B P(A)P(B|A)P(C|A) = P(A)P(C|A) \sum_B P(B|A) = P(A)P(C|A)$

$$P(A, B) = \sum_C P(A)P(B|A)P(C|A) = P(A)P(B|A)$$

$$P(A) = \sum_{B, c} P(A)P(B|A)P(C|A) = P(A)$$

So,

$$P(B|A, c) = \frac{P(A)P(B|A)P(C|A)}{P(A)P(C|A)} = P(B|A) = \frac{P(A)P(B|A)}{P(A)} \quad \checkmark \quad \checkmark$$

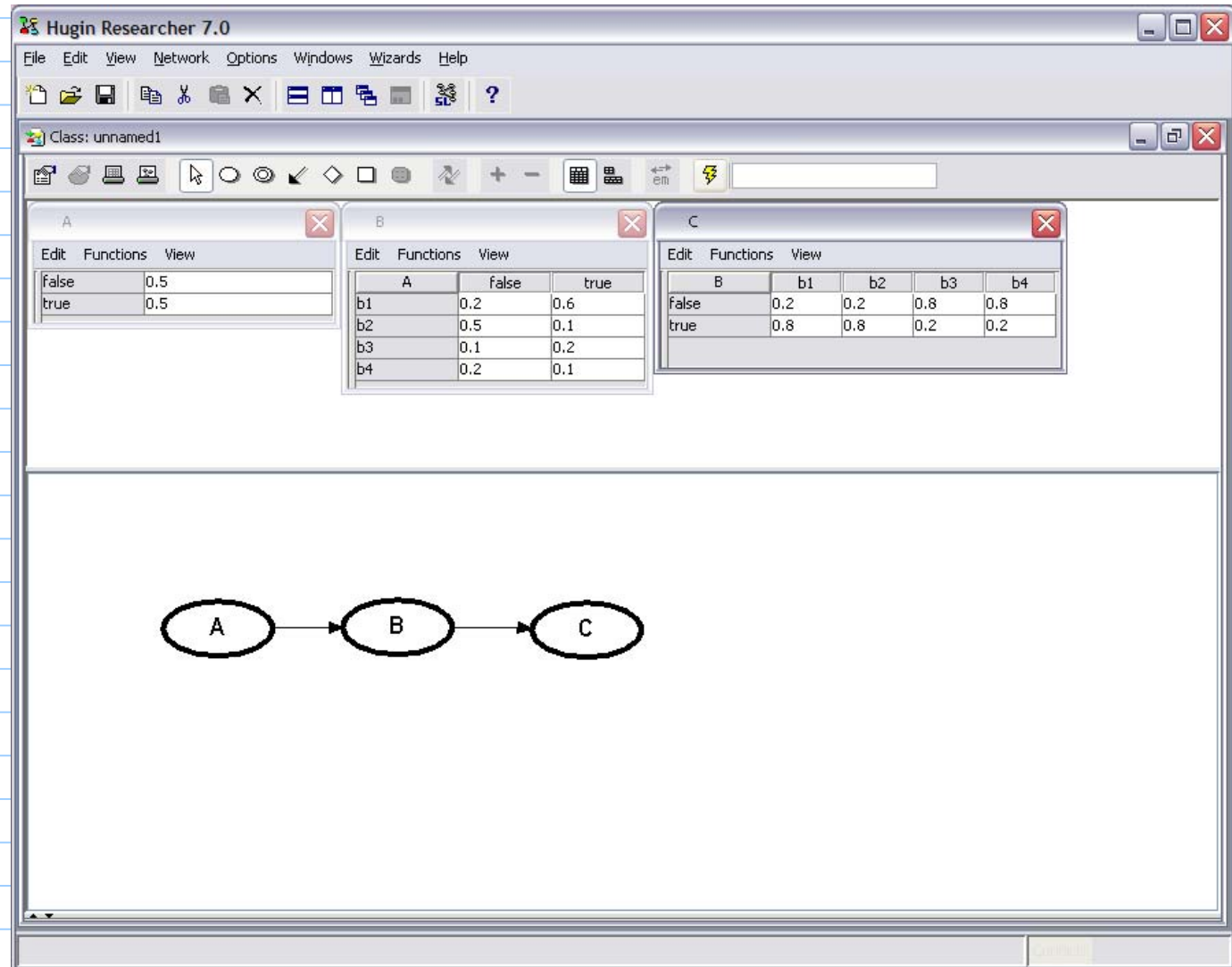
$$P(A, B, C) = P(A)P(B|A)P(C|A) = \begin{matrix} a_1 & a_2 \\ 0.1 & 0.9 \end{matrix} \times \begin{matrix} a_1 & a_2 \\ b_1 & \begin{matrix} 0.2 & 0.3 \\ 0.8 & 0.7 \end{matrix} \\ b_2 \end{matrix} \times \begin{matrix} a_1 & a_2 \\ c_1 & \begin{matrix} 0.5 & 0.6 \\ 0.5 & 0.4 \end{matrix} \\ c_2 \end{matrix} =$$

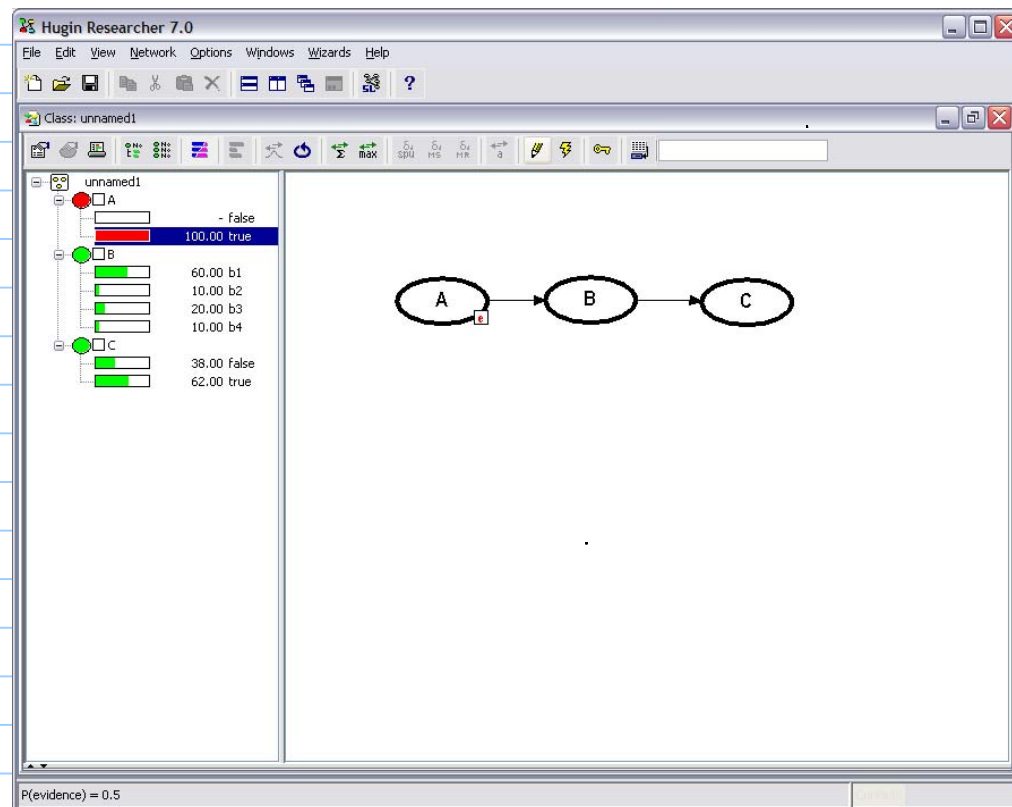
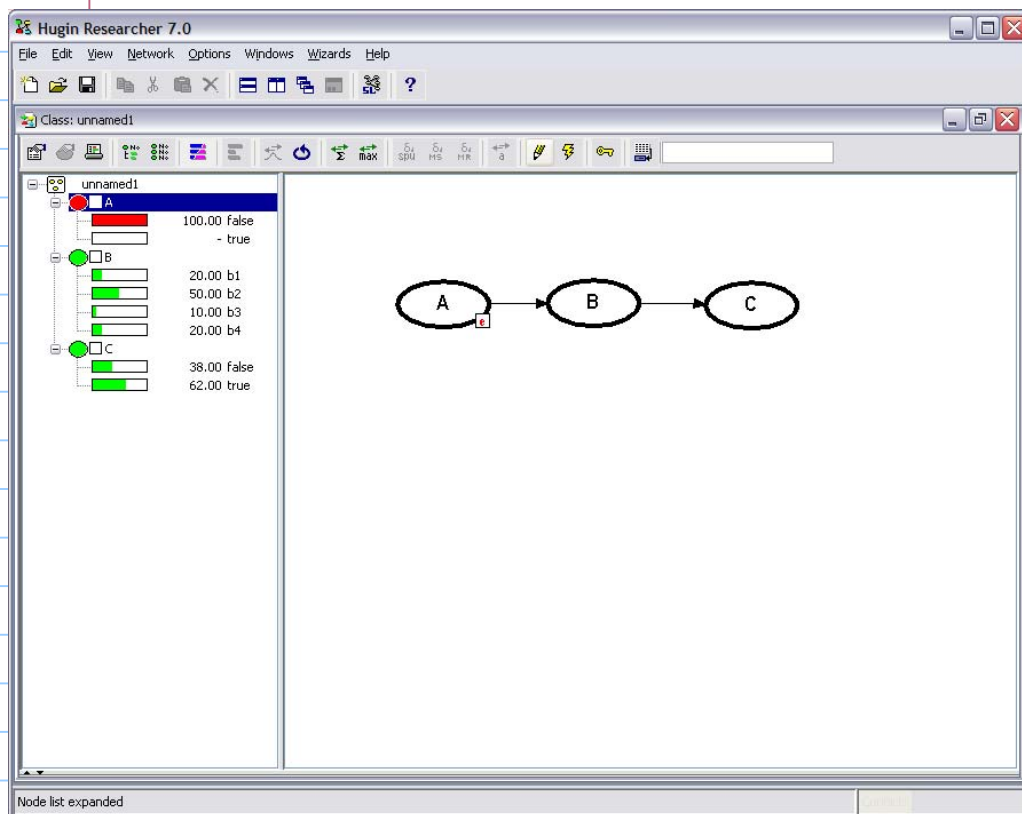
$$= \begin{matrix} a_1 & a_2 \\ b_1 & \begin{matrix} (0.1 \times 0.2 \times 0.5, 0.1 \times 0.2 \times 0.5) & (0.9 \times 0.3 \times 0.6, 0.9 \times 0.3 \times 0.4) \\ 0.01 & 0.01 & 0.162 & 0.108 \end{matrix} \\ b_2 & \begin{matrix} (0.1 \times 0.8 \times 0.5, 0.1 \times 0.8 \times 0.5) & (0.9 \times 0.7 \times 0.6, 0.9 \times 0.7 \times 0.4) \\ 0.04 & 0.04 & 0.378 & 0.252 \end{matrix} \end{matrix}$$

$$\begin{array}{r} 18x \quad 12x \\ 9 \quad 9 \\ \hline 162 \quad 108 \\ 42x \quad 28x \\ 9 \quad 9 \\ \hline 378 \quad 252 \end{array}$$

$$\text{Check: } \underbrace{0.01 + 0.01 + 0.04 + 0.04}_{0.1} + \underbrace{0.162 + 0.108}_{0.27} + \underbrace{0.378 + 0.252}_{0.63} = 1 \quad \checkmark$$

2,23





Note that d-separation implies conditional independence, but, as this example shows, the converse is not true!