

14.14 Consider the Bayes net shown in Figure 14.23.

a. Which of the following are asserted by the network structure?

(i) $P(B, I, M) = P(B)P(I)P(M)$.

(ii) $P(J | G) = P(J | G, I)$.

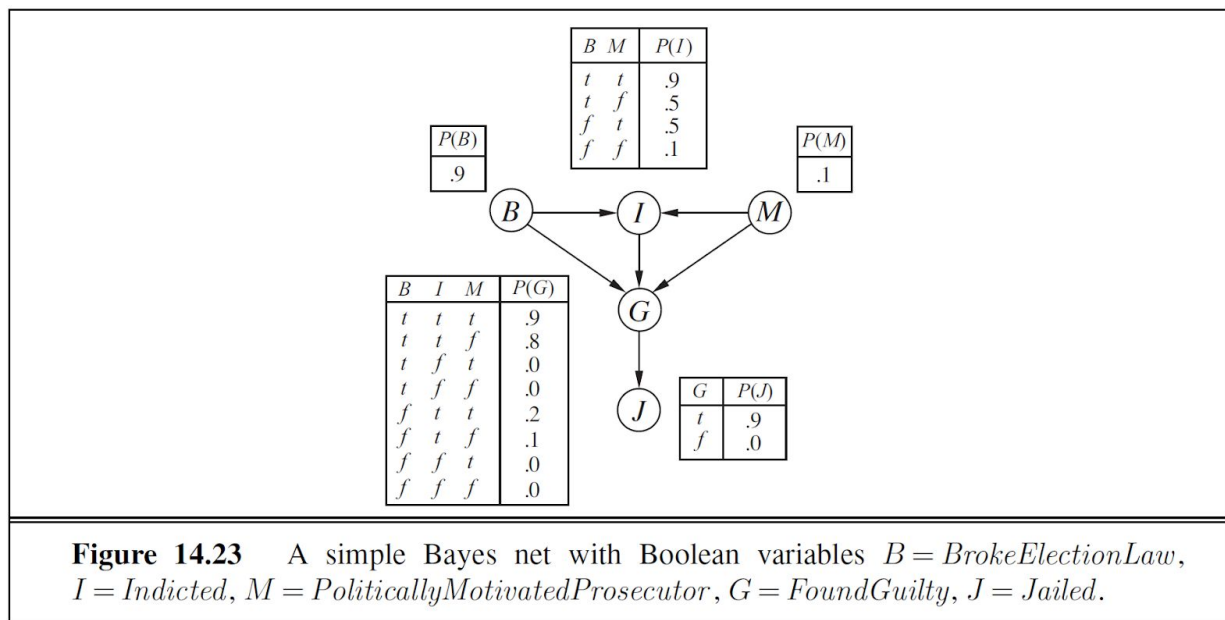
(iii) $P(M | G, B, I) = P(M | G, B, I, J)$.

b. Calculate the value of $P(b, i, -m, g, j)$.

c. Calculate the probability that someone goes to jail given that they broke the law, have been indicted, and face a politically motivated prosecutor.

d. A context-specific independence (see page 542) allows a variable to be independent of some of its parents given certain values of others. In addition to the usual conditional independences given by the graph structure, what context-specific independences exist in the Bayes net in Figure 14.23?

e. Suppose we want to add the variable P = PresidentialPardon to the network; draw the new network and briefly explain any links you add.



a. Which of the following are asserted by the network structure?

(i) $P(B, I, M) = P(B)P(I)P(M)$.

(ii) $P(J | G) = P(J | G, I)$.

(iii) $P(M | G, B, I) = P(M | G, B, I, J)$.

(i) Is not asserted by network structure

"The topological semantics specifies that each variable is conditionally independent of its non-descendants, given its parents." [AIMA3E, p. 517]

By Chain Rule and AIMA3E Equation 14.3:

$$P(B, I, M) = P(B | \text{Parents}(B)).P(I | \text{Parents}(I)).P(M | \text{Parents}(M))$$

$$P(B, I, M) = P(B).P(I | B, M).P(M)$$

$$P(B).P(I | B, M).P(M) \neq P(B).P(I).P(M)$$

(ii) Is asserted by network structure

By AIMA3E Equation 14.3:

$$P(J \mid B, I, M, G) = P(J \mid \text{Parents}(J))$$

$$P(J \mid B, I, M, G, I) = P(J \mid \text{Parents}(J), I)$$

$$P(J \mid B, I, M, G) = P(J \mid G, I)$$

(iii) Is asserted by network structure

"...a node is conditionally independent of all other nodes in the network, given its parents, children, and children's parents—that is, given its Markov blanket." [AIMA3E, p. 517] Proof in Exercise 14.7.

$$P(M \mid G, B, I, J) = P(M \mid \text{Markov_Blanket}(M))$$

$$P(M \mid G, B, I, J) = P(M \mid I, G)$$

$$P(M \mid G, B, I, J, B) = P(M \mid I, G, B)$$

$$P(M \mid G, B, I, J) = P(M \mid I, G, B)$$

b. Calculate the value of $P(b, i, \neg m, g, j)$.

By Chain Rule and Equation 14.3:

$$P(b, i, \neg m, g, j) = P(b) \cdot P(\neg m) \cdot P(i \mid b, \neg m) \cdot P(g \mid b, i, \neg m) \cdot P(j \mid g)$$

Then we only need to use the Conditional Probabilities Tables of the network:

$$P(b, i, \neg m, g, j) = 0.9 \cdot 0.9 \cdot 0.5 \cdot 0.8 \cdot 0.9$$

$$P(b, i, \neg m, g, j) = 0.2916$$

c. Calculate the probability that someone goes to jail given that they broke the law, have been indicted, and face a politically motivated prosecutor.

$$P(j \mid b, i, m) = P(j, b, i, m) / P(b, i, m)$$

Chain Rule on denominator

$$P(b, i, m) = P(b) \cdot P(m) \cdot P(i \mid b, m) = 0.9 \cdot 0.1 \cdot 0.9 = 0.081$$

Marginalization Rule on numerator:

$$P(j, b, i, m) = P(j, b, i, m, g) + P(j, b, i, m, \neg g)$$

Chain Rule on numerator:

$$P(j, b, i, m) = P(b) \cdot P(m) \cdot P(i \mid b, m) \cdot P(g \mid b, i, m) \cdot P(j \mid g) + P(b) \cdot P(m) \cdot P(i \mid b, m) \cdot P(\neg g \mid b, i, m) \cdot P(j \mid \neg g)$$

$$P(j, b, i, m) = P(b) \cdot P(m) \cdot P(i \mid b, m) \cdot (P(g \mid b, i, m) \cdot P(j \mid g) + P(\neg g \mid b, i, m) \cdot P(j \mid \neg g))$$

Using CPTs for numerator:

$$P(j, b, i, m) = 0.9 \cdot 0.1 \cdot 0.9 \cdot (0.9 \cdot 0.9 + 0.1 \cdot 0) = 0.081 \cdot 0.81$$

Then we can resolve what we were looking for:

$$P(j \mid b, i, m) = 0.081 \cdot 0.81 / 0.081 = 0.81$$

d. A context-specific independence (see page 542) allows a variable to be independent of some of its parents given certain values of others. In addition to the usual conditional independences given by the graph structure, what context-specific independences exist in the Bayes net in Figure 14.23?

Variable G is independent of variables M and B when variable I is false.

e. Suppose we want to add the variable $P = \text{PresidentialPardon}$ to the network; draw the new network and briefly explain any links you add.

If we look at the network $P(J=\text{true} \mid G=\text{true}) = 0.9$ meaning that there are other hidden vars that allows a person avoid the chair. Then, the PresidentialPardon could be one of these variables.

Then variable P will be parent of J . Then J will be evaluated given G and P : $P(J \mid G, P)$

The probability that $P=\text{True}$ when G is false is 0 (zero) because the president can not give his or her pardon to an innocent person. So variable G is parent of P . I think until here this could be a simple network that is compliant with requirement. We can complicate things a little. What if Presidential Pardon is related with the kind of crime or its reasons. Then we could add B and M as parents of P too.

