

2.9. Let G be an undirected graph. In this case, begin the breadth first search and view all the neighbors of a given node. We call this node g^0 and the set of its neighbors g_n^0 . Then we check the each of the neighbors. As we travel to them, we label each one g^1 and the set of its neighbors g_n^1 . If $g_n^0 \cap g_n^1 \neq \emptyset$, then there is a cycle. Once we have passed through all neighbors, we check the neighbors of the first node in g_n^0 and label this node g_n^0 and repeat the process. Further, if a node in is the neighborhood of the current node g^i and is not the previous node g^{i-1} (or g^{i+1}), then there is a cycle.

3.1. Let X_1, X_2, X_3 be random variables. We define the following properties:

$$P(X_1 = x_0) = .5 \quad (1)$$

$$P(X_1 = x_1) = .5 \quad (2)$$

$$P(X_2 = x_0) = .7 \quad (3)$$

$$P(X_2 = x_1) = .3 \quad (4)$$

$$P(X_3 = x_0) = .2 \quad (5)$$

$$P(X_3 = x_1) = .8. \quad (6)$$

As we can see, all the variables are independent. Then we define the probabilities

$$P(X_1 = x_0, X_2 = x_0 \mid X_3 = x_0) = .13 \quad (7)$$

$$P(X_1 = x_0, X_2 = x_0 \mid X_3 = x_1) = .02 \quad (8)$$

$$P(X_1 = x_1, X_2 = x_1 \mid X_3 = x_0) = .3 \quad (9)$$

$$P(X_1 = x_1, X_2 = x_1 \mid X_3 = x_1) = .05. \quad (10)$$

Thus we see that X_1, X_2 is dependent on X_3 .

3.2.

1. Let X_1, \dots, X_n be random variables and C be a class variable. Assume that $(X_i \perp \mathbf{X}_{-i} \mid C)$ for all i where $\mathbf{X}_{-i} = \{X_1, \dots, X_n\} - \{X_i\}$. Then

$$P(C, X_1, \dots, X_n) = P(C)P(X_1 \mid C)P(X_2, \dots, X_n \mid X_1, C) \quad (11)$$

$$= P(C)P(X_1 \mid C)P(X_2 \mid X_1, C)P(X_3, \dots, X_n \mid X_1, X_2, C) \quad (12)$$

$$= P(C) \prod_{i=1}^n P(X_i \mid C, X_{i-1}, \dots, X_1) \quad (13)$$

$$= P(C) \prod_{i=1}^n P(X_i \mid C), \quad (14)$$

where the last step used the assumption of independence.

2. Using the property proven in part (i), we see that

$$\frac{P(C = c_1 \mid X_1, \dots, X_n)}{P(C = c_2 \mid X_1, \dots, X_n)} = \frac{\frac{P(X_1, \dots, X_n \mid C=c_1)P(C=c_1)}{P(X_1, \dots, X_n)}}{\frac{P(X_1, \dots, X_n \mid C=c_2)P(C=c_2)}{P(X_1, \dots, X_n)}} \quad (15)$$

$$= \frac{P(X_1, \dots, X_n \mid C = c_1)P(C = c_1)}{P(X_1, \dots, X_n \mid C = c_2)P(C = c_2)} \quad (16)$$

$$= \frac{P(C = c_1)}{P(C = c_2)} \prod_{i=1}^n \frac{P(X_i \mid C = c_1)}{P(X_i \mid C = c_2)}. \quad (17)$$

3.3.

1. While having an earthquake does not eliminate the possibility of a burglary, having an earthquake causes the alarm to go off. So if an alarm goes off and there is an earthquake, it decreases the probability that a burglary set off the alarm because the earthquake did happen and was able to set the alarm off.
2. Assume that $P(a^1 | b^1, e^1) = P(a^1 | b^0, e^1) = 1$. Then

$$P(b^1 | a^1, e^1) = \frac{P(b^1)P(a^1, e^1 | b^1)}{P(b^0)P(a^1, e^1 | b^0) + P(b^1)P(a^1, e^1 | b^1)} \quad (18)$$

$$= \frac{P(b^1)P(a^1 | e^1, b^1)}{P(b^0)P(a^1 | e^1, b^0) + P(b^1)P(a^1 | e^1, b^1)} \quad (19)$$

$$= \frac{P(b^1)}{P(b^0) + P(b^1)} = P(b^1). \quad (20)$$

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