

Table 5.11. Data set for Exercise 8.

Instance	A	B	C	Class
1	0	0	1	-
2	1	0	1	+
3	0	1	0	-
4	1	0	0	-
5	1	0	1	+
6	0	0	1	+
7	1	1	0	-
8	0	0	0	-
9	0	1	0	+
10	1	1	1	+

- (b) Use the conditional probabilities in part (a) to predict the class label for a test sample ($A = 1, B = 1, C = 1$) using the naïve Bayes approach.
 - (c) Compare $P(A = 1), P(B = 1)$, and $P(A = 1, B = 1)$. State the relationships between A and B .
 - (d) Repeat the analysis in part (c) using $P(A = 1), P(B = 0)$, and $P(A = 1, B = 0)$.
 - (e) Compare $P(A = 1, B = 1|Class = +)$ against $P(A = 1|Class = +)$ and $P(B = 1|Class = +)$. Are the variables conditionally independent given the class?
9. (a) Explain how naïve Bayes performs on the data set shown in Figure 5.46.
 - (b) If each class is further divided such that there are four classes ($A1, A2, B1$, and $B2$), will naïve Bayes perform better?
 - (c) How will a decision tree perform on this data set (for the two-class problem)? What if there are four classes?
10. Repeat the analysis shown in Example 5.3 for finding the location of a decision boundary using the following information:
 - (a) The prior probabilities are $P(\text{Crocodile}) = 2 \times P(\text{Alligator})$.
 - (b) The prior probabilities are $P(\text{Alligator}) = 2 \times P(\text{Crocodile})$.
 - (c) The prior probabilities are the same, but their standard deviations are different; i.e., $\sigma(\text{Crocodile}) = 4$ and $\sigma(\text{Alligator}) = 2$.
 11. Figure 5.47 illustrates the Bayesian belief network for the data set shown in Table 5.12. (Assume that all the attributes are binary).
 - (a) Draw the probability table for each node in the network.