

# Conditional Probability and Bayes Theorem

**Probability Theory:  
Foundation for Data Science  
with Anne Dougherty**



# Learning Goals for Module 2

In this module, we'll learn about conditional probability and Bayes formula. At the end of this module, learners should be able to:

- ▶ **Explain the concept of conditional probability.**
- ▶ **Calculate probabilities using conditioning and Bayes Theorem.**
- ▶ Explain the concepts of independence and mutually exclusive events and provide examples.
- ▶ See the relationship between conditional and independent events in a statistical experiment.

# Conditional Probability

Suppose we have two events  $A$  and  $B$  from the same sample space  $S$ . We want to calculate the probability of event  $A$ , knowing that event  $B$  has occurred.  $B$  is the “conditioning event”. Notation:  $P(A|B)$ , the probability of event  $A$  given that  $B$  has occurred.

Example: Roll a six-sided dice twice. Recall,  $S = \{(i, j) \mid i, j \in \{1, 2, 3, 4, 5, 6\}\}$ ,  $|S| = 36$  and each of the 36 outcomes of  $S$  is equally likely.

Let  $A$  be the event that at least one of the dice shows a 3.  
 $A = \{(3, 1), (3, 2), \dots, (3, 6), (1, 3), (2, 3), (4, 3), (5, 3), (6, 3)\}$   
 $P(A) = 11/36$

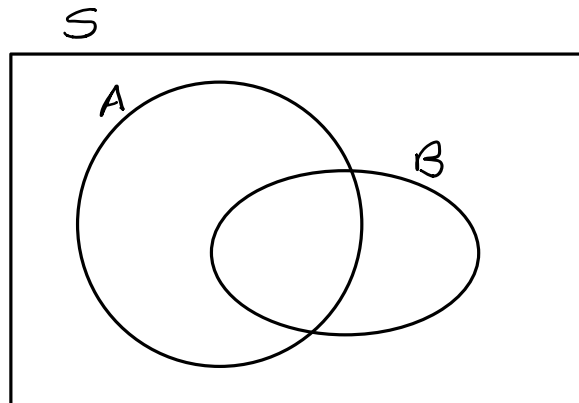
Let  $B$  be the event that the sum of the 2 dice is 9.  
 $B = \{(6, 3), (3, 6), (4, 5), (5, 4)\}$   
 $P(B) = 4/36$

## Example continued

Question: Suppose we know that  $B$  has occurred. How does this change the probability of  $A$ ? That is, find  $P(A|B)$ , the probability that at least one dice was a 3 given that the sum was 9.

►  $P(A \cap B) =$

►  $P(A|B) =$



# Bayes Theorem

**Conditional probability** is defined as:

$$P(A|B) = \frac{P(A \cap B)}{P(B)}, \quad P(B) > 0$$

This leads to the **multiplication rule**

$$P(A \cap B) = P(B)P(A|B)$$

**Bayes Theorem:** Let  $P(B) > 0$ . Then,

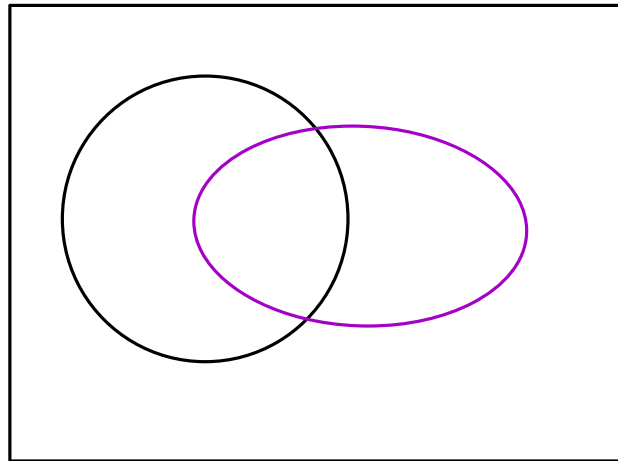
$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

# Law of Total Probability

Given two events  $A$  and  $B$  from the same sample space,

$$B = (B \cap A) \cup (B \cap A^c)$$

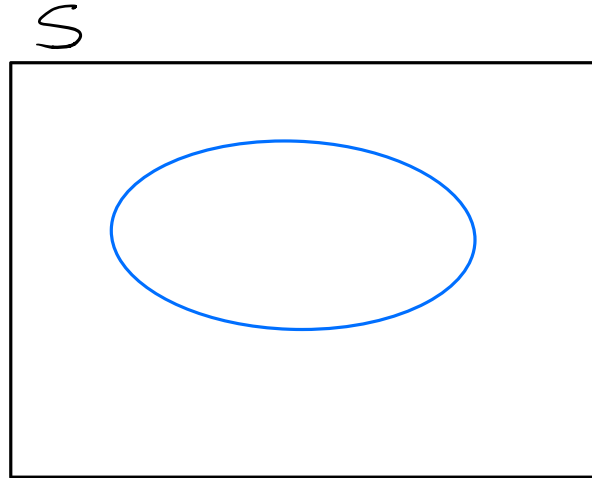
$$P(B) = P(B \cap A) + P(B \cap A^c) = P(B|A)P(A) + P(B|A^c)P(A^c)$$



# Law of Total Probability - continued

Extend this idea to  $n$  sets  $A_1, A_2, \dots, A_n$  where  $A_1 \cap \dots \cap A_n = \emptyset$  and  $\cup_{k=1}^n A_k = S$ . Then,

$$P(B) = \sum_{k=1}^n P(B|A_k)P(A_k).$$



## Example - Testing for a disease

Example: Suppose your company has developed a new test for a disease. Let event  $A$  be the event that a randomly selected individual has the disease and, from other data, you know that 1 in 1000 people has the disease. Thus,  $P(A) = .001$ . Let  $B$  be the event that a positive test result is received for the randomly selected individual. Your company collects data on their new test and finds the following:

- ▶  $P(B|A) = .99$
- ▶  $P(B^c|A) = .01$
- ▶  $P(B|A^c) = .02$

Calculate the probability that the person has the disease, given a positive test result. That is, find  $P(A|B)$ .



## Example - continued

# Example - Tree Diagram

