

# Conditional Probability and Independence

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## 1 Restricting the Sample Space - Conditional Probability

How do we modify the probability of an event in light of the fact that something is known?

If the top card is  $3\spadesuit$ , what is the probability that the second card is a three? a  $\spadesuit$ ? a king?

All of your answers have 51 in the denominator. You have mentally restricted the sample space from  $S$  with 52 outcomes to  $B = \{\text{all cards but } 3\spadesuit\}$  with 51 outcomes. We call the answer the **conditional probability**.

For equally likely outcomes, we have a formula.

$$\begin{aligned} P(A|B) &= \text{the proportion of outcomes in } A \text{ that are also in } B \\ &= \frac{\#(A \cap B)}{\#(B)} = \frac{\#(A \cap B)/\#(S)}{\#(B)/\#(S)} \end{aligned}$$

This allows us to answer the questions above.

$$P\{\text{second card is a three} | \text{first card is } 3\spadesuit\} = \frac{3}{51} = \frac{1}{17}$$

$$P\{\text{second card is a } \spadesuit | \text{first card is } 3\spadesuit\} = \frac{1}{51}$$

$$P\{\text{second card is a king} | \text{first card is } 3\spadesuit\} = \frac{1}{51}$$

The last identity for  $P(A|B)$  with equally likely outcomes can be interpreted as the ratio of probabilities:

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

We thus take this version of the identity as the general definition of conditional expectation for any pair of events  $A$  and  $B$  as long as  $P(B) > 0$

**Exercise 1.** If  $A \subset B$ , then  $P(A|B) = P(A)/P(B)$ . If  $B \subset A$ , then  $P(A|B) = 1$ .

**Example 2.** For two rolls of a die, we have 36 possible outcomes.

$$P\{\text{sum is } k\} = \frac{6 - |k - 7|}{36}.$$

Now suppose that we are to roll two dice until on of two numbers  $k_1$  and  $k_2$  is observed for the sum. Our sample space  $S$  is the set of all outcomes that sum to  $k_1$  or  $k_2$ . Thus,

$$\begin{aligned} P\{\text{sum is } k_1 | \text{sum is } k_1 \text{ or } k_2\} &= \frac{P\{\text{sum is } k_1\}}{P\{\text{sum is } k_1 \text{ or } k_2\}} = \frac{(6 - |k_1 - 7|)/36}{(6 - |k_1 - 7| + 6 - |k_2 - 7|)/36} \\ &= \frac{6 - |k_1 - 7|}{12 - |k_1 - 7| - |k_2 - 7|} \end{aligned}$$

## 2 The Multiplication Principle

Multiply both sides of the defining formula for conditional probability by  $P(B)$  to obtain **the multiplication principle**

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

Now, we can complete an earlier problem

$$\begin{aligned} P\{\text{ace on first two cards}\} &= P\{\text{ace on second card} | \text{ace on first card}\}P\{\text{ace on first card}\} \\ &= \frac{1}{17} \times \frac{1}{13}. \end{aligned}$$

We can continue this process to obtain a **chain rule**

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

Thus,

$$\begin{aligned} &P\{\text{ace on first three cards}\} \\ &= P\{\text{ace on third card} | \text{ace on first and second card}\}P\{\text{ace on second card} | \text{ace on first card}\}P\{\text{ace on first card}\} \\ &= \frac{1}{26} \times \frac{1}{17} \times \frac{1}{13}. \end{aligned}$$

Extending this 4 events, we have:

**Example 3.** In a urn with  $b$  blue balls and  $g$  green balls, the probability of green, blue, green, blue is

$$\frac{g}{b+g} \cdot \frac{b}{b+g-1} \cdot \frac{g-1}{b+g-2} \cdot \frac{b-1}{b+g-3} = \frac{(g)_2(b)_2}{(b+g)_4}$$

**Exercise 4.** What is the probability of 2 blue and 2 green in four draws from the urn above?

### 3 Law of Total Probability

**Definition 5.** A **partition** of the sample space  $S$  is a finite collection of pairwise mutually exclusive events  $\{C_1, C_2, \dots, C_n\}$  whose union is  $S$ .

Thus, every point  $s \in S$  belong to *exactly* one of the  $C_i$ .

**Theorem 6** (Law of total probability). Let  $P$  be a probability on  $S$ . and let  $\{C_1, C_2, \dots, C_n\}$  be a partition of  $S$  chosen so that  $P(C_i) > 0$  for all  $i$ . Then, for any event  $A \subset S$

$$P(A) = \sum_{i=1}^n P(A|C_i)P(C_i).$$

*Proof.* Because  $\{C_1, C_2, \dots, C_n\}$  is a partition,  $\{A \cap C_1, A \cap C_2, \dots, A \cap C_n\}$  are pairwise mutually exclusive events. By the distributive property of sets, their union is  $A$ . Thus,

$$P(A) = \sum_{i=1}^n P(A \cap C_i).$$

Finish by using the identity

$$P(A \cap C_i) = P(A|C_i)P(C_i), \quad i = 1, 2, \dots, n.$$

□

**Example 7** (craps). To play craps, first roll two dice.

- If the sum is 7 or 11, then the player wins immediately.
- If the sum is 2, 3, or 12, then the player loses immediately.
- If the sum is 4, 5, 6, 8, 9, or 10, if the player rolls this number a second time before rolling a 7, then the player wins

So

$$P\{\text{winning immediately}\} = \frac{6}{36} + \frac{2}{36} = \frac{8}{36} = \frac{2}{9}.$$

For  $k = 4, 5, 6, 8, 9$ , or  $10$ ,

$$\begin{aligned} P\{\text{winning with } k\} &= P\{\text{rolling } k \text{ before } 7 | \text{first roll is } k\} \cdot P\{\text{first roll is } k\} \\ &= \frac{6 - |k - 7|}{12 - |k - 7|} \cdot \frac{6 - |k - 7|}{36} = \frac{(6 - |k - 7|)^2}{36(12 - |k - 7|)} \end{aligned}$$

This yields the table

first roll	4	5	6	8	9	10
probability of winning	$\frac{9}{36 \cdot 9}$	$\frac{16}{36 \cdot 10}$	$\frac{25}{36 \cdot 11}$	$\frac{25}{36 \cdot 11}$	$\frac{16}{36 \cdot 10}$	$\frac{9}{36 \cdot 9}$

Thus,

$$P\{\text{winning after the first toss}\} = \frac{2}{36} \left( 1 + \frac{8}{5} + \frac{25}{11} \right) = \frac{1}{18} \left( \frac{55 + 88 + 125}{55} \right) = \frac{268}{18 \cdot 55} = \frac{134}{9 \cdot 55}$$

and

$$P\{\text{winning}\} = \frac{2}{9} + \frac{134}{9 \cdot 55} = \frac{110 + 134}{9 \cdot 55} = \frac{244}{495} = 0.493.$$

## 4 Bayes formula

**Theorem 8** (Bayes formula). *Let  $P$  be a probability on  $S$ . and let  $\{C_1, C_2, \dots, C_n\}$  be a partition of  $S$  chosen so that  $P(C_i) > 0$  for all  $i$ . Then, for any event  $A \subset S$  and any  $j$*

$$P(C_j|A) = \frac{P(A|C_j)P(C_j)}{\sum_{i=1}^n P(A|C_i)P(C_i)}.$$

*Proof.* By the law of total probability, the dominator is equal to  $P(A)$ . By the definition of conditional probability, the numerator is equal to  $P(C_j \cap A)$ . Now, make these two substitutions and use one more time the definition of conditional probability.  $\square$

**Example 9.** *The Public Health Department gives us the following information.*

- *A test for the disease yields a positive result 90% of the time when the disease is present.*
- *A test for the disease yields a positive result 10% of the time when the disease is not present.*
- *One person in 10,000 has the disease.*

*Let*

- *$C$  be the event of having the disease.*
- *$A$  be the event of testing positive for the disease.*

*Then*

$$P(A|C) = 0.90, \quad P(A|C^c) = 0.10, \quad P(C) = 0.0001$$

*and*

$$P(A) = P(A|C)P(C) + P(A|C^c)P(C^c) = 0.90 \cdot 0.0001 + 0.10 \cdot 0.9999 = 0.00009 + 0.09999 = 0.10008.$$

*Thus, the probability of having the disease given that the test was positive is*

$$P(C|A) = \frac{P(A|C)P(C)}{P(A)} = \frac{0.00009}{0.10008} = 0.00089.$$

## 5 Independence

An event  $A$  is **independent of  $B$**  if

$$P(A) = P(A|B).$$

In other words, the knowledge of  $B$  does not influence  $P(A)$ . Multiply both sides of the formula in the definition of independence by  $P(B)$  and use the multiplication rule to obtain

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

This last formula is the usual definition of independence and is symmetric in the events  $A$  and  $B$ . We can also use this to extend the definition to  $n$  independent events

The events  $A_1, \dots, A_n$  are **independent** if

$$P(A_1 \cap \dots \cap A_n) = P(A_1) \times \dots \times P(A_n).$$

**Exercise 10.** *If  $A$  and  $B$  are independent, then show that  $A^c$  and  $B$ ,  $A$  and  $B^c$ ,  $A^c$  and  $B^c$  are also independent.*

Roll two dice.

$$\begin{aligned} \frac{1}{36} &= P\{a \text{ on the first die}, b \text{ on the second die}\} \\ &= \frac{1}{6} \times \frac{1}{6} = P\{a \text{ on the first die}\}P\{b \text{ on the second die}\} \end{aligned}$$

and, thus, the outcomes on two rolls of the dice are independent.