

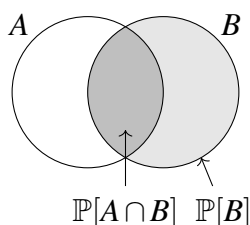
## Conditional Probability Intro

Note 14

**Conditional Probability:** Probability of event  $A$ , given that event  $B$  has happened;

$$\mathbb{P}[A | B] = \frac{\mathbb{P}[A \cap B]}{\mathbb{P}[B]}.$$

Think of like restricting our sample space:



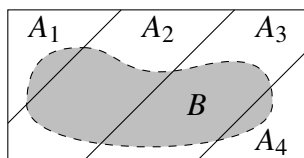
**Bayes Rule:** A consequence of conditional probability - notice  $\mathbb{P}[A \cap B] = \mathbb{P}[A | B]\mathbb{P}[B] = \mathbb{P}[B | A]\mathbb{P}[A]$ , so

$$\mathbb{P}[B | A] = \frac{\mathbb{P}[A \cap B]}{\mathbb{P}[A]} = \frac{\mathbb{P}[A | B]\mathbb{P}[B]}{\mathbb{P}[A]}.$$

**Total Probability Rule:** If disjoint events  $A_1, \dots, A_n$  form a partition on the sample space  $\Omega$ , we then have

$$\mathbb{P}[B] = \sum_{i=1}^n \mathbb{P}[B \cap A_i] = \sum_{i=1}^n \mathbb{P}[B | A_i]\mathbb{P}[A_i].$$

Visually, we're splitting an event into partitions and looking at each intersection individually:



**Independence:** Two events are independent if the following (equivalent) conditions are satisfied. The second definition is probably more intuitive -  $B$  happening does not affect the probability of  $A$  happening.

$$\mathbb{P}[A \cap B] = \mathbb{P}[A]\mathbb{P}[B]$$

$$\mathbb{P}[A | B] = \mathbb{P}[A]$$

## 1 Box of Marbles

Note 14

You are given two boxes: one of them containing 900 red marbles and 100 blue marbles, the other one contains 500 red marbles and 500 blue marbles.

- (a) If we pick one of the boxes randomly, and pick a marble what is the probability that it is blue?
- (b) If we see that the marble is blue, what is the probability that it is chosen from box 1?
- (c) Suppose we pick one marble from box 1 and without looking at its color we put it aside. Then we pick another marble from box 1. What is the probability that the second marble is blue?

### Solution:

- (a) Let  $B$  be the event that the picked marble is blue,  $R$  be the event that it is red,  $A_1$  be the event that the marble is picked from box 1, and  $A_2$  be the event that the marble is picked from box 2. Therefore we want to calculate  $\mathbb{P}[B]$ . By total probability,

$$\mathbb{P}[B] = \mathbb{P}[B | A_1] \mathbb{P}[A_1] + \mathbb{P}[B | A_2] \mathbb{P}[A_2] = 0.5 \times 0.1 + 0.5 \times 0.5 = 0.3.$$

- (b) In this part, we want to find  $\mathbb{P}[A_1 | B]$ . By Bayes rule,

$$\mathbb{P}[A_1 | B] = \frac{\mathbb{P}[B | A_1] \mathbb{P}[A_1]}{\mathbb{P}[B | A_1] \mathbb{P}[A_1] + \mathbb{P}[B | A_2] \mathbb{P}[A_2]} = \frac{0.1 \times 0.5}{0.5 \times 0.1 + 0.5 \times 0.5} = \frac{1}{6}.$$

- (c) Let  $B_1$  be the event that first marble is blue,  $R_1$  be the event that the first marble is red, and  $B_2$  be the event that second marble is blue without looking at the color of first marble. We want to find  $\mathbb{P}[B_2]$ . By total probability,

$$\mathbb{P}[B_2] = \mathbb{P}[B_2 | B_1] \mathbb{P}[B_1] + \mathbb{P}[B_2 | R_1] \mathbb{P}[R_1] = \frac{99}{999} \times 0.1 + \frac{100}{999} \times 0.9 = 0.1.$$

More generally, one can see that the probability that the  $n$ -th marble picked from box 1 is blue with probability 0.1. This is clear by symmetry: all the permutations of the 1000 marbles have the same probability, so the probability that the  $n$ -th marble is blue is the same as the probability that the first marble is blue.

## 2 Duelling Meteorologists

Note 14

Tom is a meteorologist in New York. On days when it snows, Tom correctly predicts the snow 70% of the time. When it doesn't snow, he correctly predicts no snow 95% of the time. In New York, it snows on 10% of all days.

- (a) If Tom says that it is going to snow, what is the probability it will actually snow?

- (b) Let  $A$  be the event that, on a given day, Tom predicts the weather correctly. What is  $\mathbb{P}[A]$ ?
- (c) Tom's friend Jerry is a meteorologist in Alaska. Jerry claims that she is a better meteorologist than Tom even though her overall accuracy is lower. After looking at their records, you determine that Jerry is indeed better than Tom at predicting snow on snowy days and sun on sunny day. Give an instance of the situation described above. This situation is actually an example of the famous Simpson's paradox! *Hint: what is the weather like in Alaska, as compared to in New York?*

**Solution:**

- (a) Let  $S$  be the event that it snows and  $T$  be the event that Tom predicts snow.

$$\begin{aligned}\mathbb{P}[S|T] &= \frac{\mathbb{P}[S \cap T]}{\mathbb{P}[T]} \\ &= \frac{\mathbb{P}[S] \cdot \mathbb{P}[T|S]}{\mathbb{P}[S \cap T] + \mathbb{P}[\bar{S} \cap T]} \\ &= \frac{\frac{1}{10} \times \frac{7}{10}}{\frac{1}{10} \times \frac{7}{10} + \frac{9}{10} \times \frac{5}{100}} = \frac{14}{23}\end{aligned}$$

- (b)

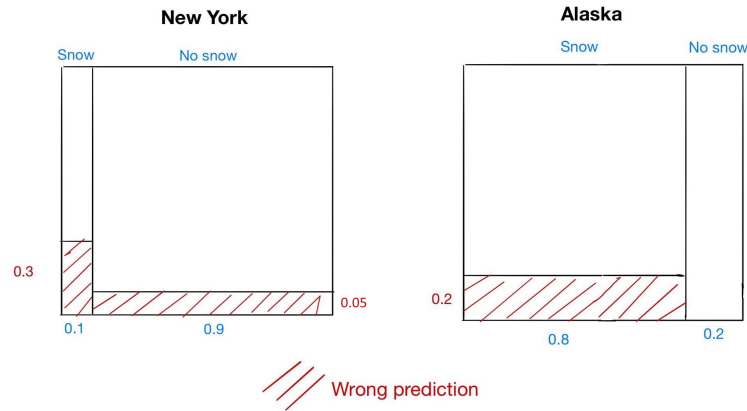
$$\begin{aligned}\mathbb{P}[A] &= \mathbb{P}[S \cap T] + \mathbb{P}[\bar{S} \cap \bar{T}] \\ &= \frac{1}{10} \times \frac{7}{10} + \frac{9}{10} \times \frac{95}{100} = \frac{37}{40}\end{aligned}$$

- (c) Even though Jerry's overall accuracy is lower, it is still possible that she is a better meteorologist if the weather is different.

For example, let's assume that it snows 80% of days in Alaska.

- When it snows, Jerry correctly predicts snow 80% of the time.
- When it doesn't snow, Jerry correctly predicts no snow 100% of the time.

Jerry's overall accuracy turns out to be less than Tom's even though she is better at predicting both categories! The following diagram gives an illustration of the situation. The intuition is that Jerry's error gets penalized more heavily than Tom because it snows more often in Alaska.



For more info on this kind of phenomena, check out [Simpson's Paradox!](#)

### 3 Pairwise Independence

Note 14

Recall that the events  $A_1$ ,  $A_2$ , and  $A_3$  are *pairwise independent* if for all  $i \neq j$ ,  $A_i$  is independent of  $A_j$ . However, pairwise independence is a weaker statement than *mutual independence*, which requires the additional condition that  $\mathbb{P}[A_1 \cap A_2 \cap A_3] = \mathbb{P}[A_1]\mathbb{P}[A_2]\mathbb{P}[A_3]$ .

Suppose you roll two fair six-sided dice. Let  $A_1$  be the event that the first die lands on 1, let  $A_2$  be the event that the second die lands on 6, and let  $A_3$  be the event that the two dice sum to 7.

- Compute  $\mathbb{P}[A_1]$ ,  $\mathbb{P}[A_2]$ , and  $\mathbb{P}[A_3]$ .
- Are  $A_1$  and  $A_2$  independent?
- Are  $A_2$  and  $A_3$  independent?
- Are  $A_1$ ,  $A_2$ , and  $A_3$  pairwise independent?
- Are  $A_1$ ,  $A_2$ , and  $A_3$  mutually independent?

#### Solution:

- We have that  $\mathbb{P}[A_1] = \mathbb{P}[A_2] = \frac{1}{6}$ , since we have a  $\frac{1}{6}$  probability of getting a particular number on a fair die.

Since there are 6 ways in which the two dice can sum to 7 (i.e.  $\{(1, 6), (2, 5), (3, 4), (4, 3), (5, 2), (6, 1)\}$ ), we have  $\mathbb{P}[A_3] = \frac{1}{6}$  as well.

- We want to determine whether  $\mathbb{P}[A_1 \cap A_2] = \mathbb{P}[A_1]\mathbb{P}[A_2]$ . We already found the probabilities of  $A_1$  and  $A_2$  from part (a), so let's look at  $\mathbb{P}[A_1 \cap A_2]$ . There's only one possible outcome where the first die is a 1 and the second die is a 6, so this gives a probability of  $\mathbb{P}[A_1 \cap A_2] = \frac{1}{36}$ .

Since  $\mathbb{P}[A_1]\mathbb{P}[A_2] = \frac{1}{6} \cdot \frac{1}{6} = \frac{1}{36} = \mathbb{P}[A_1 \cap A_2]$ , these two events are independent.

- (c) We want to determine whether  $\mathbb{P}[A_2 \cap A_3] = \mathbb{P}[A_2]\mathbb{P}[A_3]$ . We already found the probabilities of  $A_2$  and  $A_3$  from part (a), so let's look at  $\mathbb{P}[A_2 \cap A_3]$ . These two events both occur if the second die lands on a 6, and the two dice sum to 7. There's only one way that this can happen, i.e. the first die must be a 1, so the intersection has probability  $\mathbb{P}[A_2 \cap A_3] = \frac{1}{36}$ .

Since  $\mathbb{P}[A_2]\mathbb{P}[A_3] = \frac{1}{6} \cdot \frac{1}{6} = \frac{1}{36} = \mathbb{P}[A_2 \cap A_3]$ , these two events are independent.

- (d) To see whether the three events are pairwise independent, we need to ensure that all pairs of events are independent. We've already checked that  $A_1$  and  $A_2$  are independent, and that  $A_2$  and  $A_3$  are independent, so it suffices to check whether  $A_1$  and  $A_3$  are independent.

Similar to the previous two parts, the intersection  $A_1 \cap A_3$  means that the first die must land on a 1, and the two dice sum to 7. There's only one way for this to happen, i.e. the second die must land on a 6, so the probability is  $\mathbb{P}[A_1 \cap A_3] = \frac{1}{36}$ .

Since  $\mathbb{P}[A_1]\mathbb{P}[A_3] = \frac{1}{6} \cdot \frac{1}{6} = \frac{1}{36} = \mathbb{P}[A_1 \cap A_3]$ , these two events are also independent. Since we've now shown that all possible pairs of events are independent,  $A_1$ ,  $A_2$ , and  $A_3$  are indeed pairwise independent.

- (e) Mutual independence requires the additional constraint that  $\mathbb{P}[A_1 \cap A_2 \cap A_3] = \mathbb{P}[A_1]\mathbb{P}[A_2]\mathbb{P}[A_3]$ . We've found the individual probabilities of these events in part (a), so we only need to compute  $\mathbb{P}[A_1 \cap A_2 \cap A_3]$ .

Here, we must have that the first die lands on 1, the second die lands on 6, and the sum of the two dice is equal to 7. There's only one way for this to happen, i.e. the first die is a 1 and the second die is a 6, so the probability of the intersection of all three events is  $\mathbb{P}[A_1 \cap A_2 \cap A_3] = \frac{1}{36}$ .

However, since  $\mathbb{P}[A_1]\mathbb{P}[A_2]\mathbb{P}[A_3] = \frac{1}{6} \cdot \frac{1}{6} \cdot \frac{1}{6} = \frac{1}{216} \neq \frac{1}{36} = \mathbb{P}[A_1 \cap A_2 \cap A_3]$ , these three events are not mutually independent.