# Probability paradoxes

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## Introduction

The goal for the article is to demonstrate several paradoxes that are related to probability theory and how can they can be solved.

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### Chapter 1

# Base definitions of probability theory

I am going to provide several definitions. I will give the both formal and informal definitions and show how they are related each other.

#### 1.1 Example and motivation

We will start with the simplest example.

**Example 1.1.** In the example we have (see fig. 1.1) N = 5 balls. There are  $N_G = 2$  green balls and  $N_R$  red balls. I.e.  $N = N_G + N_R$ .

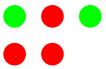


Figure 1.1: Probability example

We can define the probability to get green ball as

$$P_G = \frac{N_G}{N} = \frac{2}{5}$$

and the probability to get red ball as

$$P_R = \frac{N_R}{N} = \frac{3}{5}.$$

We can get only a red or a green ball and

$$P_G + P_R = 1.$$

We will formalize the probability calculations for such cases via the counting principle and define the probability as a ratio of number desired outcomes |A| to the number of all possible outcomes  $|\Omega|$ :

$$P = \frac{|A|}{|\Omega|}.$$

There is a simple example on counting principle below.

**Example 1.2** (Fish in a pond). The example is from a Russian Biological Olympiad. Consider a pond with fishes. 15 of them were marked. After sometime we took 15 fishes and 5 of them were marked. How many fishes are in the pond.

The accepted answer was 45 with the following explanation:

$$\frac{15}{5} = \frac{n}{15}$$

therefore n = 45.

Lets try to solve the task with probability theory and convert the question to the following one: How many fishes are in the pond if in every 15 fishes with high probability we get 5 marked ones?

Let n the number of fishes in the pond. Then the size of the sample space  $|\Omega|$  is the following

$$|\Omega| = \binom{n}{15}$$

i.e. how many ways to get 15 fishes from n.

The desired outcome has the size |A| combined from 2 ones: getting 5 fishes from 15 and get the rest (10) from non marked fishes: n-15:

$$|A| = \binom{15}{5} \cdot \binom{n-15}{10}.$$

Therefore the result can be calculated as follows

$$P_n = \frac{|A|}{|\Omega|} = \frac{\binom{15}{5} \cdot \binom{n-15}{10}}{\binom{n}{15}}.$$

Quick calculations [1] show that n = 45 is very close to real answer:

- \$ stack repl
- > map fish\_in\_pond [55, 50, 45, 40, 35, 30]
- [0.21391501072376837,0.24492699593153727,0.26162279575176545,
  - 0.2440273978093778,0.17082265318953427,5.813662441225208e-2]

**Example 1.3** (Birthday paradox). Consider n people and give a prediction that there is at least one pair of people who have a birthday at the same day. We will call the event as A and there fore are required to find P(A). The straightforward calculations are not easy and we will try to find a probability for another event  $A^c$  that is the complement of event A. I.e. event  $A^c$  states that there is no such pairs and all birthdays are different. We can calculate the desired probability via

$$P_n = P(A) = 1 - P(A^c).$$

We will use the standard counting here. There are totally  $365^n$  possible options for different birthdays in the group on n people. The number of outcomes that satisfied  $A^c$  is

$$365 \cdot 364 \cdot \dots \cdot (365 - n + 1) = \prod_{i=1}^{n} (365 - i + 1).$$

Therefore

$$P(A^c) = \frac{\prod_{i=1}^{n} (365 - i + 1)}{365^n}$$

and

$$P_n = 1 - \frac{\prod_{i=1}^n (365 - i + 1)}{365^n}.$$

Calculations gives us  $P_{10} = 0.1169$ ,  $P_{30} = 0.7063$ ,  $P_{60} = 0.9941$ . I.e. we can say with high probability that in the group of 60 people there will be at least 2 persons with the same birthday.

#### 1.2 Definitions

Now we are ready to give several formal definitions.

#### 1.2.1 $\sigma$ -algebra

**Definition 1.4** (Power set). Let  $\Omega$  is a set than the set of all possible subsets of  $\Omega$  is called *power set* and denoted as  $\mathcal{P}(\omega)$ .

**Definition 1.5** ( $\sigma$  algebra). Let  $\Omega$  is a set then a subset  $\mathcal{F}$  of Power set  $\mathcal{P}(\Omega)$  ( $\mathcal{F} \subseteq \mathcal{P}(\Omega)$ ) is called  $\sigma$  algebra if the following conditions are satisfied:

- 1.  $\mathcal{F}$  contains  $\Omega$ :  $\Omega \in \mathcal{F}$
- 2. TBD
- 3. TBD

In our example 1.1,  $\sigma$  algebra is a collection of any balls.



Figure 1.2: Probability space. It consists of elementary events: a, b, c and d, each of them has equal probability  $p_{a,b,c,d} = \frac{1}{4}$ 

#### 1.3 Conditional probability

**Definition 1.6** (Conditional probability). The *conditional probability* of event A on event B is defined as follow

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

**Example 1.7** (Conditional probability). Lets consider 6 balls each of them can be either two colors (see fig. 1.3).



Figure 1.3: Condition probability. Original probability space.  $P(A=red)=\frac{5}{6},\ P(A=blue)=\frac{3}{6},\ P(A=green)=\frac{4}{6}$ 

You can see that the probability P(A) to get red ball is  $P(A = red) = \frac{5}{6}$ , blue one is  $P(A = blue) = \frac{3}{6}$ , green one is  $P(A = green) = \frac{4}{6}$ .

Now assume that event A is to get a green ball but event B is to get red ball, how we can define P(A|B) in the case.

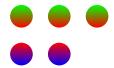


Figure 1.4: Condition probability.  $P(A = green|B = red) = \frac{3}{5}, P(A = blue|B = red) = \frac{2}{5}$ 

The situation is displayed on fig. 1.4. We have only 5 possibilities to choose a ball now instead of 6 in the original case. This is because we just got an additional information - "one of the color should be red". Only 3 of the 5 balls are green. Therefore  $P(A|B) = P(A = green|B = red) = \frac{3}{5}$ .

This result is in correlation with the formal definition of Conditional probability:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A = green \cap B = red)}{P(B = red)} = \frac{3/6}{5/6} = \frac{3}{5}.$$

The fig. 1.5 gives as the view if event B = blue occurs.



Figure 1.5: Condition probability.  $P(A = red|B = blue) = \frac{2}{3}$ ,  $P(A = green|B = blue) = \frac{1}{3}$ 

In the case we have the following conditional probabilities:  $P(A = red|B = blue) = \frac{2}{3}$ ,  $P(A = green|B = blue) = \frac{1}{3}$ .

Finally, the fig. 1.6 gives as the view if event B = green occurs.



Figure 1.6: Condition probability.  $P(A = blue|B = green) = \frac{1}{4}$ ,  $P(A = red|B = green) = \frac{3}{4}$ 

**Example 1.8** (The King's sibling). Suppose that we have a king from a family of 2 children. What's the probability that his sibling is a girl. The important assumption  $^1$  that has to be made is the following: there is no any family planning in the king family and the probability to get a boy  $P_b$  and probability to get a girl  $P_b$  are equally likely:

$$P_b = P_g = \frac{1}{2}.$$

We have 4 cases: bb, bg, gb, gg and the condition that the king is a boy pick up only 3 options for us: bb, bg, gb. All of them are equally likely and 2 have a girl as sibling. I.e.

$$P(sibling = girl|king) = \frac{2}{3}.$$

<sup>&</sup>lt;sup>1</sup> For instance if the king family assume to get a new child until the first boy (king) get then we will have the sibling is girl with probability 1.

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**Proposition 1.9** (Total probability). The total probability is defined as follows

$$P(A) = \sum_{i} P(A|B_i)$$

**Example 1.10** (Total probability). Lets assume in the Conditional probability (Example 1.7) that we are interested in the event A that the ball is green. The other color will be either blue or red. I.e.  $B_1 = blue$ ,  $B_2 = red$ .

$$\begin{split} P(A=green) &= P(A=green|B=blue)P(B=blue) + \\ &+ P(A=green|B=red)P(B=red) = \\ &= \frac{1}{3} \cdot \frac{1}{2} + \frac{3}{5} \cdot \frac{5}{6} = \frac{4}{6}. \end{split}$$

I.e. formula works.

Consider another, not so simple example

**Example 1.11** (Total probability paradox). Let we have 6 balls each of them has one color: red or green (see fig. 1.7).



Figure 1.7: Total probability example

Lets event A is an event to get a ball.  $P(A) = \frac{1}{6}$ . The event  $B_1$  is an event to get green ball:  $P(B_1) = \frac{1}{2}$ . The same one is for probability to get red ball:  $P(B_2) = \frac{1}{2}$ . Conditional probabilities can be calculated as follows:

$$P(A|B_1) = P(A|B_2) = \frac{1}{3}. (1.1)$$

As result the total probability is

$$P(A) = \frac{1}{2} \cdot \frac{1}{3} + \frac{1}{2} \cdot \frac{1}{3} = \frac{2}{6} \neq \frac{1}{6}.$$

The error is in the (1.1). When we consider a concrete ball then it either green or blue and as result one of the conditional probabilities  $P(A|B_1)$  or  $P(A|B_2)$  is zero. In the case we will get correct answer  $P(A) = \frac{1}{6}$ .

**Definition 1.12** (Independence). Two events A and B are independent if

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

**Example 1.13** (Non independent events). Consider situation shown in fig. 1.7. Let event A is that ball is green, event B is that ball is red. We have

$$A \cap B = \emptyset$$
,

i.e.  $P(A \cap B) = 0$ . This means that the events cannot be considered as independent accordingly definition 1.12 as soon as  $P(A) = P(B) = \frac{1}{2}$ .

Really the events are dependent as soon as we can say that A will not occur if B occurs and vice versa.

TBD [2]

# Chapter 2

## Paradoxes

2.1 Monty Hall problem

TBD

2.2 Waiting time on a bus stop

TBD

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