

# Probability Spaces

## Probability and Statistics for Data Science

Carlos Fernandez-Granda



These slides are based on the book [Probability and Statistics for Data Science](#) by Carlos Fernandez-Granda, available for purchase [here](#). A free preprint, videos, code, slides and solutions to exercises are available at <https://www.ps4ds.net>

# Summary

Probability spaces encode common sense

## Key idea

Model uncertain phenomena as **experiments** that can be **repeated over and over**

# Sample space

Set of all possible outcomes of the experiment

Usually denoted by  $\Omega$

## Rolling a 6-sided die

What is the sample space?

$$\Omega := \{1, 2, 3, 4, 5, 6\}$$

# Rolling a die until it lands on 6

What is the sample space?

It depends on how we model!

1.  $\Omega := \{1, 2, 3, \dots\}$  (number of rolls)
2.  $\Omega := \{6, 1 \rightarrow 6, \dots, 5 \rightarrow 6, 1 \rightarrow 1 \rightarrow 6, \dots\}$

# Weather tomorrow in NYC

What is the sample space?

Depends *a lot* on modeling choices

Just interested in temperature at a given time:  $\Omega := \mathbb{R}$



# Overview of probability space

1. Model phenomenon of interest as experiment with a sample space of mutually exclusive **outcomes**
2. Group outcomes in sets called **events**
3. Assign a **probability** to each event

# Rolling a six-sided die

Examples of events:

$$A := \{1, 3, 5\}$$

$$B := \{4\}$$

$$C := \{1, 2, 3, 4, 5, 6\}$$

If we roll a 4, which of these events have occurred?

## Rolling a die until it lands on 6

How many outcomes does the event *Rolling twice* contain?

It depends!

1.  $\Omega_1 := \{1, 2, 3, \dots\}$  (number of rolls)
2.  $\Omega_2 := \{6, 1 \rightarrow 6, \dots, 5 \rightarrow 6, 1 \rightarrow 1 \rightarrow 6, \dots\}$

Weather in NYC tomorrow  $\Omega = \mathbb{R}$

Examples of events:

$$A := [30, \infty)$$

$$B := \{35\}$$

$$C := \mathbb{R}$$

If temperature is 40 degrees, which events have occurred?

# Probability measure

The probability of an event quantifies how likely it is

(A function of a set is called a measure)

**Intuitive definition:** If we repeat the experiment many times

$$P(\text{event}) = \frac{\text{number of times event occurs}}{\text{total repetitions}}$$

# Events and probability measures

Key question:

What events should we assign probabilities to?

## Sample space

This is the event that **any** outcome occurs

From our intuitive definition:

$$\begin{aligned} P(\Omega) &= \frac{\text{outcomes in } \Omega}{\text{total}} \\ &= \frac{\text{total}}{\text{total}} \\ &= 1 \end{aligned}$$

## Union and intersection of events

If we assign probabilities to  $A$  and  $B$ , we should assign a probability to

►  $A \cup B$  (  $A$  or  $B$  happen )

►  $A \cap B$  (  $A$  and  $B$  happen )



## Union of disjoint events

Disjoint events don't share any common outcomes

From our intuitive definition:

$$\begin{aligned}P(D_1 \cup D_2) &= \frac{\text{outcomes in } D_1 \text{ or } D_2}{\text{total}} \\&= \frac{\text{outcomes in } D_1 + \text{outcomes in } D_2}{\text{total}} \\&= \frac{\text{outcomes in } D_1}{\text{total}} + \frac{\text{outcomes in } D_2}{\text{total}} \\&= P(D_1) + P(D_2)\end{aligned}$$

# Union of non-disjoint events

Non-disjoint events share some common outcomes

From our intuitive definition:

$$\begin{aligned}P(E_1 \cup E_2) &= \frac{\text{outcomes in } E_1 \text{ or } E_2}{\text{total}} \\&= \frac{\text{outcomes in } E_1 + \text{outcomes in } E_2 - \text{outcomes in } E_1 \text{ and } E_2}{\text{total}} \\&= \frac{\text{outcomes in } E_1}{\text{total}} + \frac{\text{outcomes in } E_2}{\text{total}} - \frac{\text{outcomes in } E_1 \text{ and } E_2}{\text{total}} \\&= P(E_1) + P(E_2) - P(E_1 \cap E_2)\end{aligned}$$

$$P(E_1 \cap E_2) = P(E_1) + P(E_2) - P(E_1 \cup E_2)$$

## Complement of an event

If we assign a probability to  $A$ , we should also assign a probability to its complement  $A^c$

This is the probability that  $A$  does not happen

What should  $P(A^c)$  equal as a function of  $P(A)$ ?

(Hint:  $A \cup A^c = \Omega$ )

$$1 = P(\Omega) = P(A \cup A^c) = P(A) + P(A^c)$$

$$\text{so } P(A^c) = 1 - P(A)$$

This is enough!

We are ready to define probability spaces mathematically

## Collection of events

Given a sample space  $\Omega$ , we assign probabilities to a collection  $\mathcal{C}$  of events that satisfies:

1.  $\Omega \in \mathcal{C}$
2. If an event  $A \in \mathcal{C}$  then  $A^c \in \mathcal{C}$
3. If the events  $A, B \in \mathcal{C}$ , then  $A \cup B \in \mathcal{C}$

If a countably infinite sequence  $A_1, A_2, A_3, \dots \in \mathcal{C}$   
then  $\cup_{i=1}^{\infty} A_i \in \mathcal{C}$

Collections with these properties are called  $\sigma$ -algebras

## Collection of events

If  $A$  and  $B$  are in the collection, what about  $A \cap B$ ?

De Morgan's law:  $A \cap B = (A^c \cup B^c)^c$

## Rolling a six-sided die

$$\Omega := \{1, 2, 3, 4, 5, 6\}$$

Collection of events if we want to include  $\{1\}, \{2\}, \dots, \{6\}$ ?

All possible subsets of  $\Omega$  ( $2^6 = 64$  events)

Smaller collections can also be valid

Smallest collection that contains  $\{2\}$ ?

$$\{\Omega, \emptyset, \{2\}, \{1, 3, 4, 5, 6\}\}$$

# Probability measure

Function mapping events in the collection to probabilities

1.  $P(A) \geq 0$  for any event  $A \in \mathcal{C}$
2.  $P(\Omega) = 1$
3. If  $A, B \in \mathcal{C}$  are disjoint then

$$P(A \cup B) = P(A) + P(B)$$

For countably infinite sequences of disjoint sets:  $A_1, A_2, A_3, \dots \in \mathcal{C}$

$$P\left(\lim_{n \rightarrow \infty} \cup_{i=1}^n A_i\right) = \lim_{n \rightarrow \infty} \sum_{i=1}^n P(A_i)$$



## Consequences of definition

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

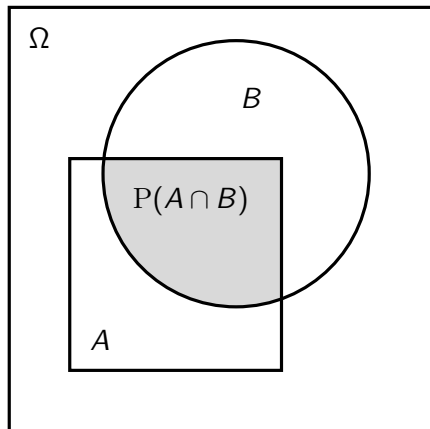
$$P(E_1 \cup E_2) = P(E_1) + P(E_2) - P(E_1 \cap E_2)$$

## Analogy with other measures

Mass, length, area or volume satisfy similar properties

We can use Venn diagrams to represent events

## Venn diagram



## Rolling a six-sided die

$$\Omega := \{1, 2, 3, 4, 5, 6\}$$

Collection: All possible subsets of  $\Omega$

Probability measure?

We need to assign **consistent** probabilities to all events...

**Idea:** Divide  $\Omega$  into smallest possible *components* and assign probabilities to them

# Partition

$A_1, A_2, \dots \in \mathcal{C}$  is a **partition** of  $\Omega$  if

- ▶  $A_i$  and  $A_j$  are disjoint for  $i \neq j$
- ▶  $\Omega = \cup_i A_i$

## Rolling a six-sided die

We assign probabilities to the partition  $\{1\}, \{2\}, \dots, \{6\}$

$$P(\{i\}) = \theta_i \text{ for } 1 \leq i \leq 6$$

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What conditions should  $\theta_1, \theta_2, \dots, \theta_6$  satisfy?

Nonnegative and

$$\begin{aligned} \sum_{i=1}^6 \theta_i &= P(\cup_{i=1}^6 \{i\}) \\ &= P(\Omega) \\ &= 1 \end{aligned}$$

$$P(\{i\}) = \theta_i \text{ for } 1 \leq i \leq 6$$

What about the rest of events in the collection?

$$\begin{aligned} P(\{2, 4, 6\}) &= P(\{2\}) + P(\{4\}) + P(\{6\}) \\ &= \theta_2 + \theta_4 + \theta_6 \end{aligned}$$



# What have we learned?

A probability space consists of

- ▶ A **sample space**  $\Omega$  containing all possible outcomes
- ▶ A **collection** of events  $\mathcal{C}$
- ▶ A **probability measure**  $P$  that assigns probabilities to the events in the collection

This sounds very complicated...

We never do this, instead we use **random variables**