

Overview of Probability

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>

Motivation

Describing uncertain phenomena

Who will win the next presidential election?

How much will a certain stock cost tomorrow?

Will the New York Knicks win the next NBA championship?

Strategy

Interpret uncertain phenomenon as an experiment that can be repeated over and over

$$P(\text{Knicks win}) = \frac{\text{times Knicks win}}{\text{total repetitions}}$$

Plan

- ▶ Probability spaces
- ▶ Conditional probability
- ▶ Estimating probabilities from data
- ▶ Independence
- ▶ Conditional independence
- ▶ The Monte Carlo method

Probability space

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

Six-sided die

Set of outcomes (sample space) $\Omega = \{1, 2, 3, 4, 5, 6\}$

Examples of events:

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

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

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

Probability measure

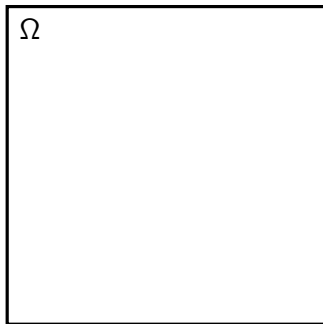
Assigns probability to each event

Intuitive definition: If we repeat the experiment many times

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

Rule 1

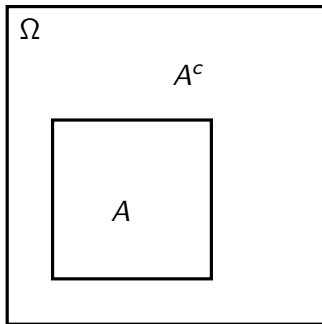
We must assign a probability to the whole sample space Ω



$$P(\text{sample space}) = 1$$

Rule 2

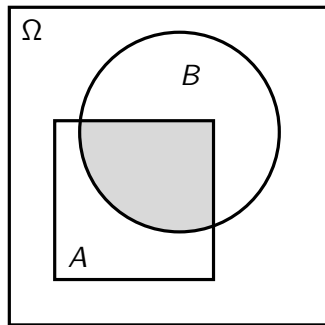
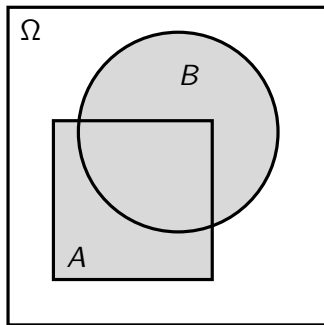
If we assign a probability to A then we need to assign a probability to A^c



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

Rule 3

If we assign a probability to A and B we need to assign a probability to $A \cup B$ and $A \cap B$



If A and B are **disjoint** $P(A \cup B) = P(A) + P(B)$

Important

Probability spaces might look very abstract, but they just implement our intuitive definition of probability

Conditional probability

Imagine that Knicks are first seed in playoffs

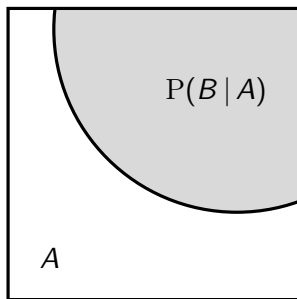
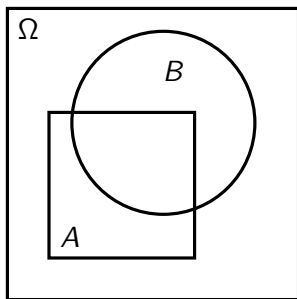
We need to **update** the probability that they win

$$P(\text{Knicks win} \mid \text{Knicks are 1st seed}) = \frac{\text{times Knicks are 1st seed and win}}{\text{times Knicks are 1st seed}}$$

Conditional probability

The conditional probability of an event B given A is

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

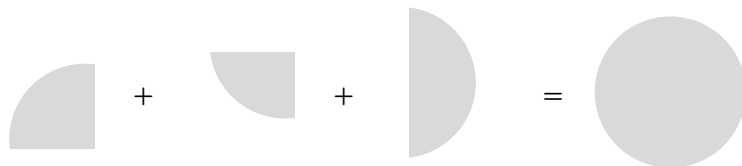
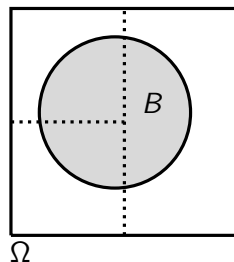
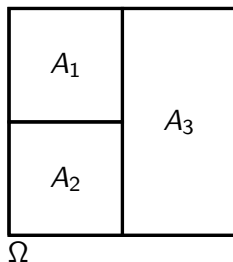


Chain rule

From definition of conditional probability $P(B|A) := \frac{P(B \cap A)}{P(A)}$

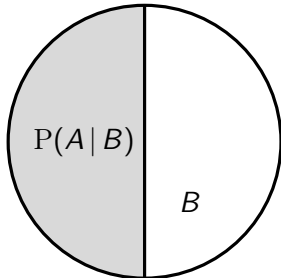
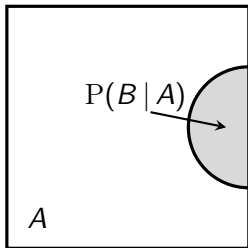
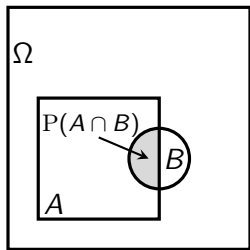
$$P(A \cap B) = P(A) P(B|A)$$

Law of Total Probability



$$P(A_1 \cap B) + P(A_2 \cap B) + P(A_3 \cap B) = P(B)$$

$$P(A|B) \neq P(B|A)$$



Bayes' Rule

For any events A and B

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

as long as $P(B) > 0$

Great, but in practice

How do we estimate estimate probabilities from data?

Intuitive definition of probability

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

Six-sided die

Data collection: We roll the die 60 times and observe 8 twos

Probability of event *Rolling a two*?

Empirical probability

Let A be an event and $X := \{x_1, x_2, \dots, x_n\}$ a dataset

The empirical probability of A is

$$P_X(A) := \frac{\sum_{i=1}^n 1_{x_i \in A}}{n}$$

where $1_{x_i \in A}$ is one if $x_i \in A$ and zero otherwise

Conditional probability

Data collection: We roll the die 60 times and observe 8 twos, 6 fours and 10 sixes

Conditional probability of event *Rolling a two* given *Roll is even*?

Empirical conditional probability

Let A and B be events, and $X := \{x_1, x_2, \dots, x_n\}$ a dataset

The empirical conditional probability of B given A is

$$P_X(B | A) := \frac{\sum_{i=1}^n 1_{x_i \in A \cap B}}{\sum_{i=1}^n 1_{x_i \in A}}$$

where $1_{x_i \in S}$ is one if $x_i \in S$ and zero otherwise

Independence of two events

Two events A, B are independent if

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

or equivalently

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

Multiple events

If A , B and C are pairwise independent, then

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

Independence of multiple events

The events $A_1, A_2, \dots, A_n \in \mathcal{F}$ are **mutually independent** if and only if for any $\{i_1, i_2, \dots, i_m\} \subseteq \{1, 2, \dots, n\}$

$$\mathbb{P} \left(\bigcap_{j=1}^m A_{i_j} \right) = \prod_{j=1}^m \mathbb{P} (A_{i_j})$$

Conditional independence

A, B are **conditionally** independent **given C** if

$$P(A | B, C) = P(A | C)$$

or equivalently

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

Independence and conditional independence

Does independence imply conditional independence? No!

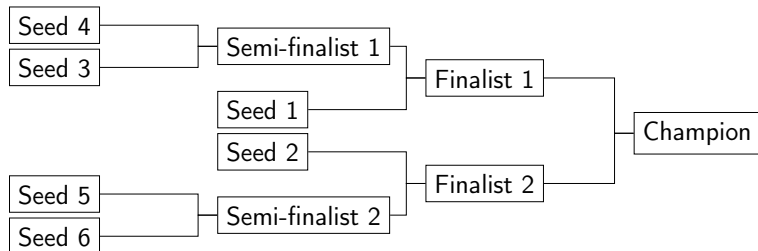
Does conditional independence imply independence? No!

Life is not a homework problem

In practice, we often cannot compute probabilities exactly, even when we have all the necessary information!

Tournament

Group stage followed by bracket



Intuitive definition of probability

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

Monte Carlo method

To approximate the probability of an event A , we

1. Generate n simulated outcomes: s_1, s_2, \dots, s_n
2. Compute the fraction of the outcomes in A ,

$$P_{\text{MC}}(A) := \frac{\sum_{i=1}^n 1_{s_i \in A}}{n}$$

where $1_{x_i \in A}$ is one if $s_i \in A$ and zero otherwise

Summary

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- ▶ Conditional probability
- ▶ Estimating probabilities from data
- ▶ Independence
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- ▶ The Monte Carlo method