

Section 2: Probability

STA 35C – Statistical Data Science III

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Fall Quarter 2025 (Sep 24 – Dec 12)
MWF, 12:10 PM – 1:00 PM, Olson 158
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Based on Chapter 1 of textbook: <https://www.probabilitycourse.com/>

1 Introduction

2 Conditional probability

- Independence

Introduction

Probability measure: introduction

Probability is a way to quantify randomness and/or uncertainty.

- e.g., coin flips, dice rolls, stocks, weather.
- Rules of probability should be intuitive and self-consistent.
- Self-consistent: the rules shouldn't lead to contradictions.
- Thus these rules must be constructed in a certain way.
- Suppose we want to assign a probability to each event in a set of possible events.
- We would like, at the very least:
 1. each probability to be a value between 0 and 1 (inclusive)
 2. the probability assigned to the full set of events to be 1
 - close to 1 \Rightarrow very likely that A occurs.
 3. the probability assigned to the empty set to be 0
 - close to 0 \Rightarrow very unlikely that A occurs.
- We need more restrictions to ensure self-consistency.

The following definition will lead to intuitive and self-consistent rules of probability.

- We assign a *probability* measure $P(A)$ to an event A.

Definition 1: Probability measure $P(\cdot)$

For a nonempty sample space Ω , the set function $P: \Omega \rightarrow [0, 1]$ is a **probability measure**, if

- $P(\Omega) = 1$,
- for any pairwise disjoint events $A_1, A_2, A_3, \dots \subset \Omega$ (i.e. $A_i \cap A_j = \emptyset$ for all i, j with $i \neq j$), holds:

$$P(A_1 \cup A_2 \cup A_3 \cup \dots) = P(A_1) + P(A_2) + P(A_3) + \dots \quad (1)$$

This definition fulfills the three desirable properties:

- $P(\Omega) = 1$: the probability of the biggest possible set is equal to 1.
- Property (1) – called the **countable additivity** property – allows us to add probabilities of disjoint sets.

Given a random experiment with a sample space Ω , how do we find the probability of an event of interest? Use:

- the specific information that we have about the random experiment.
- the probability rules induced by Definition 1.

Finding probabilities: example

Example: Roll a fair four-sided die. What is the probability of $E = \{1, 3\}$?

- Information about experiment (fair die): $P(\{1\}) = P(\{2\}) = P(\{3\}) = P(\{4\})$.
- Probability rules:

$$\begin{aligned}1 &= P(S) \\&= P(\{1\} \cup \{2\} \cup \{3\} \cup \{4\}) \\&= P(\{1\}) + P(\{2\}) + P(\{3\}) + P(\{4\}) \\&= 4P(\{1\}).\end{aligned}$$

Thus $P(\{1\}) = P(\{2\}) = P(\{3\}) = P(\{4\}) = \frac{1}{4}$. Finally,

$$P(E) = P(\{1, 3\}) = P(\{1\}) + P(\{3\}) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}.$$

Annoying to write e.g., $P(\{2\})$

- Simplify to $P(2)$
- But always keep in mind that P is a function on sets, not on individual outcomes.

Finding probabilities: more tools

Definition 1 implies the following additional properties:

Properties of $P(\cdot)$

Given a sample space Ω and arbitrary events $A, B \subset \Omega$, Definition 1 implies

1. $P(\emptyset) = 0$
2. $P(A^c) = 1 - P(A)$
3. $P(A \cup B) = P(A) + P(B) - P(A \cap B)$
4. $P(B \setminus A) = P(B) - P(A \cap B)$
5. $P(A) \leq P(B)$ if $A \subset B$.

(Pictures for intuition; for formal proofs, see “Example 1.10” in §1.3.3 of textbook)

Finding probabilities: example

Suppose we have the following information:

1. There is a 60 percent chance that it will rain today. $A = \text{event of rain today}$
2. There is a 50 percent chance that it will rain tomorrow. $B = \text{event of rain tomorrow}$
3. There is a 30 percent chance that it does not rain either day. $(A \cup B)^c$

Find the following probabilities:

- a. The probability that it will rain today or tomorrow.
- b. The probability that it will rain today and tomorrow.
- c. The probability that it will rain today but not tomorrow.
- d. The probability that it either will rain today or tomorrow, but not both.



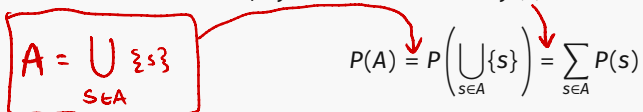
$$\begin{aligned} \text{a. } P(A \cup B) &= 1 - P(\text{does not rain either day}) = 1 - 0.3 = 0.7 \\ \text{b. } P(A \cap B) &= P(A) + P(B) - P(A \cup B) = 0.6 + 0.5 - 0.7 = 0.4 \\ \text{c. } P(A \cap B^c) &= P(A \setminus B) = P(A) - P(A \cap B) = 0.6 - 0.4 = 0.2 \\ \text{d. } P(A \cup B) - P(A \cap B) &= 0.7 - 0.4 = 0.3 \end{aligned}$$

Distinguish between two different types of sample spaces: *discrete* and *continuous*.

- Will discuss in more detail in Section 3 of the course.
- Discrete: can compute the probability of an event by adding all outcomes in the event.
- Continuous: need to use integration instead of summation.

If a sample space Ω is a countable set, this refers to a *discrete* probability model.

- Can list all elements: $\Omega = \{s_1, s_2, s_3, \dots\}$.
- For an event $A \subset \Omega$, by countable additivity (1) we can write


$$A = \bigcup_{s \in A} \{s\} \qquad P(A) = P\left(\bigcup_{s \in A} \{s\}\right) = \sum_{s \in A} P(s)$$

Thus, to find probability of an event, just need to sum the probability of individual elements in that event.

Probability models: discrete (example)

Consider a gambling game: win $k - 2$ dollars with probability $\frac{1}{2^k}$ for any $k \in \mathbb{N}$.

- What is the probability of winning at least \$1 and less than \$4?
- What is the probability of winning more than \$1?

$$\sum_{k=1}^{\infty} \frac{1}{2^k} = 1$$

| k | 1 | 2 | 3 | 4 | 5 | ... |
|---------|---------------|---------------|---------------|----------------|----------------|-----|
| winings | -1 | 0 | 1 | 2 | 3 | ... |
| prob | $\frac{1}{2}$ | $\frac{1}{4}$ | $\frac{1}{8}$ | $\frac{1}{16}$ | $\frac{1}{32}$ | ... |

Let X = amount of winnings

$$\begin{aligned} \bullet P(1 \leq X < 4) &= P(\{X=1\} \cup \{X=2\} \cup \{X=3\}) \\ &= P(X=1) + P(X=2) + P(X=3) \\ &= \frac{1}{8} + \frac{1}{16} + \frac{1}{32} = \boxed{\frac{7}{32}} \\ \bullet P(X > 1) &= 1 - P(X \leq 1) = 1 - \left[\frac{1}{2} + \frac{1}{4} + \frac{1}{8} \right] = \boxed{\frac{1}{8}} \end{aligned}$$

Probability models: discrete (equally likely outcomes)

Important special case: finite sample space Ω where each outcome is equally likely.

- Thus for any outcome $s \in \Omega$, we must have

$$P(s) = \frac{1}{|\Omega|}.$$

- In such a case, for any event A , we can write

$$P(A) = \sum_{s \in A} P(s) = \sum_{s \in A} \frac{1}{|\Omega|} = \frac{|A|}{|\Omega|}.$$

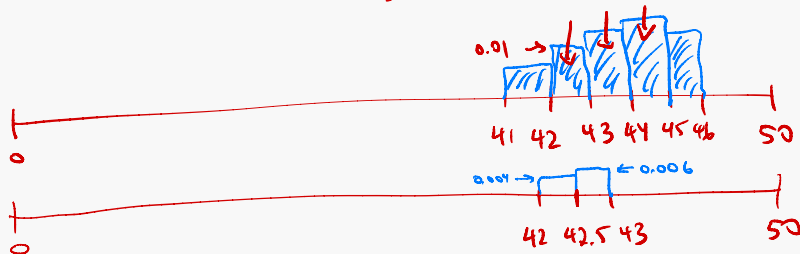
Probability models: continuous

Consider a sample space Ω that is an *uncountable* set.

- E.g., a 50-minute exam (so $\Omega = [0, 50]$), and let T_{Ant} be the time it takes Ant to finish the exam.
- What is the probability of $T_{Ant} \in [42.5, 45)$?

If uniform assumption on finishing time,

$$\text{then } P(T_{Ant} \in [42.5, 45)) = \frac{45 - 42.5}{50 - 0} = 0.05$$



Conditional probability

Introduction

As you obtain *additional information*, how should you update probabilities of events?

- For example, suppose I roll a fair die.
- Let $A = \{1, 3, 5\}$. What is the probability that the outcome is in A ?
We will write this as $P(A)$.

$$\underline{P(A)} = \frac{|A|}{|S|} = \frac{3}{6}$$

- Let $B = \{1, 2, 3\}$. What is the probability of A if I know that the outcome is in B ?
We will write this as $P(A|B)$.

$$\underline{P(A|B)} = \frac{|A \cap B|}{|B|} = \frac{|\{1, 3\}|}{3} = \frac{2}{3}$$

In the previous example, we call...

- ... $P(A)$ the *prior probability* of A ;
- ... $P(A|B)$ the *conditional probability of A given that B has occurred*.
 - ▶ Usually shortened to the *conditional probability of A given B* .

The way we obtained $P(A|B)$ in this example can be generalized by the following definition.

Definition

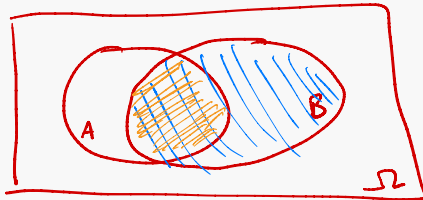
Definition 2: Conditional probability

If A and B are two events in a sample space Ω , then the **conditional probability of A given B** is defined as

$$\underline{P(A|B)} = \frac{P(A \cap B)}{\underline{P(B)}}, \text{ when } \underline{P(B) > 0}. \quad (2)$$

If we know that B has occurred, then we can discard every outcome outside of B .

- In other words, our sample space is reduced to the set B . (Picture)



- We divide $P(A \cap B)$ by $P(B)$ so that the conditional probability of the new sample space B becomes 1. $P(B|B) = 1$
- $P(A|B)$ is undefined when $P(B) = 0$ (meaning B never occurs).

Conditional probability rules

Conditional probability itself is a probability measure.

- So all probability rules learned so far can be extended to conditional probability.

For example, Definition 1 (slide 3) and other properties (slide 7)

If conditioning on event C , then slide 7 becomes

$$1. P(\emptyset | C) = 0$$

$$2. P(A^c | C) = 1 - P(A | C)$$

$$3. P(A \cup B | C) = P(A | C) + P(B | C) - P(A \cap B | C)$$

Important special cases

Plug into Definition 2

- When A and B are disjoint:

$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{0}{P(B)} = 0$$

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lecture

- When B is a subset of A :

- When A is a subset of B :

Example

I roll a fair die twice. Let

- X_1 be the result of the first roll;
- X_2 be the result of the second roll;

Given that I know $X_1 + X_2 = 7$, what is the probability that $X_1 = 4$ or $X_2 = 4$?

Chain rule

We can rearrange the formula in Definition 2 as

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

We can generalize this to 3 events:

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

We can generalize this to $n \geq 2$ events (chain rule for conditional probability):

$$P(A_1 \cap A_2 \cap \cdots \cap A_n) = P(A_1) P(A_2|A_1) P(A_3|A_2, A_1) \cdots P(A_n|A_{n-1}, A_{n-2} \cdots A_1)$$

Example

Of 100 units of a certain product, 5 are defective. If we pick three of the 100 units at random, what is the probability that none of them are defective?

Conditional probability

Independence

Let A be the event that it rains tomorrow. Let B be the event that the coin I toss (indoors) tomorrow lands heads up.

- Guess that the result of the coin toss has nothing to do with tomorrow's weather.
- The probability of A should not depend on whether or not B happens.
- Two events are independent if one does not convey any info about the other.

Definition 1: Independent events

Two events A and B are *independent* if and only if

$$P(A \cap B) = P(A)P(B). \quad (3)$$

If two events A and B are independent and $P(B) \neq 0$, then

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

- “Independence means that conditional probability of one event given another is the same as the original (prior) probability.”
- Sometimes it is obvious if two given events are independent or not.
- Other times, we need need to check if they satisfy the independence condition (3).

Example

I pick a random number from $\{1, 2, 3, \dots, 10\}$, and call it N .

- Suppose that all outcomes are equally likely.
- Let A be the event that $N < 7$, and let B be the event that N is even.

Are A and B independent?

Independence does not mean disjoint! Disjoint does not mean independence!

Independence \neq disjoint

- Two *independent* events convey no information about the other.
- Two *disjoint* events cannot occur at the same time.