Abstract Definition of Probability 1

Probability is the area of mathematics concerned with describing uncertain or random events. We will develop a mathematical framework that will allow us to quantify uncertainty in a principled way.

1.1 Axioms of Probability

The fundamental object is the probability space $(\Omega, \mathcal{F}, \mathbb{P})$, which specifies the space of outcomes Ω , the space of events \mathcal{F} , and the likelihoods of events \mathbb{P} .

Definition 1 (Probability Space). A sample space Ω is the set of all possible outcomes of a random process. The elements of $\omega \in \Omega$ are called *outcomes* and the subsets $A \subseteq \Omega$ are called *events*. The associated probability measure \mathbb{P} encodes the probability of each event occurring.

Definition 2 (Probability Measure). Let \mathcal{F} denote the set of all subsets of Ω that we can assign probabilities to. A probability measure is a function from $\mathcal{F} \to \mathbb{R}_+$ such that

- 1. Normalization: $\mathbb{P}(\Omega) = 1$
- 2. Non-Negativity: $\mathbb{P}(A) \geq 0$ for all $A \in \mathcal{F}$.
- 3. Countable Additivity: If $A_1, A_2, ... \in \mathcal{F}$ are disjoint $(A_i \cap A_j = \emptyset \text{ for } i \neq j)$, then

$$\mathbb{P}\left(\bigsqcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} \mathbb{P}(A_i).$$

From these three properties, we can recover all the natural properties a probability should satisfy.

Proposition 1 (Properties of a Probability Measure)

Any probability measure \mathbb{P} satisfies the following

- 2. $\mathbb{P}(A^c)=1-\mathbb{P}(A),$ 3. Monotonicity: If $A\subseteq B$, then $\mathbb{P}(A)\leq \mathbb{P}(B),$
- 4. Inclusion–Exclusion Principle: $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) \mathbb{P}(A \cap B)$,
- 5. Union Bound: $\mathbb{P}(A \cup B) \leq \mathbb{P}(A) + \mathbb{P}(B)$.

1.1.1 Discrete Probability Spaces

If Ω is countable, then we can always assign a probability to every outcome. Thus a probability measure is completely determined by the probabilities of each individual outcome.

Corollary 1

Let $\Omega = \{\omega_1, \omega_2, \dots\}$ and let $A \subset \Omega$ be an event. Then

$$\mathbb{P}(A) = \sum_{\omega_i \in A} \mathbb{P}(\omega_i).$$

Remark 1. On the other hand, if Ω is uncountable, then it is impossible to assign a probability to every outcome, so it is necessary to define probabilities on the set of measurable events \mathcal{F} .

A special case of a discrete probability space is the one with *equally likely outcomes*. This is often the most naive definition of a probability space. We will encounter much richer probability spaces throughout this course, since outcomes might not always be equally likely.

Definition 3 (Uniform Probability Measure). If all outcomes are equally likely, then the associated probability measure \mathbb{P} is called the *uniform probability measure* and

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

1.2 Example Problems

Problem 1.1. Suppose two six sided dice are rolled, and the number of dots facing up on each die is recorded.

- 1. Write down the sample space Ω .
- 2. Write down, as a set, the event A = "The sum of the dots is 7".
- 3. Write down, as a set, the event B^c , where B = "The sum of the numbers is at least 4".
- 4. Write down, as a set, the events $A \cap B^c$ and $A \cup B^c$.

Solution 1.1.

1. The sample space for a pair of dice is the a pair of the outcomes of each die roll

$$\Omega = \{1, \dots, 6\} \times \{1, \dots, 6\} = \{(x, y) : x, y \in \{1, 2, \dots, 6\}\}.$$

2. We can simply write down all the combinations

$$A = \{(1,6), (2,5), (3,4), (4,3), (5,2), (6,1)\}.$$

3. If $B = \{\text{sum is at least 4}\}\ \text{then } B^c = \{\text{sum is at most 3}\}\$, so

$$B^c = \{(1,1), (1,2), (2,1)\}.$$

4. Since it is impossible for the sum of dots to be 7 and at most 3 at the same time, $A \cap B^c = \emptyset$. All the possibilities the sum of dots is 7 or at most 3 is

$$A \cup B^c = \{\underbrace{(1,6),(2,5),(3,4),(4,3),(5,2),(6,1)}_{A},\underbrace{(1,1),(1,2),(2,1)}_{B^c}\}.$$

Problem 1.2. For the following experiments, describe a possible sample space Ω .

- 1. Roll a die.
- 2. Number of coin-flips until heads occurs.
- 3. Waiting time in minutes (with infinite precision, e.g., $0.2384\overline{45}$ minutes) until a task is complete.

Solution 1.2.

- 1. There are many ways we can record the outcome of a die such that no elements can occur at the same time $\Omega = \{1, 2, 3, 4, 5, 6\}$ or $\Omega = \{\text{even}, \text{odd}\}$. The choice of the best sample space will depend on the application in mind, but usually the coarsest choice is the most powerful.
- 2. There is only one natural choice here $\Omega = \{1, 2, 3, ...\} = \mathbb{N}$.
- 3. There is only one natural choice here $\Omega = [0, \infty) = \{x \in \mathbb{R} : x > 0\}.$

Problem 1.3. Suppose that two fair six sided die are rolled.

- 1. What is the probability that the dots on each die match?
- 2. What is the probability that the dots sum to 7?
- 3. What is the probability that the dots do not sum to 7?
- 4. What is the probability that the dots match and sum to 7?

Solution 1.3. The probability is uniform over the sample space $\Omega = \{1, \dots, 6\}^2$. All outcomes are equally likely, so for an event A,

$$\mathbb{P}(A) = \frac{|A|}{|\Omega|} = \frac{|A|}{36}.$$

- 1. The event is $A = \{(1, 1), (2, 2), \dots, (6, 6)\}$ with |A| = 6, so $\mathbb{P}(A) = 6/36 = 1/6$.
- 2. The event is $B = \{(1,6), (2,5), \dots, (6,1)\}$ with |B| = 6, so $\mathbb{P}(B) = 6/36 = 1/6$.
- 3. The event is B^c with $|B^c| = |S| |B| = 30$ elements, hence $\mathbb{P}(B^c) = 30/36 = 5/6 = 1 \mathbb{P}(B)$.
- 4. 7 is an odd number so it is impossible for the dots to match. Therefore, $\mathbb{P}(\emptyset) = 0$.

1.3 Proofs of Key Results

Problem 1.4. (Proposition 1) Show the *monotonicity* property of probability,

if
$$A \subseteq B$$
 then $\mathbb{P}(A) < \mathbb{P}(B)$.

Solution 1.4. This follows directly from the axioms. If $A \subseteq B$, then $B = A \cup A \setminus B$ and the sets A and $A \setminus B$ are disjoint. Therefore, by countable additivity,

$$\mathbb{P}(B) = \mathbb{P}(A) + \mathbb{P}(A \setminus B) \ge \mathbb{P}(A)$$

since $\mathbb{P}(A \setminus B) \geq 0$ by the non-negativity property.

Problem 1.5. (Proposition 1) Show that the axiomatic defintiion of a probability implies that

$$0 \le \mathbb{P}(A) \le 1$$

for any event A.

Solution 1.5. Suppose for the sake of contradiction that $\mathbb{P}(A) > 1$ for some event A. By the monontonicity property, since $A \subseteq \Omega$,

$$\mathbb{P}(\Omega) \ge \mathbb{P}(A) > 1$$

which contradicts the fact that $\mathbb{P}(\Omega) = 1$. Therefore, $\mathbb{P}(A) \leq 1$.

Problem 1.6. (Proposition 1) Show that the axiomatic definition of a probability implies that

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

for any event A.

Solution 1.6. Notice that $A \cup A^c = \Omega$ and A and A^c are disjoint. From finite additivity, we conclude that

$$\mathbb{P}(A) + \mathbb{P}(A^c) = \mathbb{P}(\Omega) = 1 \implies \mathbb{P}(A) = 1 - \mathbb{P}(A^c).$$

Problem 1.7. (Proposition 1) Show that the axiomatic definition of a probability implies that

$$\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A \cap B),$$

for any event A and B.

Solution 1.7. Notice that A and $(B \setminus A)$ are disjoint events such that $A \cup (B \setminus A) = A \cup B$, so by countable additivity

$$\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B \setminus A)$$

Next, notice that $B \setminus A$ and $A \cap B$ are exclusive and $(A \cap B) \cup (B \setminus A) = A \cap B$,

$$\mathbb{P}(B) = \mathbb{P}(A \cap B) + \mathbb{P}(B \setminus A) \implies \mathbb{P}(B \setminus A) = \mathbb{P}(B) - \mathbb{P}(A \cap B)$$

which implies our result.

Problem 1.8. (Proposition 1) Show that the axiomatic definition of a probability implies that

$$\mathbb{P}(A \cup B) \le \mathbb{P}(A) + \mathbb{P}(B),$$

for any event A and B.

Solution 1.8. By the inclusion–exclusion property,

$$\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A \cap B) < \mathbb{P}(A) + \mathbb{P}(B)$$

since $\mathbb{P}(A \cap B) \geq 0$. Notice that this implies that the union bound is sharp and is attained when A and B are disjoint sets, which is implied by countable additivity.