# **Probability Theory**

Raymond Bian

January 16, 2024

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### Lecture 1: Intro to Probability

# 1 Basics of Probability

What data do you need to specify probability? You need the **set of all outcomes**, a list of everything that could possibly occur as a consequence, and the likelihood of each event.

**Example.** For a roll of a dice, the set of all outcomes would be  $\{1,2,3,4,5,6\}$ . The list could include things like "the result is 3", or "the result is  $\geq 4$ ", and the likelihood would be  $\frac{1}{6}$  for each of the results.

### 1.1 Basics of Set Theory

**Definition 1.** A **set** is an unordered collection of elements. **Elements** are objects within sets.

**Definition 2.** A set *A* is a **subset** of a set *B* if  $a \in A \Rightarrow a \in B$ 

**Definition 3.** The **union** of two sets A and B is the collection of elements that are in A or B.

**Definition 4.** The **intersection** of two sets A and B is the collection of elements that are in both A and B.

**Definition 5.** The **complement** of a set A is everything not in A.

**Definition 6.** A **finite set** is a set with finite number of elements.

**Definition 7.** The **cartesian** product of two sets A and B denoted  $A \times B$  is

$$\{(a, b): a \in A \land b \in B\}.$$

Then,  $|A \times B| = |A| \cdot |B|$ .

## 1.2 Back to Probability

**Definition 8.** A **sample space** is the set of al possible outcomes in an experiment.

**Example.** The sample space  $\Omega$  for a coin flip is  $\{H, T\}$ .

Note that **events** are just subsets of the sample space, and **elementary events** are just elements of the sample space.

**Example.** For a dice roll:  $\Omega = \{1, 2, 3, 4, 5, 6\}$ , some events could be  $\{1, 2\}$ ,  $\{3, 6\}$ ,  $\{3\}$ . There are a total of  $2^6$  events.

**Definition 9.** If  $\Omega$  is a finite set, a probability P on  $\Omega$  is a function:  $P \colon 2^{\Omega} \to [0,1]$  such that  $\mathbb{P}(\varnothing) = 0$  and  $\mathbb{P}(\Omega) = 1$ .

**Lemma 1.** If  $A_1, \ldots, A_{\alpha} \subset \Omega$  are disjoint,  $\mathbb{P}(\bigcup_i A_i) = \sum_i \mathbb{P}(A_i)$ .

**Proposition 1.** Let  $A = \{a_1, a_2, \dots a_l\}$  such that  $a_i$  are elementary events. Then,

$$\mathbb{P}(A) = \sum_{i=1}^{l} \mathbb{P}(\{a_i\}).$$

**Example.** For the dice roll, if  $A = \{1, 3, 5\}$ , then  $\mathbb{P}(A) = 3 \cdot \frac{1}{6} = \frac{1}{2}$ .

**Definition 10. Equiprobable outcomes**: Let's say we have the set  $\Omega = \{\omega_1, \dots, \omega_N\}$  and  $\mathbb{P}(\omega_i) = \mathbb{P}(\omega_j)$  for all i and j. Then,  $\mathbb{P}(\omega) = \frac{1}{N}$  for all  $\omega \in \Omega$  and  $\mathbb{P}(A) = \frac{|A|}{N}$ . In other words, when outcomes are probable,

 $\mathbb{P}(\text{event}) = \frac{\text{number of outcomes for that event}}{\text{number of possible outcomes}}$ 

## 1.3 Counting

Suppose 2 experiments are being performed. Let's say that experiment 1 has m possible outcommes, and experiment 2 has n possible outcomes. Then together, there are total of  $n \cdot m$  total outcomes.

**Example.** Rolling a dice and then flipping a coin, how many possible outcomes are there?

**Proof.** You have  $6 \cdot 2 = 12$  outcomes.

**Example.** Let's say you have a college planning committee that consists of 3 freshman, 4 sophomores, 5 juniors, and 2 seniors. How many ways are there to select a subcommittee of 4 with one person from each grade?

**Proof.** There are 4 events with 3, 4, 5, and 2 possible outcomes for each. Therefore, there are  $3 \cdot 4 \cdot 5 \cdot 2 = 120$  total subcommittees.

**Example.** How many 7-place license plates are there if the first 3 are letters and the last 4 are numbers?

**Proof.** There are  $26^3 \cdot 10^4$  license plates.

**Definition 11.** A **permutation** is an ordering of elements in a set. The number of ways to order n elements is given by n!.

**Example.** Alex has a bunny ranch with 10 bunnies. They are going to run an obstacle course and ranked 1-10 based on completion time. How many possible rankings are there (no ties)?

**Proof.** There are 10! possible rankings.

**Example.** Assume 6 bunnies have straight ears and 4 have floppy ears. We rank the bunnies separately. How many possible rankings are there?

**Proof.** There are  $6! \cdot 4!$  possible outcomes.

**Definition 12.** A **combination** denotes the number of ways to choose k elements from n total elements (counting subsets).

**Example.** How many ways are there to pick a 2 person team from a set of 5 people?

**Proof.** There are  $C(5, 2) = {5 \choose 2} = {5! \over 2! \cdot 3!} = 10$  ways.

**Example.** How many committees consisiting of 2 women and 3 men can be formed from a group of 5 women and 7 men?

**Proof.** We have  $C(5,2) \cdot C(7,3)$  possible committees.

**Example.** What if two of the men do not want to serve on the committee together?

**Proof.** The number of ways to choose the women stays the same. However, for the men we must subtract the number of committees that have both men. Therefore, we have  $C(5,2)\cdot (C(7,3)-C(5,1))$  possible committees.

**Example.** How many ways can we divide a 10 person class into 3 groups, sizes 3, 3, and 4?

**Proof.** We just have 3 events, multiplying:  $C(10,3) \cdot C(7,3) \cdot C(4,4)$ .

**Definition 13.** This is known as a **multinomial**, and is given by

$$\binom{n}{n_1, n_2, \dots, n_r} = \frac{n!}{n_1! \cdot n_2! \cdot \dots \cdot n_r!}.$$

It counts the number of ways to partition a set of size n into sets of sizes  $n_1, n_2, \ldots, n_r$ .

### 1.4 Back to Probability Again

**Example.** Flip 10 fair coins. What is the likelihood of flipping 3 heads?

**Proof.** Number of events of 3 heads is C(10, 3). Total number of events is  $2^{10}$ . Therefore,

$$\mathbb{P}(10 \text{ heads}) = \frac{C(10,3)}{2^{10}}.$$

In general, we have  $\sum_{k=0}^{n} \mathbb{P}(k \text{ heads}) = 1$ . In **Lecture 2: More Probabiliv** other words.

$$\frac{1}{2^{10}} \cdot \sum_{k=0}^{10} \binom{10}{k} = 1.$$

such that

$$\sum_{k=0}^{10} \binom{10}{k} = 2^{10}.$$

More generally,

**Definition 14.** The **binomial theorem** states that for all  $x, y \in \mathbb{R}$ ,  $n \ge 1$ ,  $n \in \mathbb{N}$ ,

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}.$$

**Example.** Rolling 10 dice, what is the likelihood of exactly 2 outcomes each of 1,2,3,4, 1 outcome of 6, and 1 outcome of 5.

**Proof.** There are total  $6^{10}$  outcomes, and there are  $\binom{10}{2,2,2,2,1,1}$  desired outcomes. Therefore, the probability of this event is  $\frac{\binom{10}{2,2,2,2,1,1}}{6^{10}}$ 

**Definition 15.** The multinomial theorem states that  $(x_1 + \ldots + x_r)^n =$ 

$$\sum_{n_1+\ldots+n_r=n} \binom{n}{n_1,\ldots,n_r} x_1^{n_1} x_2^{n_2} \ldots x_r^{n_r}.$$

#### 1.5 Measure Theory

This is just a generalization of what we have seen before.

**Definition 16.** Let  $\mathcal{F} \subset 2^{\Omega}$  be an "event space". A mapping  $P: \mathcal{F} \to \mathbb{R}$  is a **probability measure** on  $(\Omega, \mathcal{F})$  if

- $\mathbb{P}(A) \geq 0 \quad \forall A \in \mathcal{F}$
- $\mathbb{P}(\emptyset) = 0, \mathbb{P}(\Omega) = 1$

If A<sub>1</sub>, A<sub>2</sub>, . . . are disjoint,

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

### 1.6 **Properties of Event Spaces**

**Definition 17.** A collection  $\mathcal{F}$  of subsets of the sample space  $\Omega$  is called an **event space** if

- $\bullet$   $\mathcal{F}$  is non-empty.
- if  $A \in \mathcal{F}$  then  $\Omega \setminus A \in \mathcal{F}$ .
- if  $A_1, A_2, \ldots \in \mathcal{F}$  then  $\bigcup_{i=1}^{\infty} A_i \in \mathcal{F}$ .

**Theorem 1.** If  $A \in \mathcal{F}$ , then  $\mathbb{P}(A) + \mathbb{P}(\Omega \setminus A) =$ 

**Proof.** Notice that A and  $\Omega \setminus A$  are disjoint. And, that  $A \cup (\Omega \setminus A) = \Omega$ . Then,

$$\mathbb{P}(A \cup (\Omega \setminus A)) = \mathbb{P}(\Omega) = 1.$$

**Theorem 2.** If  $A, B \in \mathcal{F}$  then  $\mathbb{P}(A \cup B) + \mathbb{P}(A \cap B)$  $B) = \mathbb{P}(A) + \mathbb{P}(B).$ 

**Proof.** Note that  $A \cup B = (A \setminus B) \cup (A \cap B)$  $(B \setminus A)$ . This is a union of disjoint sets, such that  $\mathbb{P}(A \cup B) = \mathbb{P}(A \setminus B) + \mathbb{P}(A \cap B) + \mathbb{P}(B \setminus A)$ . Then, we have  $\mathbb{P}(A \cup B) + \mathbb{P}(A \cap B) = \mathbb{P}(A \setminus B)$  $(B) + \mathbb{P}(A \cap B) + \mathbb{P}(B \setminus A) + \mathbb{P}(A \cap B)$ , of which the RHS simplifies to  $\mathbb{P}(A) + \mathbb{P}(B)$ .

**Theorem 3.** If  $A, B \in \mathcal{F}$ , and  $A \subseteq B$ , then  $\mathbb{P}(A) \leq \mathbb{P}(B)$ .

**Proof.** We wish to show  $A \subseteq B \Rightarrow \mathbb{P}(A) \leq$  $\mathbb{P}(B)$ . Then,  $B = (B \setminus A) \cup (B \cap A) = (B \setminus A) \cup A$ , such that  $\mathbb{P}(B) = \mathbb{P}(B \setminus A) + \mathbb{P}(A) \geq \mathbb{P}(A)$ because  $\mathbb{P}(B \setminus A) \geq 0$ .

#### 1.7 **Examples**

**Example.** What is the probability that one is dealt a full house?

**Proof.** This is the number of ways one can get a full house, divided by the total number of poker hands (5 card). The total number of poker hands is  $\binom{52}{5}$ . The number of full houses is  $\frac{52\cdot\binom{3}{2}\cdot48\cdot3}{2!3!}$ . Another way we can count the number of full houses is  $\binom{13}{1}\cdot\binom{4}{3}\cdot\binom{12}{1}\cdot\binom{4}{2}$ . The result of the division is our answer.

**Example.** A box contains 3 marbles, 1 red 1 green and 1 blue. Consider an experiment that cnsists of us taking 1 marble, replacing it, and drawing another marble. What is the sample space?

Proof.

$$\Omega = \{ (r, r), (r, b), (r, g), (b, r), (b, g), (b, b), (g, r), (g, g), (g, b) \}.$$

**Example.** What about if we don't replace the first marble?

**Proof.** Everything without (r, r), (b, b), (g, g).

**Example.** What is the probability of being dealt a flush?

**Proof.** This is just number of flushses divided by number of poker hands. The number of flushes is  $\binom{4}{1} \cdot \binom{13}{5}$ .

**Example.** What is the probability of being dealt a straight?

**Proof.** We can do the probability of any straight, minus probability of straight flush. The number of straights is 10 number-wise. Therefore, the number of straights is  $10 \cdot (4^5 - 4)$ . The probability can be then calculated.

**Example.** An urn contains n balls. If k balls are withdrawn one at a time, what is the probability that a special ball is chosen?

**Proof.**  $\mathbb{P}(\text{special}) = 1 - \mathbb{P}(\text{special}^c)$ . If the special ball is not chosen, it would be  $\frac{(n-1)!}{(n-k-1)!}$ . The total number of withdrawings is  $\frac{n!}{k!}$ . Then, the total probability is  $1 - \frac{n-k}{n}$ .

**Example.** If n people are present in a room, what is the prob that no two celebrate their birthday on the same date? How large must n be such that this probability is  $<\frac{1}{2}$ .

**Proof.**  $\mathbb{P}$ (no people with same birthday) is the number of no same birthday situations divided by the number of possibilities. Total possibilities is  $365^n$ . No same birthday situatons is  $\mathbb{P}(365, n) = \frac{365!}{(365-n)!}$ . For the second question, n = 23.

### 1.8 Conditional Probability

**Definition 18.** If  $A, B \in \mathcal{F}$  and  $\mathbb{P}(B) > 0$  then the **conditional probability** if A given B is denoted by  $\mathbb{P}(A \mid B)$  and defined by

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

**Theorem 4.** If  $B \in \mathcal{F}$  and  $\mathbb{P}(B) > 0$  then  $(\Omega, \mathcal{F}, \mathbb{Q})$  is a pobability space where  $\mathbb{Q} : \mathcal{F} \to \mathbb{R}$  is defined by  $\mathbb{Q}(A) = \mathbb{P}(A \mid B)$ 

**Example.** Let's say a coin is flipped twice. What is the conditional probability that both flips land on heads, given that the first flip lands on heads?

$$\begin{array}{lll} \textbf{Proof.} & \frac{\mathbb{P}(\mathsf{two \ heads} \cap \mathsf{first \ heads})}{\mathbb{P}(\mathsf{first \ heads})} & = & \frac{\mathbb{P}(\mathsf{two \ heads})}{\mathbb{P}(\mathsf{first \ heads})} \end{array}$$
 This is just  $\frac{1}{2}$ .

**Example.** What if given at least one lands on heads?

$$\begin{array}{c} \textbf{Proof.} \ \ \frac{\mathbb{P}(\mathsf{two \ heads} \cap \mathsf{at \ least \ one \ head})}{\mathbb{P}(\mathsf{at \ least \ one \ head})} = \\ \frac{\mathbb{P}(\mathsf{two \ heads})}{\mathbb{P}(\mathsf{at \ least \ one \ head})} = \frac{2}{3}. \end{array}$$

**Example.** In the card game bridge, the 52 cards are dealt equally. If North and South have a total of 8 spades among them, what is the probability that East has 3 of the 5 remaining spades?

**Proof.** No rule: 
$$\mathbb{P}(\mathsf{E} \text{ has 3 spades}) = \frac{\binom{5}{3} \cdot \binom{21}{10}}{\binom{26}{3}}$$
.

**Theorem 5.** Probability of intersection of three sets (insert from canvas).

**Definition 19.** We call two events *A*, *B* **independent** if the occurrence of one does not affect the other. Formally,

$$\mathbb{P}(A \mid B) = \mathbb{P}(A)$$
 and  $\mathbb{P}(B \mid A) = \mathbb{P}(B)$ .

We can also check that  $\mathbb{P}(A \cap B) = \mathbb{P}(A) \cdot \mathbb{P}(B)$ .

**Example.** Flip three fair coins. What is likelihood that all three come up heads?

**Proof.** With the sample space approach:  $\Omega = \{H, T\}^3$ . Of 8 total elementary events, 1 has three heads, so the probability is  $\frac{1}{8}$ .

With independence: we know that each event is independent, and all events are  $\frac{1}{2}$ , so the probability is  $\left(\frac{1}{2}\right)^3 = \frac{1}{8}$ .

**Definition 20.** Independence can be expanded to more than just two events (insert from canvas). However, note that events can be pairwise independent, but may not be all together independent.

### Lemma 2.

$$\mathbb{P}(B \mid A) = \mathbb{P}(A \mid B) \frac{\mathbb{P}(B)}{\mathbb{P}(A)}.$$

**Proof.** The RHS is the same as  $\frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)} \cdot \frac{\mathbb{P}(B)}{\mathbb{P}(A)} = \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(A)} = \mathbb{P}(B \mid A).$ 

**Example.** There are n balls that are sequentially chosen without replacement from r red balls and b blue balls. Given that k of the n balls are blue, what is the conditional probability that the first chosen is blue?

Proof.

$$\mathbb{P}(\text{first is blue} \mid k \text{ are blue})$$

$$= \mathbb{P}(k \text{ are blue} \mid \text{first is blue})$$

$$\cdot \frac{\mathbb{P}(\text{first is blue})}{\mathbb{P}(k \text{ are blue})} \dots$$

### Lecture 3

Continuing on with conditional probabiltiy from last time,

**Example.** A total of n balls are sequentially and randomly chosen without replacement from an urn containing r red balls and b blue balls ( $n \le r + b$ ). Given that k of the n balls are blue, what is the conditional probability that the first ball chosen is blue?

Proof. We can use Lemma 2. Then, we have

$$\mathbb{P}(\text{first blue}) = \frac{b}{r+b}$$

$$\mathbb{P}(\text{first } k \text{ are blue}) = \frac{\binom{n}{k} P(b,k) P(r,n-k)}{P(r+b,n)}.$$

 $\mathbb{P}(k-1 \text{ of remaining } n-1 \text{ slots are blue.}) =$ 

$$\frac{\binom{n-1}{k-1}P(b-1,k-1)\cdot P(r,n-k)}{P(r+b-1,n-1)}.$$

 $\mathbb{P}(k-1 \text{ of rest } n-1 \text{ are blue}) \cdot \frac{\mathbb{P}(\text{first blue})}{\mathbb{P}(\text{first } k \text{ are blue})}$ 

will then be our answer.