

Conditional Probability and Independence

B39AX — Fall 2023

Heriot-Watt University

Intuition

Let A and B be two events.

If we know B happened, what does this tell us about A ?

Example 1:

$A =$ “I am left-handed”

$B =$ “One of my parents is left-handed”

Example 2:

$A =$ “Today it rains”

$B =$ “Today 270 babies were born”

Intuition

Let $\mathbb{P}(A | B)$ be the probability that A occurs *given that B occurred*.

In the frequentist interpretation, repeat an experiment T times:

$$\mathbb{P}(A | B) \simeq \frac{\# \text{ times } A \text{ and } B \text{ occurred}}{\# \text{ times } B \text{ occurred}} = \frac{T \cdot f_T(A \cap B)}{T \cdot f_T(B)} \simeq \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)}$$

If B *does not inform* about A , then $\mathbb{P}(A | B) = \mathbb{P}(A)$, or

$$\mathbb{P}(A \cap B) = \mathbb{P}(A) \cdot \mathbb{P}(B)$$

In this case, we say that A and B are *independent*.

Definitions

Conditional probability:

Let A and B be events such that $\mathbb{P}(B) > 0$.

The *conditional probability of A given B* is

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

Independence:

The events A and B are *independent* if $\mathbb{P}(A \cap B) = \mathbb{P}(A) \cdot \mathbb{P}(B)$

Exercise

Roll a fair die.

- $A = \text{"Get number 3"}$

$B = \text{"Get an odd number"}$

Compute $\mathbb{P}(A | B)$ and $\mathbb{P}(B | A)$.

- $A = \text{"Get an even number"}$

$B = \text{"Get number 3"}$

Are A and B independent?

Exercise

A family has two children. What is the probability that both are boys given that at least one is a boy?

Answer: $1/3$

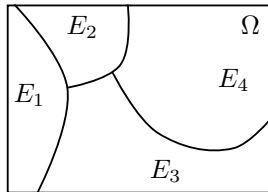
Properties of independence

- A and B independent implies
 - A and B^c independent
 - A^c and B independent
 - A^c and B^c independent
- If $\mathbb{P}(B) > 0$ and A and B are independent, then $\mathbb{P}(A | B) = \mathbb{P}(A)$
- Let $\mathbb{P}(B) > 0$. Then, $\mathbb{Q}(\cdot) := \mathbb{P}(\cdot | B)$ is a probability measure (satisfies the axioms)

Partition

$\{E_i\}_{i=1}^n$ is a **partition** of Ω if

- $E_i \neq \emptyset$ for all i
- $E_i \cap E_j = \emptyset$ for all $i \neq j$
- $\bigcup_{i=1}^n E_i = \Omega$



Example: A and A^c are a partition of Ω .

Three Important Theorems

Multiplication Rule

Motivation

Suppose we have a random password generator that creates passwords using only lowercase characters $\{a, b, c, \dots, z\}$ and never repeats the same character in a password.

If we ask it to generate a 3-length password, what is the probability that it is “abc”?

Multiplication Rule

Theorem (Multiplication Rule)

Let A_1, A_2, \dots, A_n be events and $\mathbb{P}(A_1 \cap \dots \cap A_{n-1}) > 0$. Then,

$$\mathbb{P}(A_1 \cap \dots \cap A_n) = \mathbb{P}(A_1) \cdot \mathbb{P}(A_2 | A_1) \cdots \mathbb{P}(A_n | A_1 \cap \dots \cap A_{n-1})$$

Ans (prev prob.): $\frac{1}{26} \cdot \frac{1}{25} \cdot \frac{1}{24}$

Total Probability Theorem

Motivation

Suppose I have two coins:

- Coin 1 is fair: H and T have the same probability
- Coin 2 is biased: H is twice as likely as T

I select one of the coins with equal probability and flip it.

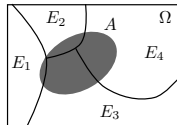
What is the probability of obtaining H ?

Total Probability Theorem

Theorem (Total Probability Theorem)

Let $\{E_i\}_{i=1}^n$ be a partition of Ω . For any event A ,

$$\mathbb{P}(A) = \sum_{i=1}^n \mathbb{P}(A | E_i) \cdot \mathbb{P}(E_i)$$



Ans (prev prob.): $\frac{7}{12}$

Bayes Rule

Motivation

Suppose I have two coins:

- Coin 1 is fair: H and T have the same probability
- Coin 2 is biased: H is twice as likely as T

I select one of the coins with equal probability and flip it.

If I obtained H , what is the probability that I chose coin 2?

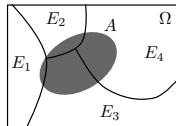
Bayes Rule

Theorem (Bayes Rule)

Let $\{E_i\}_{i=1}^n$ be a partition of Ω . If $\mathbb{P}(A) > 0$, then

$$\mathbb{P}(E_i | A) = \frac{\mathbb{P}(A | E_i) \cdot \mathbb{P}(E_i)}{\sum_{j=1}^n \mathbb{P}(A | E_j) \cdot \mathbb{P}(E_j)}$$

for any $i = 1, \dots, n$.



Allows updating “beliefs” when A happens: $\mathbb{P}(E_i) \rightarrow \mathbb{P}(E_i | A)$

Ans (prev prob.): $\frac{4}{7}$

Example

Consider a disease and a test for the disease.

- The test has accuracy 95%
- The test gives false positives with probability 5%
- The disease affects 0.1% of the population

I was randomly selected to take the test, which turned out positive.

What is the probability of having the disease? *Should I worry?*

80% of people in a survey answered 95% [The Economist, 20 Feb, 1999]

Ans: 1.87%

Proofs of the Theorems

Proof of Multiplication Rule

Apply conditional probability recursively: $\mathbb{P}(A \cap B) = \mathbb{P}(A | B) \cdot \mathbb{P}(B)$

(a formal proof can be done via induction)

$$\mathbb{P}(A_1 \cap A_2) = \mathbb{P}(A_2 | A_1) \cdot \mathbb{P}(A_1)$$

$$\mathbb{P}(A_1 \cap A_2 \cap A_3) = \mathbb{P}((A_1 \cap A_2) \cap A_3)$$

$$= \mathbb{P}(A_3 | A_1 \cap A_2) \cdot \mathbb{P}(A_1 \cap A_2)$$

$$= \mathbb{P}(A_3 | A_1 \cap A_2) \cdot \mathbb{P}(A_2 | A_1) \cdot \mathbb{P}(A_1)$$

$$\vdots$$

$$\mathbb{P}(A_1 \cap \cdots \cap A_n) = \mathbb{P}(A_n | A_1 \cap \cdots \cap A_{n-1}) \cdots \mathbb{P}(A_2 | A_1) \cdot \mathbb{P}(A_1)$$



Proofs of the Theorems

Proof of Total Probability

For any event A , we have $A = A \cap \Omega$. And because the sets E_1, E_2, \dots, E_n are a partition of Ω ,

$$\begin{aligned}\mathbb{P}(A) &= \mathbb{P}(A \cap \Omega) \\ &= \mathbb{P}\left(A \cap (E_1 \cup E_2 \cup \dots \cup E_n)\right) \\ &= \mathbb{P}\left((A \cap E_1) \cup (A \cap E_2) \cup \dots \cup (A \cap E_n)\right) \\ &= \mathbb{P}(A \cap E_1) + \mathbb{P}(A \cap E_2) + \dots + \mathbb{P}(A \cap E_n) \\ &= \mathbb{P}(A | E_1) \cdot \mathbb{P}(E_1) + \mathbb{P}(A | E_2) \cdot \mathbb{P}(E_2) + \dots + \mathbb{P}(A | E_n) \cdot \mathbb{P}(E_n) .\end{aligned}$$

The 4th equality is due to $E_i \cap E_j = \emptyset \implies (A \cap E_i) \cap (A \cap E_j) = \emptyset$.



Proofs of the Theorems

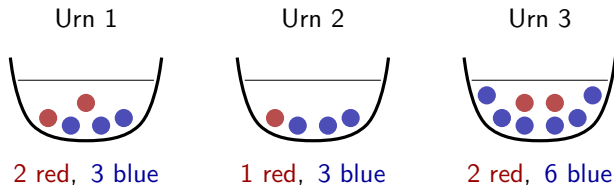
Proof of Bayes Rule

$$\begin{aligned}\mathbb{P}(E_i | A) &= \frac{\mathbb{P}(A \cap E_i)}{\mathbb{P}(A)} \\ &= \frac{\mathbb{P}(A | E_i) \cdot \mathbb{P}(E_i)}{\mathbb{P}(A)} \\ &= \frac{\mathbb{P}(A | E_i) \cdot \mathbb{P}(E_i)}{\sum_{j=1}^n \mathbb{P}(A | E_j) \cdot \mathbb{P}(E_j)},\end{aligned}$$

where the last step used the total probability theorem.



Exercise



Question 1: I select one of the urns uniformly at random, and draw a ball. What is the probability that it is blue? Ans: $\frac{7}{10}$

Question 2: From urn 3, I draw 3 balls consecutively *without* putting them back in.* What is the probability that all of them are blue? $\frac{5}{14}$

*This is called **sampling without replacement**. If we put the balls back in, it is called **sampling with replacement**.

Counting

Computing probabilities often requires counting # of elements in sets

$$\text{Uniform probabilities} \implies \mathbb{P}(A) = \frac{\# \text{ of elements in } A}{\# \text{ of elements in } \Omega}$$

Example: {Alistair, Beatrix, Colin, Donald, Eleanor} are waiting in line.

- How many different configurations can they form?
- Only 3 can be served. How many configurations now? (order matters)
- How many ways of splitting them into 3 served, 2 not served?

Counting

Given a set with n elements,

Permutations: $n! = n \cdot (n - 1) \cdot (n - 2) \cdots 2 \cdot 1$

k -permutations: $A_k^n = \frac{n!}{(n - k)!} = n \cdot (n - 1) \cdots (n - k + 1)$

k -combinations: $\binom{n}{k} = \frac{A_k^n}{k!} = \frac{n!}{(n - k)!k!}$

Convention: $0! = 1$