Probability Theory

Applications for Data Science Module 2: Conditional Probability

Anne Dougherty

February 6, 2021



TABLE OF CONTENTS

Learning Goals

Learning Goals for Module 2

In this module, we'll learn about conditional probability and Bayes formula. At the end of this module, learners should be able to:

- Explain the concept of conditional probability.
- Calculate probabilities using conditioning and Bayes Theorem.
- ► Explain the concepts of independence and mutually exclusive events and provide examples.

Independence

Two events are **independent** if knowing the outcome of one event does not change the probability of the other.

Examples:

- ► Flip a two-sided coin repeatedly. Knowing the outcome of one flip does not change the probability of the next.
- Roll a dice repeatedly.
- What about polling? What if you ask two randomly selected people about their political affiliation? What if the two people are friends?

 If the 2 people are randomly chosen, knowing one person's affiliation shouldn't tell you anything about the other's affiliation, the other's affiliation, they are friends.

 This is not necessarily true if they are friends

Definition

Two events, A and B, are **independent** if P(A|B) = P(A), or equivalently, if P(B|A) = P(B).

Recall: $P(A) = P(A|B) = \frac{P(A \cap B)}{P(B)}$ where independent, we get the multiplication rule for independent.

rule for independent events:

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

Dhis is for 2 wents to be indep. What if we have neverts.

Definition Events A_1, \dots, A_n are mutually independent if for every k (k = 2, 3, ... n) and every subset of indices i_1, i_2, \ldots, i_k :

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$
where definition of independence in two ways:

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2}) \cdots P(A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1})P(A_{i_2} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap A_{i_2} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_1} \cap \cdots \cap A_{i_k})$$

$$P(A_{i_1} \cap \cdots \cap A_{i_k}) = P(A_{i_$$

Use the definition of independence in two ways:

- ▶ We can use the definition to show two events A and B are (or are not) independent. To do this, we calculate P(A), P(B), and $P(A \cap B)$ to check if $P(A \cap B) = P(A)P(B)$.
- ▶ If we know two events are independent, we can find the probability of their intersection.

Example 1

Example: Roll a six-sided dice twice. Recall, $S = \{(i, j) \mid i, j \in \{1, 2, 3, 4, 5, 6\}\}, |S| = 36$ and each of the 36 outcomes of S is equally likely.

Let E be the event that the sum is 7.

Let F be the event that the first roll is a 4.

Let G be the event that the second roll is a 3.

What can you say about the independence of E, F and G?

So, any pair of E, F + G are indep.

Mutual indep? P(ENFNG) = P(3433) = 36 + P(E) F(E) P(G)

So not mutually indep. Think about this. If you

know a of the events has occurred, then the 3 minutes has occurred.

Example 2

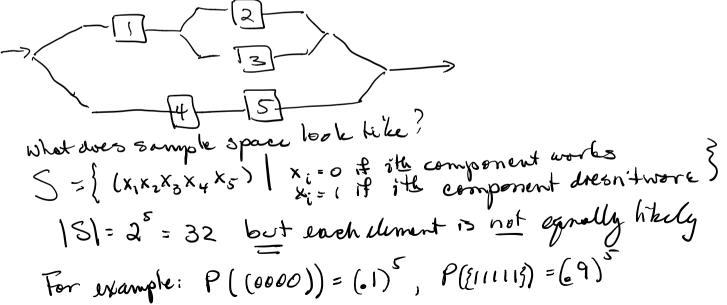
In a school of 1200 students, 250 are juniors, 150 students are taking a statistics course, and 40 students are juniors and also taking statistics. One student is selected at random from the entire school. Let J be the event the selected student is a junior. Let S be the event that the selected student is taking statistics.

If the randomly chosen student is a junior, then what is the probability that they are also taking stats? Are J and S independent? $P(5) = \frac{250}{1200}$, $P(5) = \frac{150}{1200}$, $P(5) = \frac{40}{1200}$.

(Also, note: $P(5) = \frac{40}{1200} \neq P(5)$, $P(5) = \frac{250}{1200}$. $P(5) = \frac{150}{1200}$. $P(5) = \frac{150}{1200}$. $P(5) = \frac{150}{1200}$. $P(5) = \frac{150}{1200}$.

Example 3

Suppose you have a system of components as in the diagram. Let A_i be the event that the i^{th} component works and assume $P(A_i) = .9$ for i = 1, 2, 3, 4, 5. Assume the components work independently of each other. For the system to work, you need a path of working components from the start to the finish. Find the probability that the system works.



Example 3 - continued Paper P(ANB) = P(A)+P(B) - P(A NB)

P(ANB)C) = P(A)+P(B)+P(C) - P(ANB)

P(system works) = $P(A_1 \cap A_2) \cup (A_1 \cap A_3) \cup (A_2 \cap A_5) \cup (A_3 \cap A_4) \cup (A_4 \cap A_5) \cup (A_5 \cap A_5) \cup$

 $-P(A_1 \cap A_2 \cap A_3) - P(A_1 \cap A_2 \cap A_4 \cap A_5) - P(A_1 \cap A_2 \cap A_4 \cap A_5) - P(A_1 \cap A_2 \cap A_4 \cap A_5)$ $= 3(.9)^2 - (.9)^3 - 2(.9)^4 + (.9)^5 = .97929$ $= 3(.9)^2 - (.9)^3 - 2(.9)^4 + (.9)^5 = .97929$ overall prob.

So P(A) (Az) = (.9) = .81

probagatem

with

components 1+2

Extend: prob. of various

Extend: prob. of various

companints might not be

the same, etc. But basic
idea is the same.

One final question: Suppose you know two events A and B are mutually exclusive, that is, $A \cap B = \emptyset$. Are A and B independent?

Jon might think yes since if A&B we mustually exclusive then knowing the probof one down't influence the other.

But:

P(A|B) = P(A \ \ B) = O

P(B)

If we know B has occurred, then A cannot occurred, then A cannot