

Lecture 6: Independence and Probability Calculation

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Recap: conditional probability

Definition: Conditional probability of A given B is defined as

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

Multiplication rule:

- $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B|A)$
- $\mathbb{P}(A \cap B \cap C) = \mathbb{P}(A)\mathbb{P}(B|A)\mathbb{P}(C|A \cap B)$

Rule of total probability: If $\Omega = A_1 \cup A_2 \cup \dots \cup A_k$ and A_1, \dots, A_k are mutually exclusive (such A_1, \dots, A_k are called a **partition** of the sample space), then for any event B ,

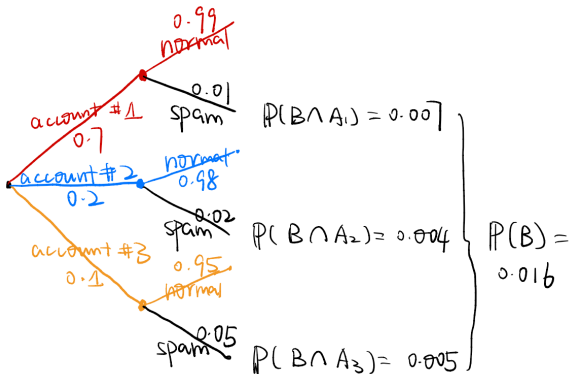
$$\mathbb{P}(B) = \mathbb{P}(A_1)\mathbb{P}(B|A_1) + \mathbb{P}(A_2)\mathbb{P}(B|A_2) + \dots + \mathbb{P}(A_k)\mathbb{P}(B|A_k)$$

Recap: Bayes' Theorem

Bayes' Theorem: If A_1, \dots, A_k are a partition of sample space Ω , then for any event B ,

$$\mathbb{P}(A_j|B) = \frac{\mathbb{P}(A_j \cap B)}{\mathbb{P}(B)} = \frac{\mathbb{P}(B|A_j)\mathbb{P}(A_j)}{\sum_{i=1}^k \mathbb{P}(B|A_i)\mathbb{P}(A_i)}, \quad j = 1, \dots, k.$$

Tree diagram:

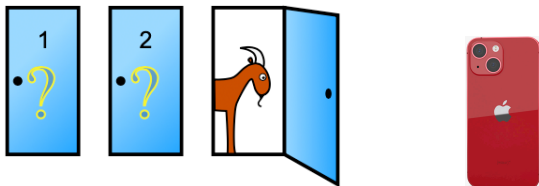


Last lecture: Monty Hall problem

You're given the choice of three doors: Behind one door is an iPhone 13; behind the others, goats.

You **randomly** pick a door, say No.1, and the host, **who knows what's behind the doors**, opens another door, say No.3, which has a goat. He then says to you, "Do you want to pick door No.2?"

Question: Would you switch your choice?



Underlying assumptions:

- If the iPhone is behind the door we pick, then the host **randomly** picks one door from the remaining two to open.
- Otherwise, the host opens the door of goat.

Last lecture: Monty Hall problem

Solution (by Bayes Theorem):

$$\circ \mathbb{P}(\text{iPhone is behind door } i) = 1/3, i = 1, 2, 3$$

Then

$$\mathbb{P}(\text{win iPhone by staying with \#1})$$

$$= \mathbb{P}(\text{iPhone behind \#1} | \text{host opens \#3 and see a goat})$$

$$= \frac{\mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#1}) \mathbb{P}(\text{iPhone behind \#1})}{\sum_{i=1}^3 \mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#}i) \mathbb{P}(\text{iPhone behind \#}i)}$$

$$= \frac{\frac{1}{2} \times \frac{1}{3}}{\frac{1}{2} \times \frac{1}{3} + 1 \times \frac{1}{3} + 0 \times \frac{1}{3}}$$

$$= \frac{1}{3}.$$

Last lecture: Monty Hall problem

Solution (by Bayes Theorem):

$$\circ \mathbb{P}(\text{iPhone is behind door } i) = 1/3, i = 1, 2, 3$$

Then

$$\mathbb{P}(\text{win iPhone after switching to \#2})$$

$$= \mathbb{P}(\text{iPhone behind \#2} | \text{host opens \#3 and see a goat})$$

$$= \frac{\mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#2}) \mathbb{P}(\text{iPhone behind \#2})}{\sum_{i=1}^3 \mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#}i) \mathbb{P}(\text{iPhone behind \#}i)}$$

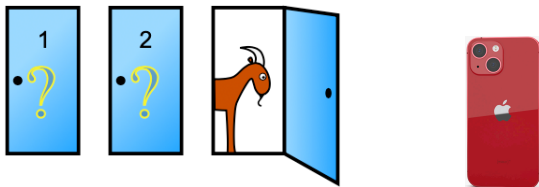
$$= \frac{1 \times \frac{1}{3}}{\frac{1}{2} \times \frac{1}{3} + 1 \times \frac{1}{3} + 0 \times \frac{1}{3}}$$

$$= \frac{2}{3} > \frac{1}{3} \implies \text{We should switch!}$$

Last lecture: Monty Hall problem (modified)

You **randomly** pick a door, say No.1, and the host, **who doesn't know what's behind the doors**, **randomly** opens another door among the remaining two, say No.3, which has a goat. He then says to you, "Do you want to pick door No.2?"

Question: Would you switch your choice?



Underlying assumptions:

- No matter which one you originally pick, the host always **randomly** picks one door from the remaining two to open.

Last lecture: Monty Hall problem (modified)

Solution (by Bayes Theorem):

$$\circ \mathbb{P}(\text{iPhone is behind door } i) = 1/3, i = 1, 2, 3$$

Then

$$\mathbb{P}(\text{win iPhone by staying with \#1})$$

$$= \mathbb{P}(\text{iPhone behind \#1} | \text{host opens \#3 and see a goat})$$

$$= \frac{\mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#1}) \mathbb{P}(\text{iPhone behind \#1})}{\sum_{i=1}^3 \mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#}i) \mathbb{P}(\text{iPhone behind \#}i)}$$

$$= \frac{\frac{1}{2} \times \frac{1}{3}}{\frac{1}{2} \times \frac{1}{3} + \frac{1}{2} \times \frac{1}{3} + 0 \times \frac{1}{3}}$$

$$= \frac{1}{2}.$$

Last lecture: Monty Hall problem (modified)

Solution (by Bayes Theorem):

$$\circ \mathbb{P}(\text{iPhone is behind door } i) = 1/3, \quad i = 1, 2, 3$$

Then

$$\mathbb{P}(\text{win iPhone after switching to \#2})$$

$$= \mathbb{P}(\text{iPhone behind \#2 and see a goat} | \text{host opens \#3})$$

$$= \frac{\mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#2}) \mathbb{P}(\text{iPhone behind \#2})}{\sum_{i=1}^3 \mathbb{P}(\text{host opens \#3 and see a goat} | \text{iPhone behind \#i}) \mathbb{P}(\text{iPhone behind \#i})}$$

$$= \frac{\frac{1}{2} \times \frac{1}{3}}{\frac{1}{2} \times \frac{1}{3} + \frac{1}{2} \times \frac{1}{3} + 0 \times \frac{1}{3}}$$

$$= \frac{1}{2} \quad \implies \text{It doesn't matter!!}$$

Reflection on Monty Hall problem

- Why is the answer so different?
- A more intuitive explanation of the **original** Monty Hall problem:

Behind door 1	Behind door 2	Behind door 3	Result if staying at door #1	Result if switching to the door offered
Goat	Goat	iPhone	Wins goat	Wins iPhone
Goat	iPhone	Goat	Wins goat	Wins iPhone
iPhone	Goat	Goat	Wins iPhone	Wins goat

The host always helps us kick out a wrong choice in the remaining two!

- But in the modified version, the host selects the door **randomly**, which doesn't benefit switching.

Today's goal

- Understand the definition of independence of events
- Know how to decide whether events are independent or not
- Comprehensively use all properties of probability we mentioned before to calculate relative problems

Independence

Intuition

In previous examples, often $\mathbb{P}(A|B) \neq \mathbb{P}(A)$, which means the information "B has occurred" resulted in a change in the likelihood of A occurring.

But sometimes, $\mathbb{P}(A|B) = \mathbb{P}(A)$, which means the chance that A will occur or has occurred is not affected by knowledge that B has occurred.

Example: A box contains 4 red tickets and 6 white tickets. Pick a ticket at random from the box.



- What is the chance of getting a winning ticket?(5/10)
- If you see that it is a white one, what is the chance of getting a winning ticket?(3/6)

Independence between two events

Definition: Two events A and B are **independent** if $\mathbb{P}(A|B) = \mathbb{P}(A)$, denoted as $A \perp B$. They are **dependent** otherwise, denoted as $A \not\perp B$.

Question: What can we say about \emptyset and Ω ?

Equivalent definitions for $A \perp B$:

- $\mathbb{P}(B|A) = \mathbb{P}(B)$
- $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$ (very useful!)

Properties: A and B are independent iff (if and only if) any of the following holds:

- A and B^c are independent
- A^c and B are independent
- A^c and B^c are independent

You will prove these in HW2!

Example

Consider a gas station with six pumps numbered 1, 2, ..., 6. Suppose that

$$\mathbb{P}(\{1\}) = \mathbb{P}(\{6\}) = 0.1, \quad \mathbb{P}(\{2\}) = \mathbb{P}(\{5\}) = 0.15, \quad \mathbb{P}(\{3\}) = \mathbb{P}(\{4\}) = 0.25.$$

Define events $A = \{2, 4, 6\}$, $B = \{1, 2, 3\}$, $C = \{2, 3, 4, 5\}$.

Therefore:

- $\mathbb{P}(A) = 0.5$, $\mathbb{P}(B) = 0.5$, $\mathbb{P}(C) = 0.8$.
- $\mathbb{P}(A \cap B) = \mathbb{P}(\{2\}) = 0.15$
- $\mathbb{P}(B \cap C) = \mathbb{P}(\{2, 3\}) = 0.4$
- $\mathbb{P}(A \cap C) = \mathbb{P}(\{2, 4\}) = 0.4$

Hence:

- $\mathbb{P}(A \cap B) \neq \mathbb{P}(A)\mathbb{P}(B) \implies A \not\perp B$
- $\mathbb{P}(B \cap C) = \mathbb{P}(B)\mathbb{P}(C) \implies B \perp C$
- $\mathbb{P}(A \cap C) = \mathbb{P}(A)\mathbb{P}(C) \implies A \perp C$

Independence between multiple events

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

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

- When $n = 2$: it reduces to $\mathbb{P}(A_1 \cap A_2) = \mathbb{P}(A_1)\mathbb{P}(A_2)$
- When $n = 3$: it reduces to (all the following conditions have to hold at the same time):
 - ▷ $\mathbb{P}(A_1 \cap A_2) = \mathbb{P}(A_1)\mathbb{P}(A_2)$
 - ▷ $\mathbb{P}(A_2 \cap A_3) = \mathbb{P}(A_2)\mathbb{P}(A_3)$
 - ▷ $\mathbb{P}(A_1 \cap A_3) = \mathbb{P}(A_1)\mathbb{P}(A_3)$
 - ▷ $\mathbb{P}(A_1 \cap A_2 \cap A_3) = \mathbb{P}(A_1)\mathbb{P}(A_2)\mathbb{P}(A_3)$

Therefore, independence is often an idealistic assumption in practice!

Independence between multiple events

Theorem: If A_1, \dots, A_n are mutually independent, then for any index sets $\{i_1, \dots, i_k\}$ and $\{j_1, \dots, j_r\}$ where $\{i_1, \dots, i_k\} \cap \{j_1, \dots, j_r\} = \emptyset$,

intersections/unions between $A_{i_1}, \dots, A_{i_k} \perp\!\!\!\perp$
intersections/unions between A_{j_1}, \dots, A_{j_r} .

The events A_{i_1}, \dots, A_{i_k} and A_{j_1}, \dots, A_{j_r} can be replaced by their complements.

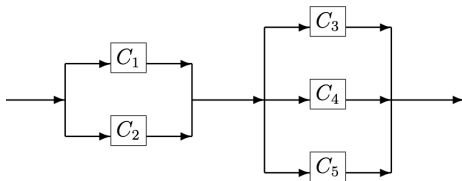
For example:

- $(A_1 \cup A_2) \perp\!\!\!\perp (A_3 \cap A_4)$
- $(A_1 \cap A_2^c \cup A_5) \perp\!\!\!\perp (A_3 \cup A_4)$
- $A_2 \perp\!\!\!\perp (A_1^c \cap A_4 \cap A_3)$
- ...

Example: parallel system

A system consists of 5 **independent** components C_1, \dots, C_5 .

$\mathbb{P}(C_i \text{ works properly}) = 0.9, i = 1, \dots, 5$. The chance that the system will work properly =?



$$\{\text{The whole system works}\} = (C_1 \cup C_2) \cap (C_3 \cup C_4 \cup C_5)$$

$$\begin{aligned}\mathbb{P}(\text{The whole system works}) &= \mathbb{P}((C_1 \cup C_2) \cap (C_3 \cup C_4 \cup C_5)) \\ &= \mathbb{P}(C_1 \cup C_2) \times \mathbb{P}(C_3 \cup C_4 \cup C_5).\end{aligned}$$

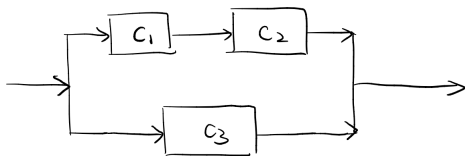
- $\mathbb{P}(C_1 \cup C_2) = 1 - \mathbb{P}(C_1^c \cap C_2^c) = 1 - \mathbb{P}(C_1^c)\mathbb{P}(C_2^c) = 0.99$;
- $\mathbb{P}(C_3 \cup C_4 \cup C_5) = 1 - \mathbb{P}(C_3^c)\mathbb{P}(C_4^c)\mathbb{P}(C_5^c) = 1 - 0.1^3 = 0.999$

$$\implies \mathbb{P}(\text{The whole system works}) = 0.99 \times 0.999 = 0.989$$

Example: parallel system 2

A system consists of 3 **independent** components C_1, C_2, C_3 .

$\mathbb{P}(C_i \text{ works properly}) = 0.7, i = 1, \dots, 3$. The chance that the system will work properly =?



$$\{\text{The whole system works}\} = (C_1 \cap C_2) \cup C_3$$

$$\begin{aligned}\mathbb{P}(\text{The whole system works}) &= \mathbb{P}((C_1 \cap C_2) \cup C_3) \\ &= \mathbb{P}(C_1 \cap C_2) + \mathbb{P}(C_3) - \mathbb{P}(C_1 \cap C_2 \cap C_3) \\ &= \mathbb{P}(C_1)\mathbb{P}(C_2) + \mathbb{P}(C_3) - \mathbb{P}(C_1)\mathbb{P}(C_2)\mathbb{P}(C_3) \\ &= 0.7^2 + 0.7 - 0.7^3 \\ &= 0.847\end{aligned}$$

Example: coffee purchase

The accompanying table gives information on the type of coffee selected by someone purchasing a single cup.

	Small	Medium	Large
Regular	14%	20%	26%
Decaf	20%	10%	10%

Consider randomly selecting such a coffee purchaser.

- Is "buying decaf coffee" independent with "buying medium-size coffee"?
- Are "buying regular coffee", "buying small-size coffee" and "buying medium-size coffee" mutually independent?

$$\mathbb{P}(\text{decaf} \cap \text{medium}) = 10\% \neq 40\% \times 30\% = \mathbb{P}(\text{decaf}) \times \mathbb{P}(\text{medium})$$

\Rightarrow decaf $\not\perp$ medium

Can we claim that {regular}, {small} and {medium} are NOT mutually independent?

Review: Probability Calculation

Review: Probability Calculation

$\mathbb{P}(\cdot) : A \in$ a set of all events \rightarrow a number $\mathbb{P}(A)$ (a mapping/function which maps an event to a number)

Three axioms:

- $0 \leq \mathbb{P}(A) \leq 1$ for any event $A \subseteq \Omega$
- $\mathbb{P}(\Omega) = 1, \mathbb{P}(\emptyset) = 0$
- If $A_1, A_2, A_3, \dots, A_n$ is a collection of disjoint (mutually exclusive) events, then

$$\mathbb{P}(A_1 \cup A_2 \cup \dots \cup A_n) = \sum_{i=1}^n \mathbb{P}(A_i).$$

Theorem: In an experiment consisting of N outcomes **with equal probability**, for any event A ,

$$\mathbb{P}(A) = \frac{N(A)}{N},$$

where we use counting techniques to calculate $N(A)$ and N .

Review: Probability Calculation

General idea to calculate probability:

(1) Translate the conditions and the event of interest by probability language

(2) Calculate.

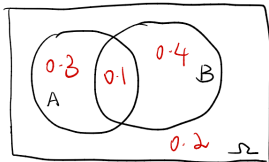
- ▷ For equally likely outcomes, consider $\mathbb{P}(A) = \frac{N(A)}{N}$ combined with counting techniques
- ▷ If conditioning probability is easier to calculate or it is given, consider conditional multiplication rule, rule of total probability, and Bayes' Theorem
- ▷ See whether there are independence or not, which may help us to do the calculation

Example 1

Given A and B , such that $\mathbb{P}(A) = \mathbb{P}(A^c \cap B) = 0.4$ and $\mathbb{P}(A \cap B) = 0.1$.

- Find $\mathbb{P}(A \cup B)$
 - Are A and B independent?
-

- Venn diagram:



Or:

- ▷ $\mathbb{P}(B) = \mathbb{P}(A^c \cap B) + \mathbb{P}(A \cap B) = 0.4 + 0.1 = 0.5$
- ▷ Then $\mathbb{P}(A \cup B) = \mathbb{P}(A) + \mathbb{P}(B) - \mathbb{P}(A \cap B)$
 $= 0.4 + 0.5 - 0.1 = 0.8$
- No, because $\mathbb{P}(A \cap B) = 0.1 \neq 0.4 \times 0.5 = \mathbb{P}(A) \times \mathbb{P}(B)$.

Example 2

5 faculty members (3 pro-A and 2 pro-B) vote for the department chair from candidate A and B **one by one in a random order**.

The probability that "A remains ahead of B throughout the vote count" = ? (e.g., "AABAB" is OK, but not "ABBAA")

Some observations:

- Permutation problem
- The last one has to vote for *B*. (Why?)
- The first one must vote for *A*. (Why?)
- So it has to be "ABAAB", "AABAB" or "AAABB"

$$N(\Omega) = P_{5,5} = 5! = 120$$

$$N(\text{A remains ahead of B}) = 3 \times P_{3,3} \times P_{2,2} = 36$$

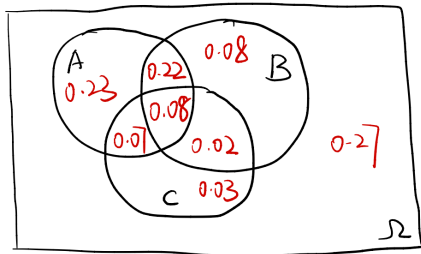
$$\mathbb{P}(\text{A remains ahead of B}) = \frac{N(\text{A remains ahead of B})}{N(\Omega)} = 0.3$$

Example 3

Randomly select a student at Columbia:

- $A = \{\text{the selected student has a American Express card}\}$
 $B = \{\text{the selected student has a Barclays card}\}$
 $C = \{\text{the selected student has a Chase card}\}$
- $\mathbb{P}(A) = 0.6, \mathbb{P}(B) = 0.4, \mathbb{P}(C) = 0.2.$
 $\mathbb{P}(A \cap B) = 0.3, \mathbb{P}(A \cap C) = 0.15, \mathbb{P}(B \cap C) = 0.1$
 $\mathbb{P}(A \cap B \cap C) = 0.08$

- (1) The probability that the selected student has ≥ 1 of 3 types of cards=?
- (2) $\mathbb{P}(B|A) = ?, \mathbb{P}(A|B) = ?$



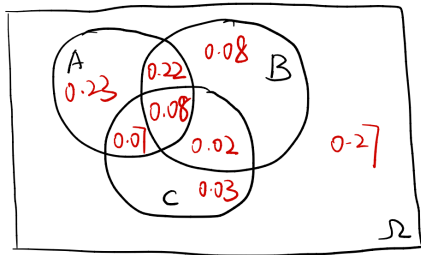
- $\mathbb{P}(A \cup B \cup C) = 0.73$
- $\mathbb{P}(B|A) = \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(A)} = 0.5,$
 $\mathbb{P}(A|B) = \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)} = 0.75$

Example 3

Randomly select a student at Columbia:

- $A = \{\text{the selected student has a American Express card}\}$
 $B = \{\text{the selected student has a Barclays card}\}$
 $C = \{\text{the selected student has a Chase card}\}$
- $\mathbb{P}(A) = 0.6, \mathbb{P}(B) = 0.4, \mathbb{P}(C) = 0.2.$
 $\mathbb{P}(A \cap B) = 0.3, \mathbb{P}(A \cap C) = 0.15, \mathbb{P}(B \cap C) = 0.1$
 $\mathbb{P}(A \cap B \cap C) = 0.08$

(3) Given that the student has an American Express card, what is the probability that she/he has ≥ 1 of the other two types of cards?



$$\begin{aligned} \circ \mathbb{P}(B \cup C | A) &= \frac{\mathbb{P}((B \cup C) \cap A)}{\mathbb{P}(A)} = \\ &= \frac{0.22 + 0.08 + 0.07}{0.6} = 0.617 \end{aligned}$$

Example 4

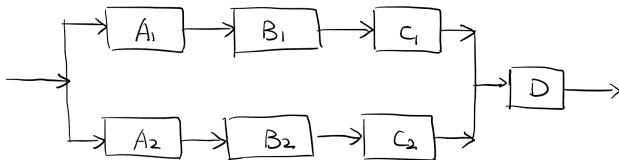
A system consists of 7 **independent** components A_1, \dots, D .

$$\mathbb{P}(A_1 \text{ works}) = \mathbb{P}(A_2 \text{ works}) = 0.9$$

$$\mathbb{P}(B_1 \text{ works}) = \mathbb{P}(B_2 \text{ works}) = 0.8$$

$$\mathbb{P}(C_1 \text{ works}) = \mathbb{P}(C_2 \text{ works}) = 0.7, \mathbb{P}(D \text{ works}) = 0.95$$

The chance that the system will work properly = ?



$$\begin{aligned}\mathbb{P}(\{\text{the system works}\}) &= \mathbb{P}(((A_1 \cap B_1 \cap C_1) \cup (A_2 \cap B_2 \cap C_2)) \cap D) \\ &= \mathbb{P}((A_1 \cap B_1 \cap C_1) \cup (A_2 \cap B_2 \cap C_2))\mathbb{P}(D) \\ &= [2\mathbb{P}(A_1 \cap B_1 \cap C_1) - \mathbb{P}((A_1 \cap B_1 \cap C_1) \cap (A_2 \cap B_2 \cap C_2))]\mathbb{P}(D) \\ &= [2\mathbb{P}(A_1)\mathbb{P}(B_1)\mathbb{P}(C_1) - \mathbb{P}(A_1 \cap B_1 \cap C_1)\mathbb{P}(A_2 \cap B_2 \cap C_2)]\mathbb{P}(D) \\ &= [2 \times 0.9 \times 0.8 \times 0.7 - (0.9 \times 0.8 \times 0.7)^2] \times 0.95 \\ &= 0.717\end{aligned}$$

Conditional independence

Independence: Two events A and B are **independent** if $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$, denoted as $A \perp B$. They are **dependent** otherwise, denoted as $A \not\perp B$.

Conditional independence: Two events A and B are **independent** conditioning on event C if $\mathbb{P}(A \cap B|C) = \mathbb{P}(A|C)\mathbb{P}(B|C)$, denoted as $(A \perp B)|C$. They are **dependent** otherwise, denoted as $(A \not\perp B)|C$.

Can be extended to the case of multiple (≥ 2) events too.

Example 5

- 5% of the population has a certain disease
- A diagnostic test:
 - ▷ correctly detects the presence of the disease 98% of the time
 - ▷ correctly detects the absence of the disease 99% of the time
- Randomly select an individual \rightarrow run tests twice \rightarrow Both positive
- The two test results are independent **given the presence/absence of the disease on the selected individual**

Question: The probability that this individual has the disease = ?

- $P_i = \{i\text{-th test positive}\}$, $i = 1, 2$
- $D = \{\text{this individual has the disease}\}$
- Goal: $\mathbb{P}(D|P_1 \cap P_2) = ?$
- What we know:
 - ▷ $\mathbb{P}(D) = 0.05$. $\mathbb{P}(P_i|D) = 0.98$, $\mathbb{P}(P_i^c|D^c) = 0.99$, $i = 1, 2$
 - ▷ $P_1 \perp\!\!\!\perp P_2$ given D , $P_1 \perp\!\!\!\perp P_2$ given D^c
- **Bayes' Theorem!**

Example 5

- $P_i = \{i\text{-th test positive}\}$, $i = 1, 2$
- $D = \{\text{this individual has the disease}\}$
- Goal: $\mathbb{P}(D|P_1 \cap P_2) = ?$
- What we know:
 - ▷ $\mathbb{P}(D) = 0.05$. $\mathbb{P}(P_i|D) = 0.98$, $\mathbb{P}(P_i^c|D^c) = 0.99$, $i = 1, 2$
 - ▷ $P_1 \perp\!\!\!\perp P_2$ given D , $P_1 \perp\!\!\!\perp P_2$ given D^c

By Bayes' Theorem,

$$\begin{aligned}\mathbb{P}(D|P_1 \cap P_2) &= \frac{\mathbb{P}(P_1 \cap P_2|D)\mathbb{P}(D)}{\mathbb{P}(P_1 \cap P_2|D)\mathbb{P}(D) + \mathbb{P}(P_1 \cap P_2|D^c)\mathbb{P}(D^c)} \\ &= \frac{\mathbb{P}(P_1|D)\mathbb{P}(P_2|D)\mathbb{P}(D)}{\mathbb{P}(P_1|D)\mathbb{P}(P_2|D)\mathbb{P}(D) + \mathbb{P}(P_1|D^c)\mathbb{P}(P_2|D^c)\mathbb{P}(D^c)} \\ &= \frac{0.98 \times 0.98 \times 0.05}{0.98 \times 0.98 \times 0.05 + 0.01 \times 0.01 \times 0.95} = 0.998.\end{aligned}$$

Reading list (optional)

- "Probability and Statistics for Engineering and the Sciences" (9th edition):
 - ▷ Chapter 2.5
- "OpenIntro statistics" (4th edition, free online, download [[here](#)]):
 - ▷ Chapter 3.1.7 and 3.2.6

Many thanks to

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- Joyce Robbins
- Chengliang Tang
- Owen Ward
- Wenda Zhou
- And all my teachers in the past 25 years