


# IMD0033 - Probabilidade

## Aula 23 - Probabilidade Condicional

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Junho, 2019



# Agenda

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- Solving Complex Probability Problems
- Conditional Probability
- Bayes' theorem

# Atualizar o repositório

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```
git clone https://github.com/ivanovitchm/imd0033_2019_1.git
```

Ou ....

```
git pull
```

# Complex Probability Problems

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1. What is the probability that it takes three flips or more for a coin to land heads up?
2. What is the probability of a coin landing heads up 18 times in a row?
3. What is the probability of getting at least one 6 in four throws of a single six-sided die?
4. What is the probability of getting at least one double-six in 24 throws of two six-sided dice?
5. What is the probability of getting four aces in a row when drawing cards from a standard 52-card deck?

# Previously on last lesson

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## Addition Rule

$$P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

An advertisement company runs a quick test and shows two ads on the same web page (ad "A" and ad "B") to 100 users. At the end of the trial, they found:

- 12 users clicked on ad "A"
- 17 users clicked on ad "B"
- 3 users clicked on both ad "A" and ad "B"

# Opposite Events

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$$B = \{2\}$$

$$\text{non-}B = \{1, 3, 4, 5, 6\}$$



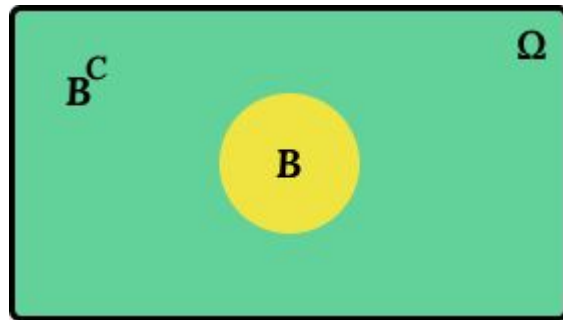
$$P(B \cup \text{non-}B) = P(B) + P(\text{non-}B) = \frac{1}{6} + \frac{5}{6} = 1$$

# Opposite Events (Set Notation)

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$$B = \{2\}$$

$$B^C = \{1, 3, 4, 5, 6\}$$



$$P(B \cup B^C) = P(B) + P(B^C) = 1$$

$$P(B) = 1 - P(B^C)$$

# Example Walk-Through

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- What is the probability that it takes three flips or more for a coin to land heads up?

**Event A:** corresponds to the following outcomes, where each number represents the number of flips it takes until we first get heads up

$$A = \{3, 4, 5, 6, \dots, 100, 101, \dots\}$$











# Example Walk-Through

	First flip	Second flip	Combined outcomes
$A = \{3, 4, 5, 6, \dots, 100, 101, \dots\}$	 Heads	 Heads	→ HH
$non-A = \{1, 2\}$	 Heads	 Tails	→ HT
$P(A) = 1 - P(non-A)$	 Tails	 Heads	→ TH
$P(A) = 1 - 0.75 = 0.25$	 Tails	 Tails	→ TT

# The Multiplication Rule

What is the probability of a coin landing heads up two times in a row?

First flip	Second flip	Combined outcomes
 Heads	 Heads	→ HH
 Heads	 Tails	→ HT
 Tails	 Heads	→ TH
 Tails	 Tails	→ TT

$$P(HH) = \frac{1}{4} = 0.25$$

$$P(H_1 \cap H_2) = P(H_1) \times P(H_2)$$

$$P(H_1 \cap H_2) = 0.5 \times 0.5 = 0.25$$

# Independent Events

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Consider now the following two events, which are associated with rolling a fair six-sided die:

- A: we get a number less than 4; event A corresponds to the outcomes  $\{1, 2, 3\}$
- B: we get an even number; event B corresponds to the outcomes  $\{2, 4, 6\}$

$$P(A) = \frac{3}{6} \text{ and } P(B) = \frac{3}{6}$$

If event A happens:  $P(B) = \frac{1}{3}$

# Independent Events

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$$P(E_1 \cap E_2 \cap E_3 \dots E_n) = P(E_1) \times P(E_2) \times P(E_3) \times \dots P(E_n)$$

For instance, this is how we find the probability that a coin lands heads up four times in a row:

$$P(H_1 \cap H_2 \cap H_3 \cap H_4) = H_1 \times H_2 \times H_3 \times H_4$$

$$P(H_1 \cap H_2 \cap H_3 \cap H_4) = 0.5 \times 0.5 \times 0.5 \times 0.5 = 0.5^4 = 0.0625$$

# Independent Events

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- What is the probability of a coin landing heads up 18 times in a row?

```
0.5**18
```

```
3.814697265625e-06
```

# Combining formulas

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$$P(E) = 1 - P(E^C)$$

$$P(E_1 \cap E_2 \cap E_3 \dots E_n) = P(E_1) \times P(E_2) \times P(E_3) \times \dots P(E_n)$$

**What is the probability of getting at least one 6 in four throws of a single six-sided die?**

# Combining formulas













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- Event  $A$  is getting at least one 6 in four throws
- Event  $A^C$  is not getting any 6 in four throws
- So event  $A^C$  is equivalent to getting any of the outcomes  $\{1, 2, 3, 4, 5\}$  four times in a row.

$$P(A^C) = \left(\frac{5}{6}\right)^4 = 0.4823 \quad P(A) = 1 - \left(\frac{5}{6}\right)^4 = 0.5177$$

# Combining formulas

What is the probability of getting at least one double-six in 24 throws of two six-sided dice?

		First die					
Second die	&						
		1,1	2,1	3,1	4,1	5,1	6,1
		1,2	2,2	3,2	4,2	5,2	6,2
		1,3	2,3	3,3	4,3	5,3	6,3
		1,4	2,4	3,4	4,4	5,4	6,4
		1,5	2,5	3,5	4,5	5,5	6,5
		1,6	2,6	3,6	4,6	5,6	6,6

$$P(A) = 1 - \left(\frac{35}{36}\right)^{24}$$



# Sampling With(out) Replacement

What is the probability of getting four aces in a row when drawing cards from a standard 52-card deck?

# Sampling With(out) Replacement

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$$P(ace) = \frac{4}{52} \qquad P(AAAA) = \left(\frac{4}{52}\right)^4 = 0.000035 \quad ?$$

$$P(AAAA) = \frac{4}{52} \times \frac{3}{51} \times \frac{2}{50} \times \frac{1}{49} = 0.00000369$$

# Sampling Without Replacement

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1. Getting two kings in a row
2. Getting a seven of hearts, followed by a queen of diamonds
3. Getting a jack, followed by a queen of diamonds, followed by a king, followed by another jack

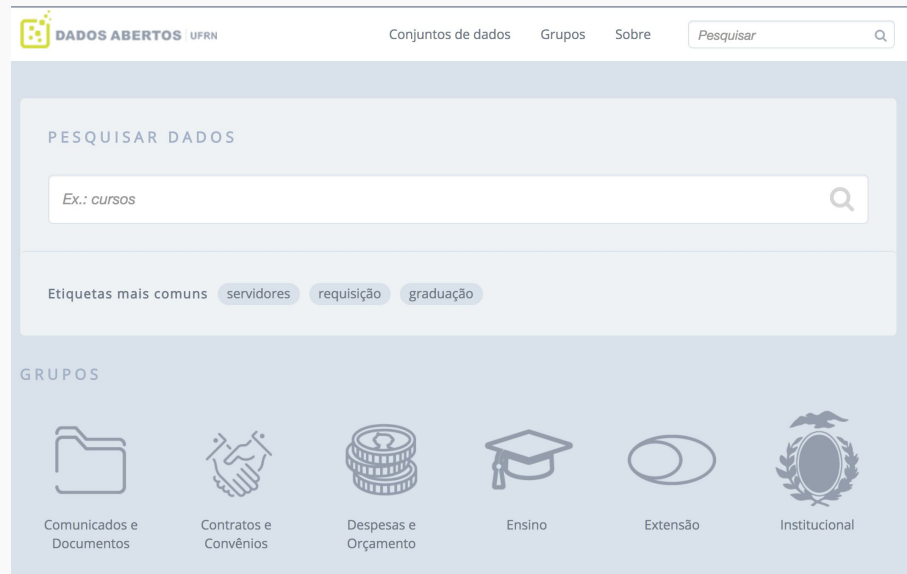
# Conditional Probability

When new evidence comes our way, it helps us update our beliefs and create a new belief

# Conditional Probability

How likely is a student to pass IMD0033?

Since a student took a grade higher than 5 (five) in the first unit of IMD0033, how likely is it to pass in the curricular component?



# Conditional Probability

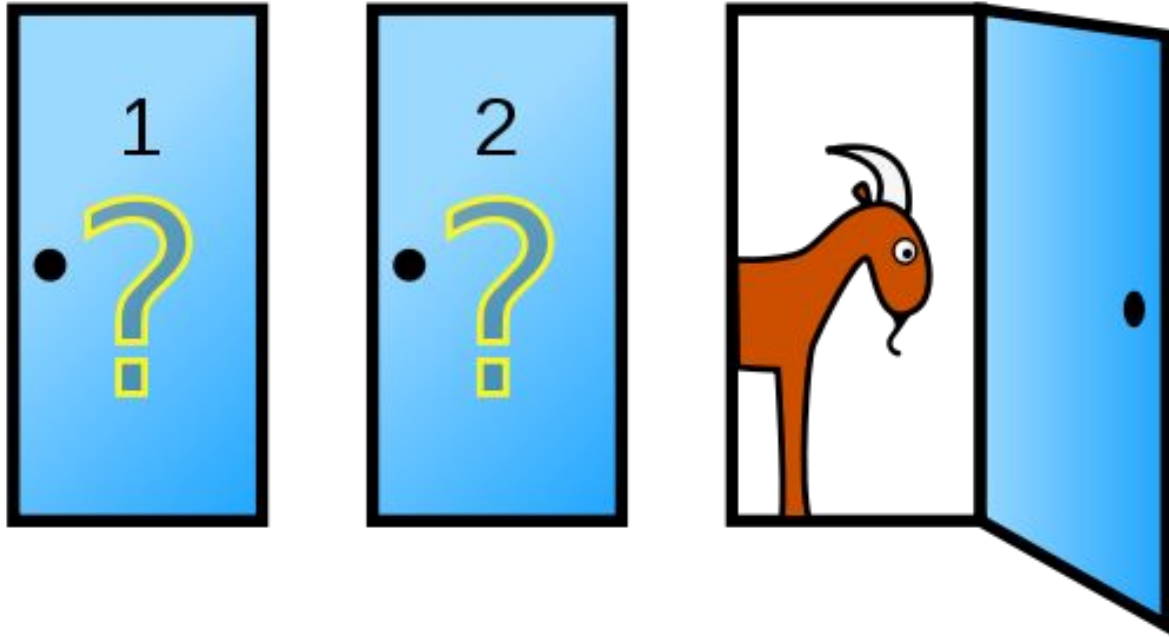
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How likely is it to rain tomorrow?

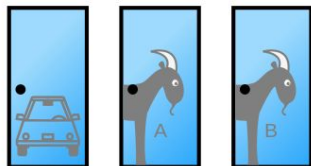
How likely is it to rain tomorrow  
since the night is cloudy?



# Monty Hall Problem



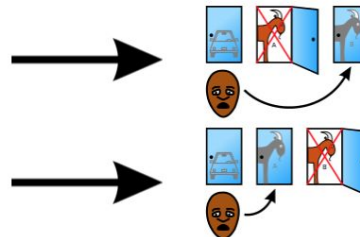
1.



Jogador escolhe carro

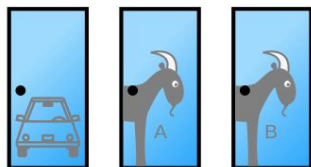
(probabilidade 1/3)

*Apresentador revela  
um dos bodes*



Trocar perde.

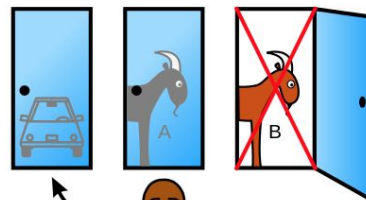
2.



Jogador escolhe Bode A

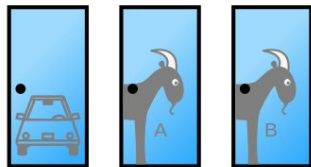
(probabilidade 1/3)

*Apresentador tem que  
revelar Bode B*



Trocar ganha.

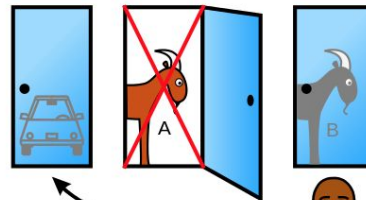
3.



Jogador escolhe Bode B

(probabilidade 1/3)

*Apresentador tem que  
revelar Bode A*



Trocar ganha.

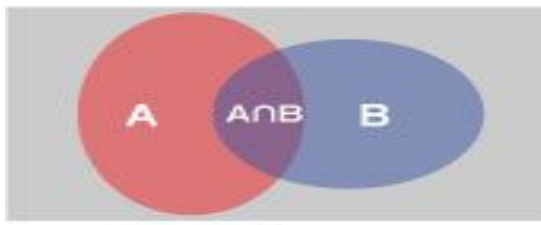


# Seeing Theory

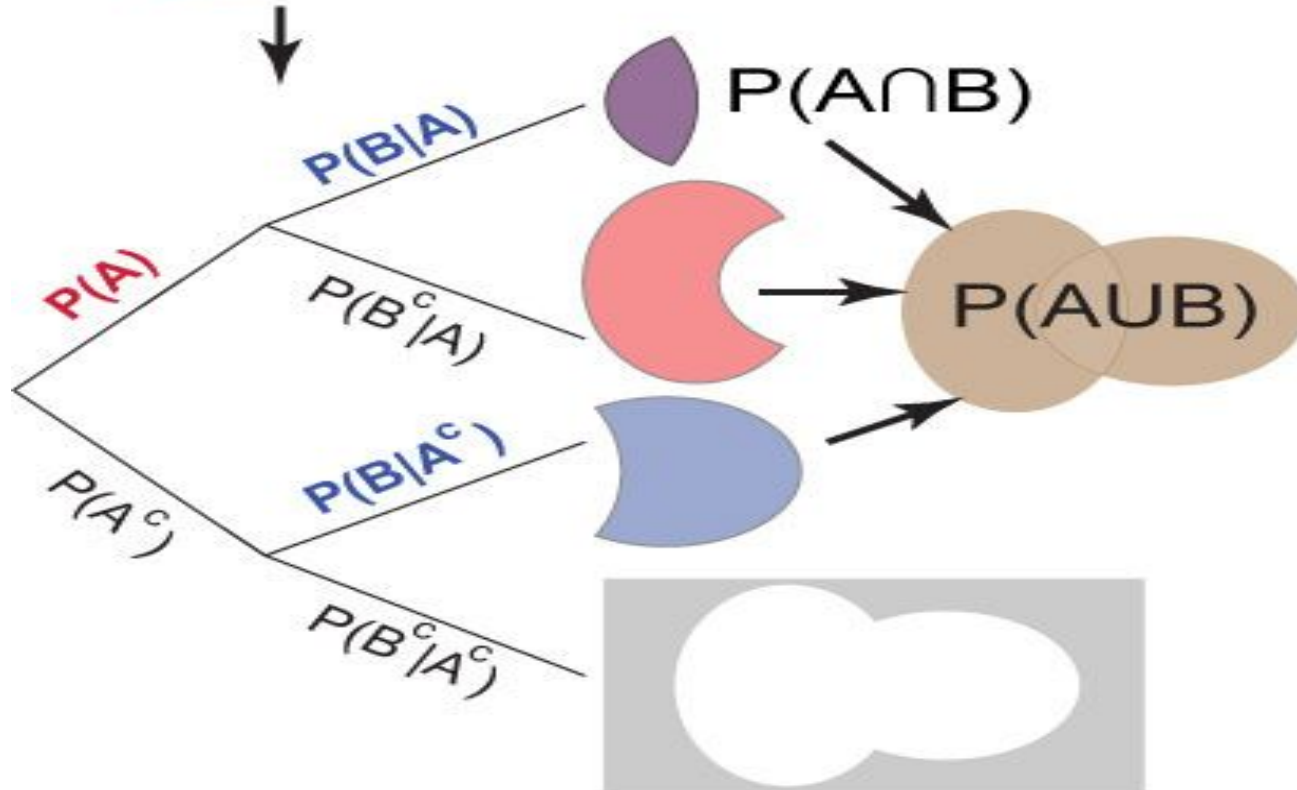
A visual introduction to probability and statistics.

Start





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



# Bayes' Rule or Bayes' Theorem

The diagram illustrates the derivation of Bayes' Theorem. At the top, two basic probability formulas are shown:  $P(B|A) = \frac{P(A \cap B)}{P(A)}$  on the left and  $P(A|B) = \frac{P(B \cap A)}{P(B)}$  on the right. Arrows from these formulas point towards a central equation,  $P(B|A) = \frac{P(A|B)P(B)}{P(A)}$ . An arrow labeled "From the data" points to the  $P(A|B)$  term in the numerator. An arrow labeled "Prior Probability" points to the  $P(B)$  term in the numerator. An arrow labeled "Normalization" points to the  $P(A)$  term in the denominator. The label "Posterior probability" is placed to the left of the central equation, with an arrow pointing to the  $P(B|A)$  term.

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

From the data

Prior Probability

Posterior probability

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

Normalization

What if  $P(A)$  and  $P(B)$  were independent?

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$$P(B|A) = P(B)$$

$$P(A|B) = P(A)$$

# Bayes's Theorem Explained

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- Bayes' Theorem helps us update a belief based on new evidence by creating a new belief.
- Bayes' Theorem helps us revise a probability when given new evidence.
- Bayes' Theorem helps us change our beliefs about a probability based on new evidence.
- Bayes' Theorem helps us update a hypothesis based on new evidence.”

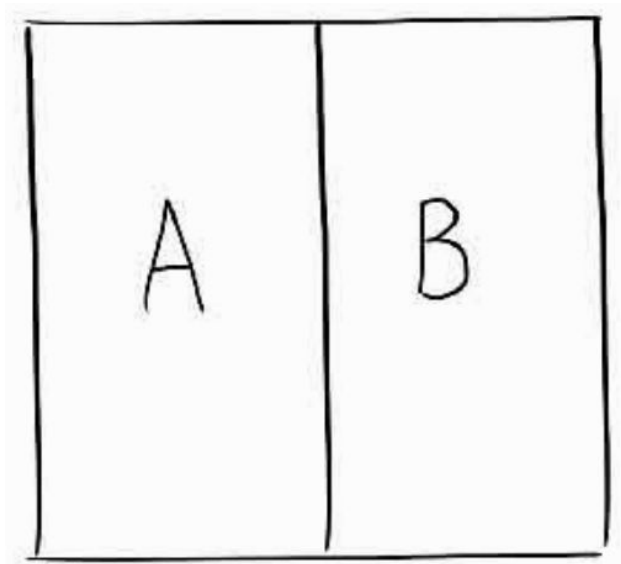
# Bayes's Theorem: here are few examples

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- You just had a test for cancer and it came back positive. What is the probability that you have cancer if the test is positive?
- Your friend claims that stock prices will decrease if interest rates increase. What is the probability stock prices will decrease if interest rates increase?
- You were just pulled over by the police and given a breathalyzer test. It came back positive. What is the probability you are truly drunk, given that the test is positive?

# Bayes's Theorem: here are few examples

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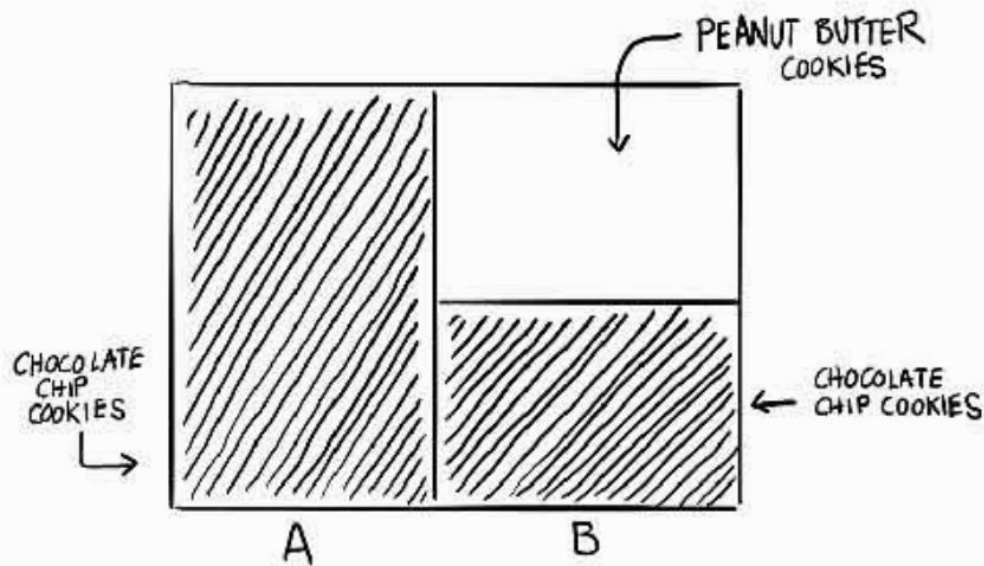


Box A is filled with 10 chocolate chip cookies

Box B is an even mix of 5 peanut butter cookies and 5 chocolate chip cookies

# Bayes's Theorem: here are few examples

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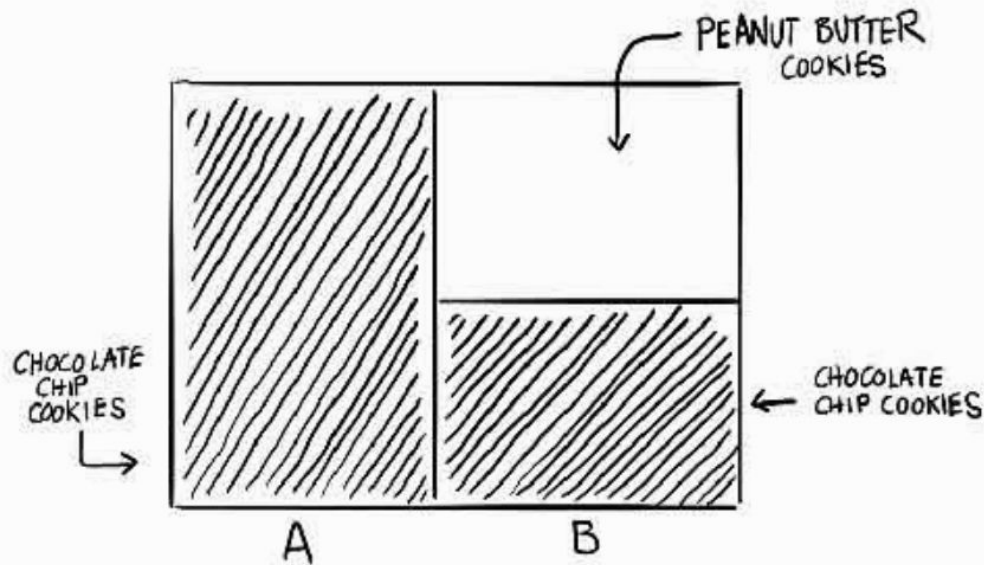
Now, what if you were to close your eyes and have both boxes shuffled, and then reach and select a cookie from one of them - and it was chocolate chip?

If you had to guess what box the cookie came from, what box would you select?



# Bayes's Theorem: here are few examples

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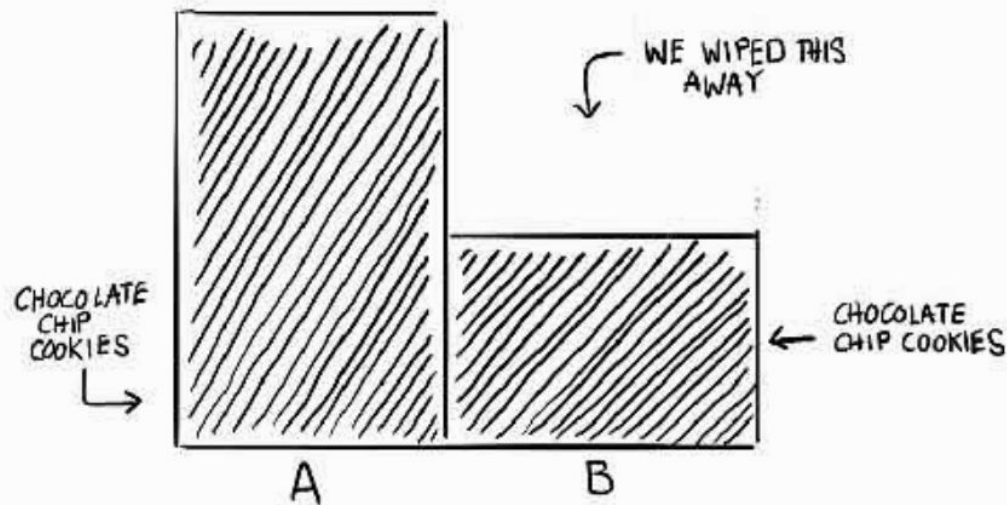


This calculation is a very basic, natural use of Bayes' Theorem.

Given evidence (the amount and type of cookies in each box), you were able to quickly come to the conclusion that Box A has a greater probability of being selected than Box B.

# Bayes's Theorem: here are few examples

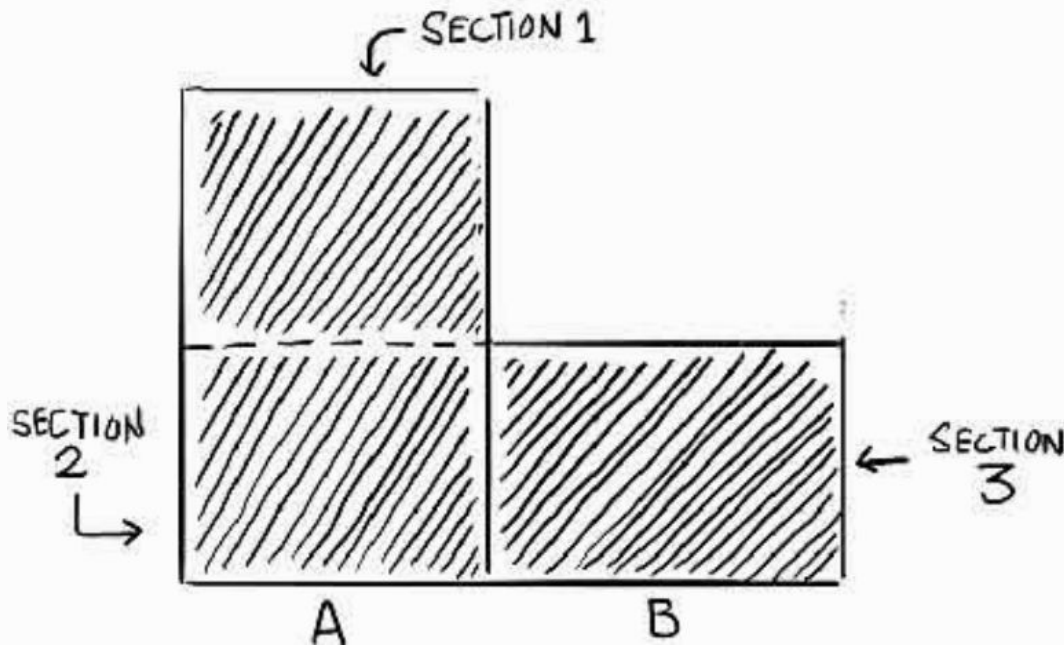
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When your hand selected a chocolate chip cookie, something disappeared: the probability of selecting a peanut butter cookie is now gone.

# Bayes's Theorem: here are few examples

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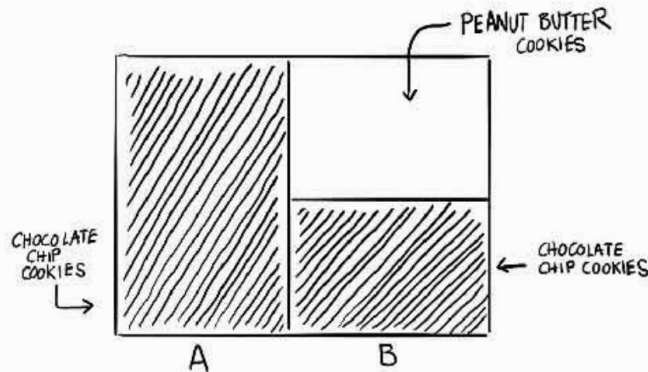


By looking at this, we can see that Box B has a probability of  $\frac{1}{3}$ , or ~33% of being selected.

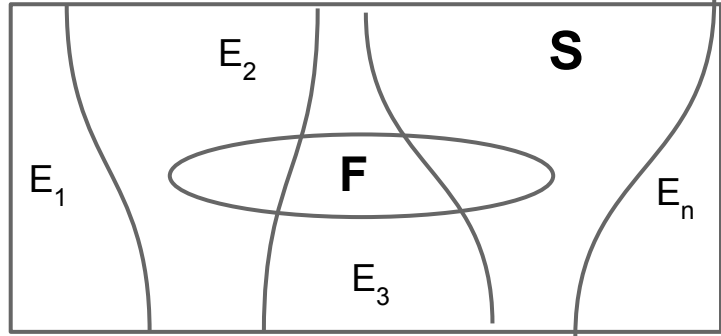
Box A has a probability of  $\frac{2}{3}$ , ~66% of being selected.

# Bayes's Theorem: here are few examples

$$P(\text{BOX A} | \text{CC COOKIE}) = \frac{P(\text{CC COOKIE} | \text{BOX A}) P(\text{BOX A})}{P(\text{CC COOKIE})}$$



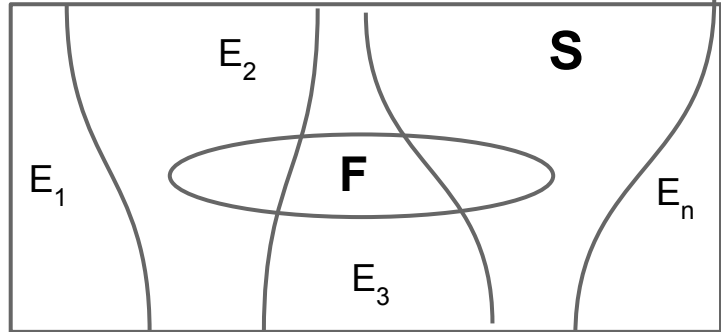
$$\begin{aligned} P(\text{BOX A} | \text{CC COOKIE}) &= \frac{1 \times .5}{.75} \\ &= \frac{.5}{.75} \\ &= 0.\overline{66} \\ &= 66\% \end{aligned}$$



## Law of total probability

Let  $F$  be an event of the sample space  $S$ , and  $E_1, E_2, \dots, E_n$  a partition of  $S$ , we can calculate the probability of  $F$  as follows:

$$\Pr(F) = \sum_{i=1}^n \Pr(F \cap E_i) = \sum_{i=1}^n \Pr(F | E_i) \cdot \Pr(E_i)$$



## Complete Bayes's Rule

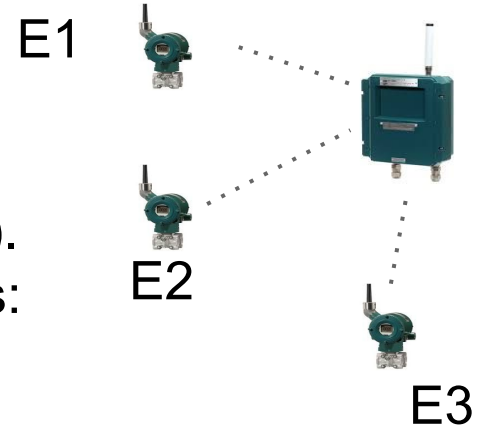
The Law of Total Probability describes that:

$$\Pr(F) = \sum_{i=0}^n \Pr(F \cap E_i) = \sum_{i=0}^n \Pr(F | E_i) \cdot \Pr(E_i)$$

$$\Pr(E_j | F) = \frac{\Pr(F \cap E_j)}{\Pr(F)} = \frac{\Pr(F \cap E_j)}{\sum_{i=0}^n \Pr(F \cap E_i)} = \frac{\Pr(F | E_j) \cdot \Pr(E_j)}{\sum_{i=0}^n \Pr(F | E_i) \cdot \Pr(E_i)}$$

# Example

A gateway receives data from 3 devices (E1, E2, E3).  
The proportion of packets generated by equipment is:  
E1 = 20%, E2 = 30% and E3 = 50%.

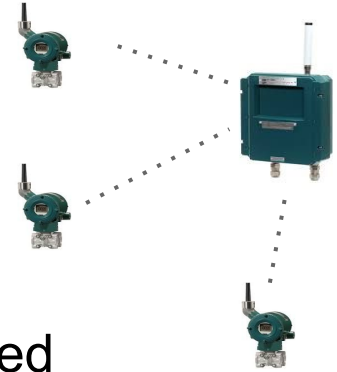


The fraction of corrupted packets is: E1 = 5%, E2 = 3%, E3 = 1%

If a package is chosen at random, knowing that the package is corrupted, how likely is that the package originated from the E3 device?

# Example

- Assume  $E_i$  to be the probability of a equipment is to be randomly selected.
- Assume  $C$  as the probability of a packet being corrupted



$$\Pr(E_1) = 0.20 \quad \Pr(E_2) = 0.30 \quad \Pr(E_3) = 0.50$$

The conditional probabilities  $\Pr(C | E_i)$  have already been given by the question  
 $\Pr(C|E_1) = 0.05 \quad \Pr(C|E_2) = 0.03 \quad \Pr(C|E_3) = 0.01$

We need to find  $P(C)$ :

$$\Pr(C) = \sum_i \Pr(C|E_i) \Pr(E_i) = 0.05 \times 0.20 + 0.03 \times 0.30 + 0.01 \times 0.50 = 0.024 = 2.4\%$$

Thus:

$$\Pr(E_3|C) = \Pr(C|E_3) \times \Pr(E_3) / \Pr(C) = 0.01 \times 0.50 / 0.024 = 0.2083 = 20.83\%$$



# Data driven example

<https://towardsdatascience.com/bayes-rule-applied-75965e4482ff>