

BAYES' THEOREM



BAYES THEOREM

- one of the most famous equations in statistics and probability

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

A FRAMEWORK FOR UPDATING OUR BELIEFS

- What's the point of probability ? => **decision making under uncertainty**
- Our knowledge about the world is never totally exact, so how do we decide whether to go ahead with a decision or not ?
- **Bayes' Theorem gives us a quantitative framework for updating our beliefs as the facts around us change...**

THE INTUITION WITH AN EXAMPLE

- It's 9AM on Monday morning, and you receive an email from your boss. You notice that it seems a little different from her usual notes: the message contains several grammatical errors, and ends by asking you to provide your social security number. Though you first assumed it was a legitimate email, the grammar mistakes and suspicious request convince you to send it right to the spam folder.

(<https://medium.com/opex-analytics/bayes-theorem-101-6a9a1ea5d4a6>)

- When making that quick decision to ignore the email from your “boss,” you unconsciously estimated several different probabilities.
 - First, you judged the likelihood of a work email’s legitimacy to be fairly high.
 - But then you assessed the probability that such a weird email could come from your boss to be low. You also have some general sense that phishing emails tend to be weird in a few specific ways, and you know that phishing scams are common enough that this particular email could plausibly be harmful.

- With all this information swirling around in your head, you decide that the email is most likely spam.
- That's pretty much all bayes theorem is: **updating** our prior beliefs given some particular piece of information.

TAKING A CLOSER LOOK AT THE FORMULA

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

- **P(A|B)** – is the probability of A given that B has already happened.
- **P(B|A)** – is the probability of B given that A has already happened. It looks circular and arbitrary for now...
- **P(A)** – is the unconditional probability of A occurring.
- **P(B)** – is the unconditional probability of B occurring.

TAKING A CLOSER LOOK AT THE FORMULA

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

$P(A|B)$ is a conditional probability – one that measures probability over only certain states of the world (states where B has occurred).

$P(A)$ is an example of an unconditional probability and is measured over all states of the world.

VIDEO EXPLANATION

- <https://www.youtube.com/watch?v=HZGCoVF3YvM&t=542s>

PRACTICE PROBLEM 1

Using Bayes for NLP to predict spams based on the content of an email.

- Assume that the word *offer* occurs in 80% of the spam messages
- Also assume *offer* occurs in 10% of desired e-mails (hams)
- If 30% of the received e-mails are considered to be spam, and I receive a new message which contains *offer*, what is the probability that this new email is a spam?
- Draw a tree diagram to help you and consider the case where you have a sample of 100 emails : you can first find the solution by counting, and then try and find it using the theorem.

100 e-mails

$$100 * 0.3 = 30$$

$$100 * 0.7 = 70$$

Spam

Desired

contains
'offer'

NOT contains
'offer'

contains
'offer'

NOT contains
'offer'

$$30 * 0.8 = 24$$

$$30 * 0.2 = 6$$

$$70 * 0.1 = 7$$

$$70 * 0.9 = 63$$

PRACTICE PROBLEM 1

- $P(\text{contains } offer | spam) = 0.8$ (given in the question)
- $P(spam) = 0.3$ (given in the question)
- $P(\text{contains } offer) = 0.3 * 0.8 + 0.7 * 0.1 = 0.31$

$$P(spam | \text{contains } offer) = \frac{P(\text{contains } offer | spam) * P(spam)}{P(\text{contains } offer)}$$

PRACTICE PROBLEM 1

- Both results should be the same :

$$P(\text{spam}|\text{contains offer}) = \frac{0.8 * 0.3}{0.31} = 0.774$$

PRACTICE PROBLEM 2

- Covid-19 tests are common nowadays, but some test results can be wrong...
- Let's assume:
 - a diagnostic test has 99% accuracy
 - and 60% of all people have Covid-19.
- If a patient tests positive, what is the probability that they actually have the disease?
- Same as previously: take a sample of 100 patients first and find the probability using counts and then use the theorem.

100 units

$$100 * 0.6 = 60$$

COVID-19

True

Diagnose
(positive)

$$60 * 0.99 = 59.4$$

False

Diagnose

$$60 * 0.01 = 0.6$$

$$100 * 0.4 = 40$$

NOT COVID-19

True

Diagnose

$$40 * 0.99 = 39.6$$

False

Diagnose
(positive)

$$40 * 0.01 = 0.4$$

$$P(\text{covid19}|\text{positive}) = \frac{P(\text{positive}|\text{covid19}) * P(\text{covid19})}{P(\text{positive})}$$

- $P(\text{positive}|\text{covid19}) = 0.99$
- $P(\text{covid19}) = 0.6$
- $P(\text{positive}) = 0.6*0.99+0.4*0.01=0.598$

$$P(\text{covid19}|\text{positive}) = \frac{0.99 * 0.6}{0.598} = 0.993$$

PRACTICE PROBLEM 3 (MONTY HALL IS BACK)

- You're on a gameshow called "**Let's Make a Deal**". There are 3 closed doors in front of you.
- Behind each door is a prize. One door has a **car**, one door has **breath mints**, and one door has a **bar of soap**. You'll get the prize behind the door you pick, but you don't know which prize is behind which door. Obviously you want the car!
- Imagine you pick **door A**.
- After picking **door A**, the host of the show, Monty Hall, now opens **door B**, revealing a bar of soap. He then asks you if you'd like to change your guess. Should you?
- By working through Bayes Theorem, we can calculate the actual odds of winning the car if we stick with **door A**, or switch to **door C**.

PRACTICE PROBLEM 3

- The posteriors we want to compute :

1. $P(\text{prize}=A|\text{opened}=B)$ vs. 2. $P(\text{prize}=C|\text{opened}=B)$

PRACTICE PROBLEM 3

- **Priors**

- The probability of any door being correct before we pick a door is $1/3$. Prizes are randomly arranged behind doors and we have no other information. So the **prior**, $P(A)$, of any door being correct is **$1/3$** .

1. $P(\text{prize} = A)$, the prior probability that door A contains the car = $1/3$
2. $P(\text{prize} = C)$, the prior probability that door C contains the car = $1/3$

PRACTICE PROBLEM 3

- **Likelihood**

- If the car is behind door A, then Monty can open door B or C. So the probability of opening either is 50%.

1. $P(\text{opens} = B | \text{prize} = A) = \frac{1}{2}$, the likelihood Monty opened door B if door A is correct

- If the car is in fact behind door C then Monty can only open door B. He cannot open A, the door we picked. He also cannot open door C because it has the car behind it.

2. $P(\text{opens} = B | \text{prize} = C) = 1$, the likelihood Monty opened door B if door C is correct

PRACTICE PROBLEM 3

- **Numerator: $P(A) \times P(B|A)$**
- $P(\text{prize} = A) \times P(\text{opens} = B | \text{prize} = A) = 1/3 \times 1/2 = 1/6$
- $P(\text{prize} = C) \times P(\text{opens} = B | \text{prize} = C) = 1/3 \times 1 = 1/3$

PRACTICE PROBLEM 3

- **Normalize**

- This is the marginal probability $P(\text{opens}=B)$ which is the total probability, removing dependence from any event:
 - In this case:

$$\sum P(\text{opens} = B | \text{prize} = A)P(\text{prize} = A), P(\text{opens} = B | \text{prize} = C)P(\text{prize} = C)$$

$$P(\text{opens} = B) = \frac{1}{6} + \frac{1}{3} = \frac{1}{2}$$

- Putting everything together:

$$1. \quad P(\textit{prize} = A | \textit{opens} = B) = \frac{\frac{1}{6}}{\frac{1}{2}} = \frac{1}{3}$$

$$2. \quad P(\textit{prize} = C | \textit{opens} = B) = \frac{\frac{1}{3}}{\frac{1}{2}} = \frac{2}{3}$$

=> the prize is more likely to be hidden behind door C, so we should switch !