

# Computational Methods for Linguists

## Ling 471

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# Reminders

- Assignment 3 due today
  - ...how are people doing?
- Blog responses due today
- Assignment 4 will be published soon
  - I will send out an additional announcement
  - due date moved to May 25

May 20	Working with linguistic corpora	TBA	
May 25	Visualization and Communication	TBA	Assignment 4
May 27	Visualization and Communication	TBA	Blogs 5
June 1	Presentations		
June 3	Presentations		
June 8			Assignment 5

From class syllabus



# Corrections

Thank you!

- **Age** is of course **not** a Gaussian
  - Thanks for doubting!
  - You can imagine situations where it will be but it is very different from e.g. height
  - Other examples of actually Gaussian stuff:
    - amount of hair on people's heads
    - weight
    - age when children acquire syntax
- **"Discrete"** variable is not spelled "discreet" :)





# Plan for today

- Precision and Recall review
- Theory:
  - The Bayes Theorem
    - Activity
  - Next week: Naive Bayes classification algorithm
- Practice:
  - Packages:
    - pip
    - pandas and dataframes

# Precision and Recall review

- Context: Object apple 🍏 retrieval
  - Array of objects: [0, 1, 2,3, 4,5, 6, 7]
  - Ground Truth: [🍏🍏🍏🍏🍊🍊🍊🍊]
  - Our System: [🍏🍊🍏🍏🍊🍊🍏🍊]
- Reference table for the **four types of label**
- True Positive:** 0,2,3
- False Positive:** 6
- True Negative:** 4,5,7
- False Negative:** 1
- Compute Precision and Recall as per **definitions**

	Predicted class POSITIVE (spam 📧 )	Predicted class NEGATIVE (normal 📧 )	
Actual class POSITIVE (spam 📧 )	TRUE POSITIVE (TP) 📧 📧 320	FALSE NEGATIVE (FN) 📧 📧 43	$\text{Recall} = \frac{TP}{TP + FN}$ $= \frac{320}{320 + 43} = 0.882$
Actual class NEGATIVE (normal 📧 )	FALSE POSITIVE (FP) 📧 📧 20	TRUE NEGATIVE (TN) 📧 📧 538	
	$\text{Precision} = \frac{TP}{TP + FP}$ $= \frac{320}{320 + 20} = 0.941$		

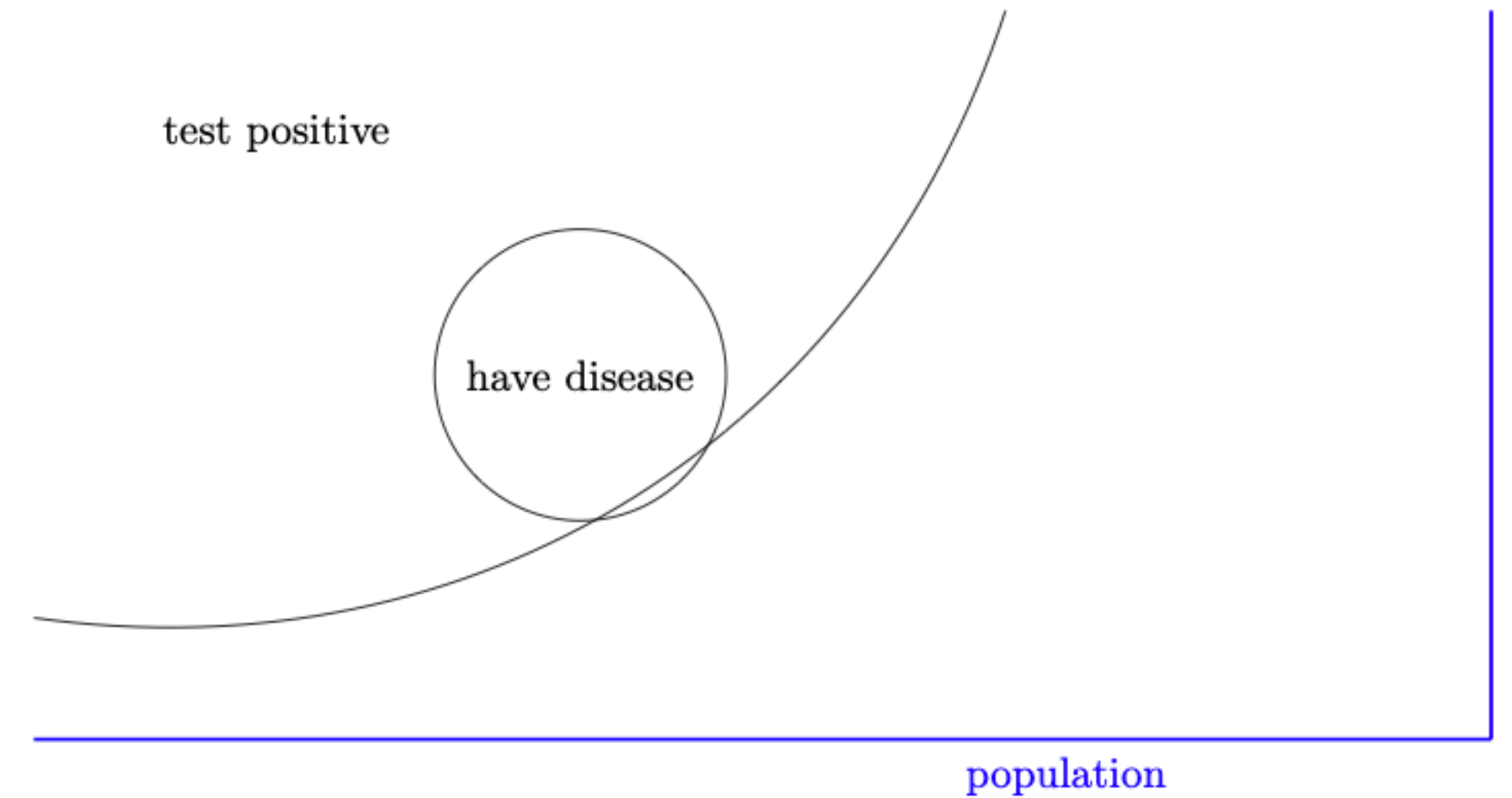
<https://www.knime.com/blog/from-modeling-to-scoring-confusion-matrix-and-class-statistics>

		True Class	
		Positive	Negative
Predicted Class	Positive	TP	FP
	Negative	FN	TN

<https://towardsdatascience.com/confusion-matrix-for-your-multi-class-machine-learning-model-ff9aa3bf7826>

# A classic example (teaser)

- Suppose:
  - 1% of population have cancer
  - 80% of tests detect it correctly while 20% of tests fail to detect it (“false negative”)
  - 9.6% of tests detect it when it is not there (“false positive”) while 90.4% correctly return negative
- Q: If you get a positive result, what is the probability of you having the disease?
  - Many people say “80%”
  - ...but that is not so:
    - the event of “testing” is separate from the event of “having the disease”!
    - they have different probabilities!
    - Stay tuned.



<https://towardsdatascience.com/3-ways-to-think-about-bayes-rule-b6f5b4ef87d6>

# Bayes Theorem

# Bayes Theorem

## in probability theory

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

The Bayes Theorem

- Recall:

- Conditional probability

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

- notation:  $A \cap B$  = "A and B" both occurred

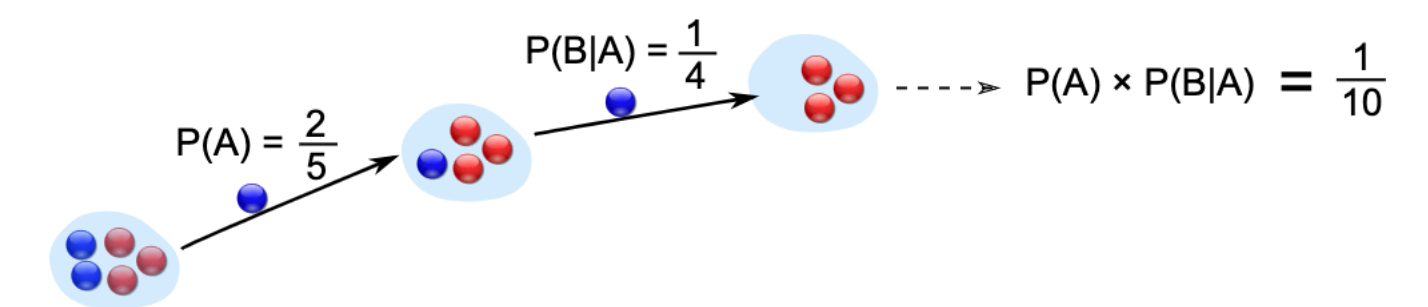
- "Intuition":

- How many times I saw A after I also saw B?
    - Derive the formula from the marble example
    - Sequence of A and B => product of P(A) and P(B|A)
    - ...then just rewrite the equation to express P(B|A) in terms of P(A and B) and P(A)

- By the way:

- In a sequence of two marble draws, what's P(second marble is blue)?
  - call it **P(A)**
  - $P(A) = P(\text{second is blue}) \cdot \mathbf{P(\text{first is blue})} + P(\text{second is blue}) \cdot \mathbf{P(\text{first is red})}$
  - The first marble is there!**

So the probability of getting **2 blue marbles** is:



And we write it as

$$P(\text{A and B}) = P(\text{A}) \times P(\text{B} | \text{A})$$

*"Probability Of"* (above P(A and B))  
*"Given"* (above P(B | A))  
*Event A* (below P(A))  
*Event B* (below P(B | A))

"Probability of **event A and event B** equals the probability of **event A** times the probability of **event B given event A**"

Conditional probability

<https://www.mathsisfun.com/data/probability-events-conditional.html>



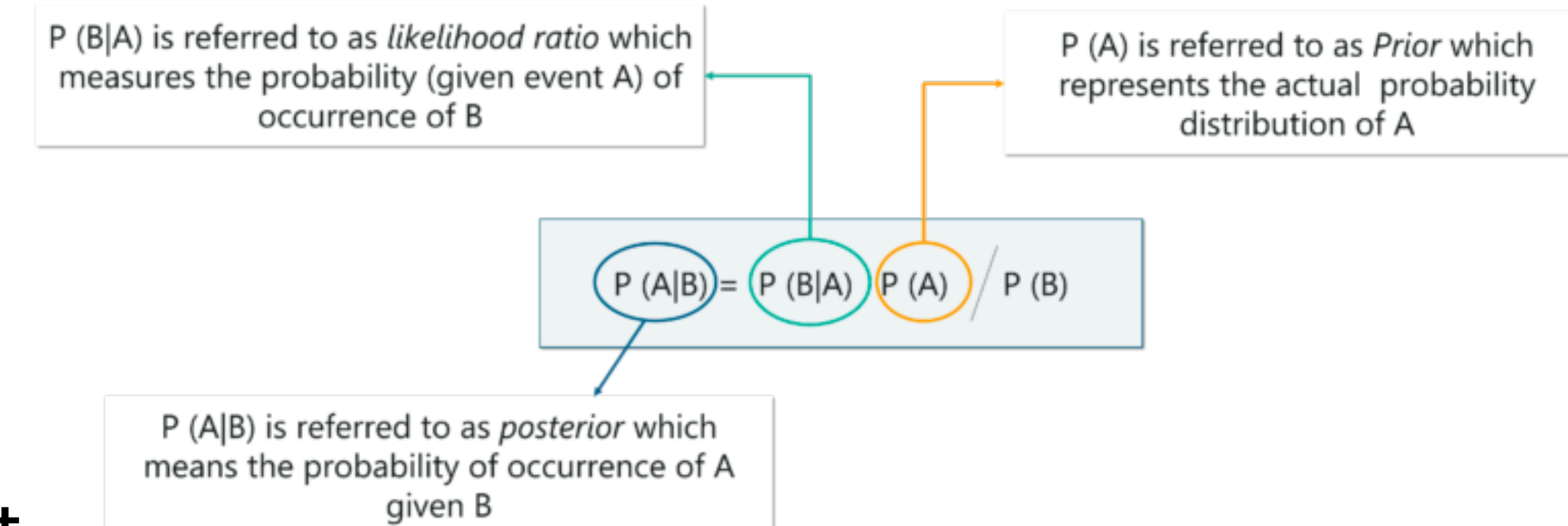
# Bayes Theorem

## in probability theory

- The Bayes Theorem is related to **conditional probability** and to **sequences**
- Intuition:
  - “**Bayes rule** provides us with a way to update our **beliefs** based on the arrival of new, relevant pieces of **evidence**.” (Devin Soni)

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

The Bayes Theorem



<https://www.edureka.co/blog/statistics-and-probability/#Bayes%20Theorem>

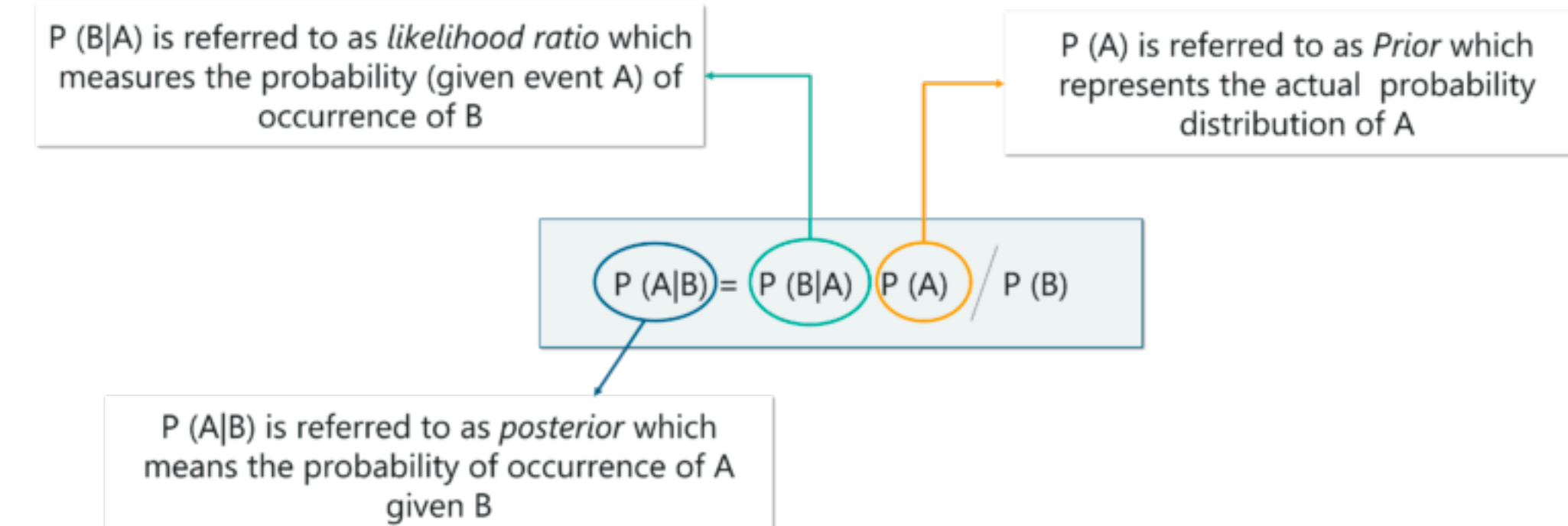
# Bayes Theorem

## derivation

- Note:
  - $P(A,B) = P(A|B) \cdot P(B)$
- Note:
  - $P(B,A) = P(B|A) \cdot P(A)$
- Note:
  - $P(A,B)$  is the same as  $P(B,A)$ !
- Therefore:
  - $P(A|B) \cdot P(B) = P(B|A) \cdot P(A)$
- Therefore:
  - $P(A|B) = P(B|A) \cdot P(A)$  divided by  $P(B)$ !
- Why is this derivation meaningful/interesting?
  - Sometimes, we **know**  $P(B|A)$  but **not**  $P(A|B)$ !

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

The Bayes Theorem



<https://www.edureka.co/blog/statistics-and-probability/#Bayes%20Theorem>

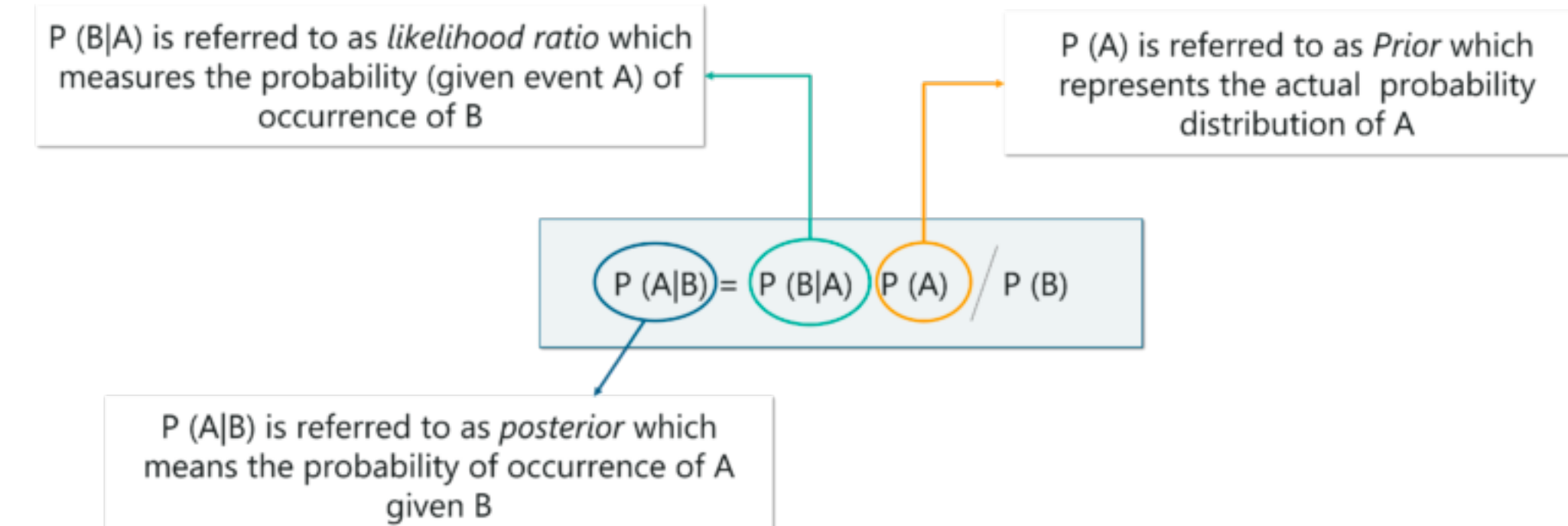
# Bayes Theorem

## an example

- Suppose:
- $P(\text{having cancer}) = 0.05$ 
  - (5% of people have it)
  - $P(A)$
- $P(\text{be a smoker}) = 0.10$ 
  - (10% of people smoke)
  - $P(B)$
- $P(\text{smoker}|\text{cancer}) = 0.20$ 
  - (20% of those who have cancer are smokers)
  - $P(B|A)$
- Find:  $P(\text{cancer}|\text{smoker})$ :
  - $P(\text{cancer}|\text{smoker}) = 0.20 * 0.05 / 0.10 = 0.10$

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

The Bayes Theorem



<https://www.edureka.co/blog/statistics-and-probability/#Bayes%20Theorem>



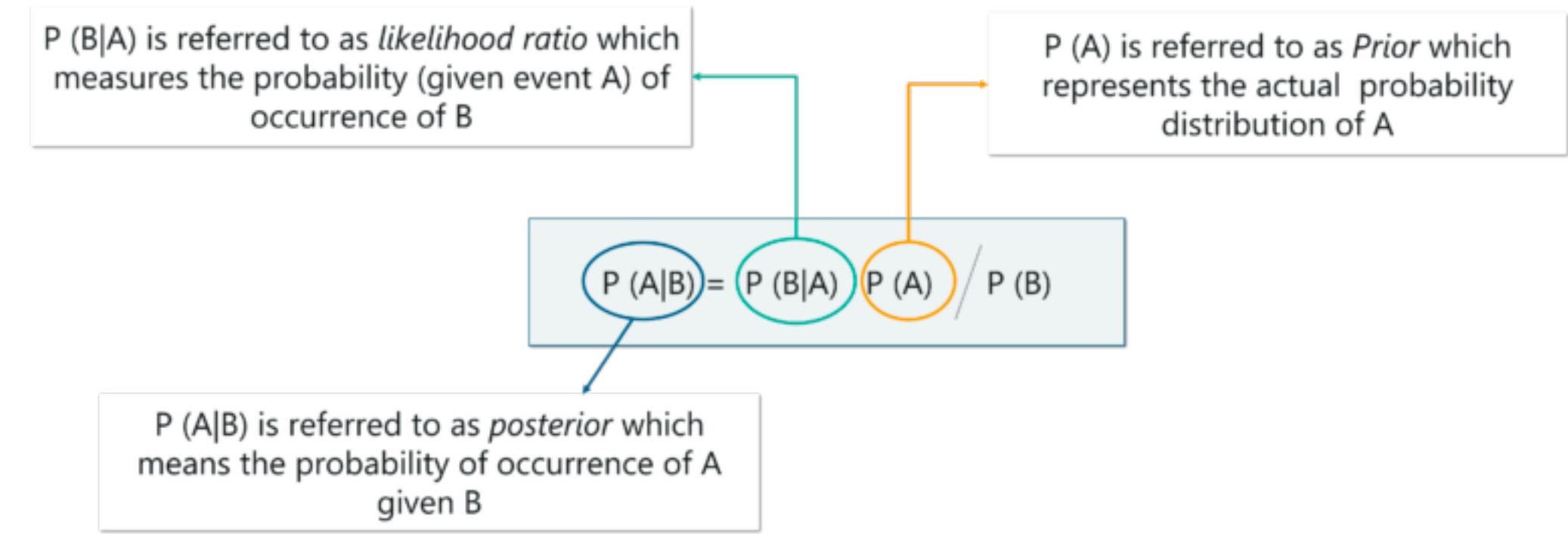
# Bayes Theorem

## a classic example

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

The Bayes Theorem

- Suppose:
  - 1% of population have cancer
  - 80% of tests detect it correctly while 20% of tests fail to detect it ("false negative")
  - 9.6% of tests detect it when it is not there ("false positive") while 90.4% correctly return negative
- Q: If you get a positive result, what is the probability of you having the disease?
  - Work it out in a **group activity**: <https://olzama.github.io/Ling471/assignments/activity-May6.html>
  - Hint: "P(B) is the P(positive test). But P(positive test) is **not directly given** to you!"
    - Positive test outcome** means: [the test is positive AND person has cancer] **OR** [the test is positive and there is NO cancer!]
    - Use the marbles example: P(**two** events) is similar to P(**two** marbles)



So the probability of getting **2 blue marbles** is:

And we write it as

"Probability Of"  $P(\mathbf{A} \text{ and } \mathbf{B}) = P(\mathbf{A}) \times P(\mathbf{B} | \mathbf{A})$  "Given"

Event A Event B

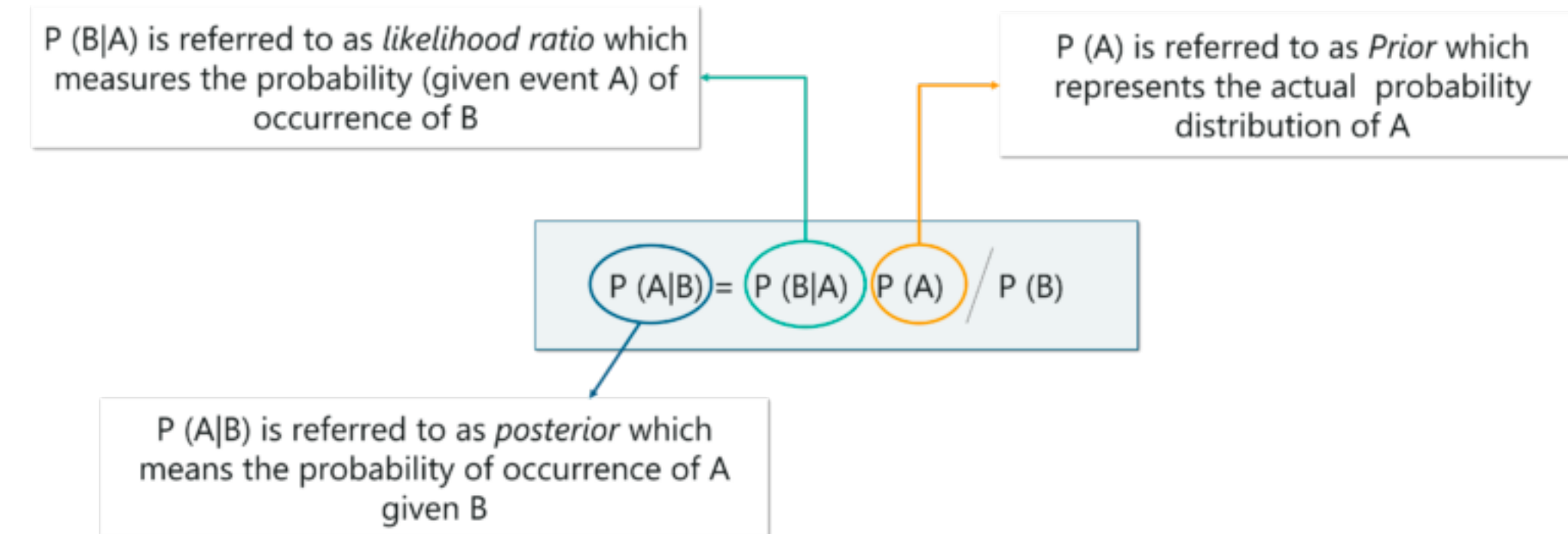
"Probability of **event A and event B** equals the probability of **event A** times the probability of **event B given event A**"

# Bayes Theorem

## a classic example

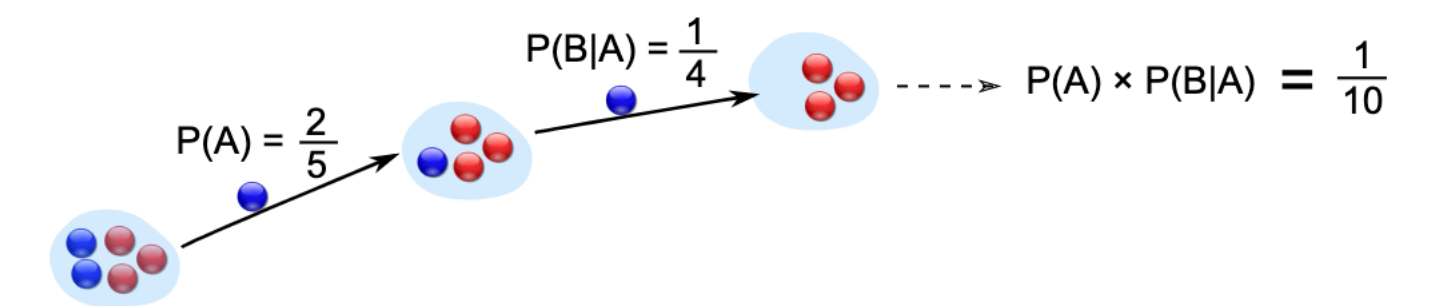
$$P(A | B) = \frac{P(B | A) \cdot P(A)}{P(B)}$$

The Bayes Theorem



- Suppose:
  - 1% of population have cancer
  - 80% of tests detect it correctly while 20% of tests fail to detect it ("false negative")
    - This 80% is out of people who **do** have the disease!
  - 9.6% of tests detect it when it is not there ("false positive") while 90.4% correctly return negative
- **Q:** If you get a positive result, what is the probability of you having the disease?
  - **Answer: 7.8%**
  - Useful reading: <https://towardsdatascience.com/3-ways-to-think-about-bayes-rule-b6f5b4ef87d6>

So the probability of getting **2 blue marbles** is:



And we write it as

"Probability Of" "Given"

$$P(\text{A and B}) = P(\text{A}) \times P(\text{B} | \text{A})$$

Event A Event B

"Probability of **event A and event B** equals the probability of **event A** times the probability of **event B given event A**"

# Dataframes and pandas package



# Installing packages with pip

- We will need several packages for next HW
- They are best installed via pip
- pip is included in python distribution (starting from python **3.8**)
  - **Usually**, it just works
  - Some people are having issues on Windows 10
  - See instructions here:
    - <https://phoenixnap.com/kb/install-pip-windows>
  - ...and here:
    - <https://stackoverflow.com/questions/23708898/pip-is-not-recognized-as-an-internal-or-external-command>
- In any case, start with checking whether you have pip already
  - `pip --version`
  - `python -m pip --version`
  - `py -m pip --version`



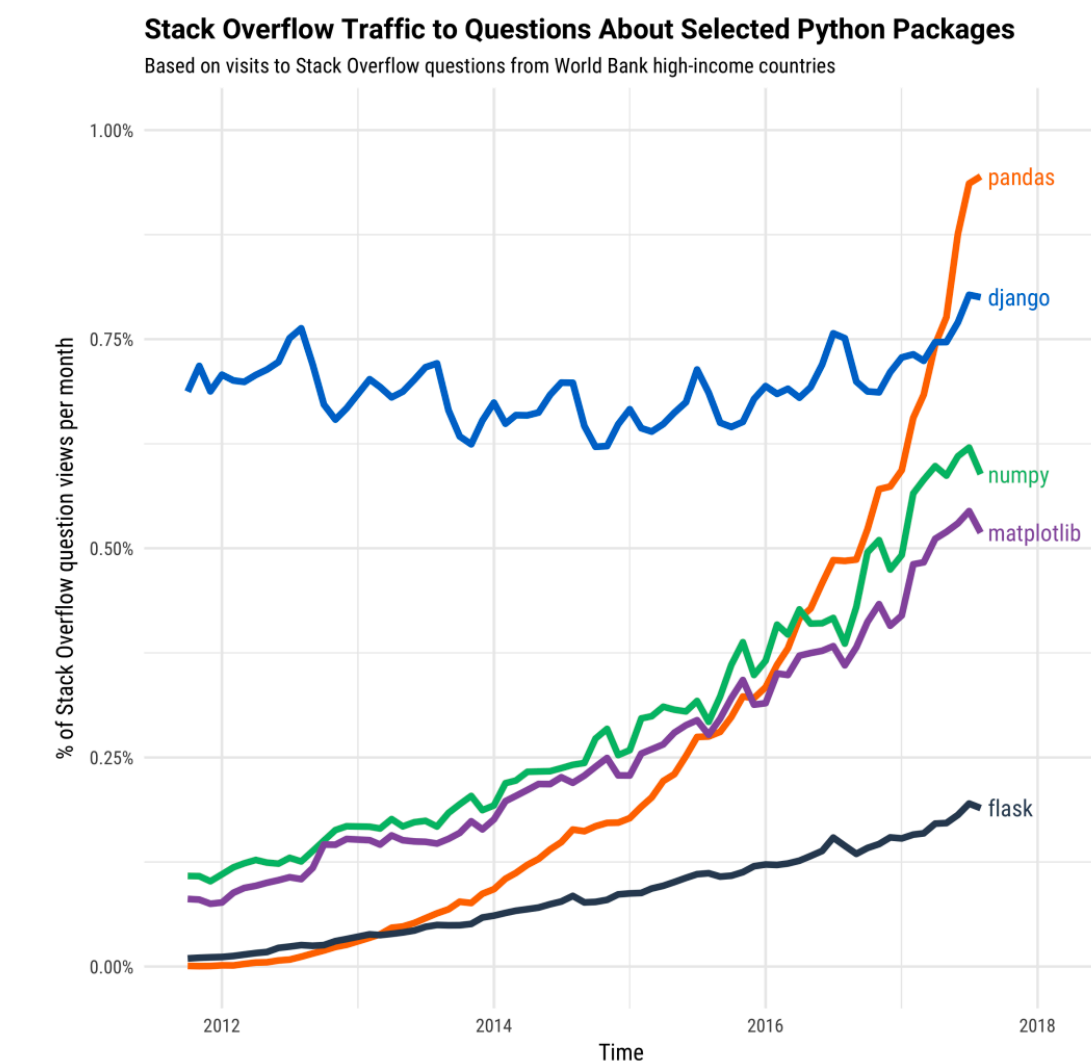
# Pandas

a popular data science package

- Stores data in convenient tables
- Allows for fast data access and manipulation
- Why store data as tables?
  - In data science/statistics/machine learning:
    - you work with “observations” (=data points)
    - each data point is a row
    - What are columns?
      - data point can have different features
      - e.g. word counts!



<https://www.kdnuggets.com/2020/03/python-pandas-data-discovery.html>



# Data as tables

- **rows:**
  - observations, datapoints
- **columns:**
  - “features”
  - can be many or few!
- Many ML algorithms involve linear algebra
  - Linear algebra includes matrix multiplication
    - Matrices are tables!

	Name	Team	Number	Position	Age
0	Avery Bradley	Boston Celtics	0.0	PG	25.0
1	John Holland	Boston Celtics	30.0	SG	27.0
2	Jonas Jerebko	Boston Celtics	8.0	PF	29.0
3	Jordan Mickey	Boston Celtics	NaN	PF	21.0
4	Terry Rozier	Boston Celtics	12.0	PG	22.0
5	Jared Sullinger	Boston Celtics	7.0	C	NaN
6	Evan Turner	Boston Celtics	11.0	SG	27.0

<https://www.geeksforgeeks.org/python-pandas-dataframe/>

	label	text
0	1	For a movie that gets no respect there sure ar...
1	1	Bizarre horror movie filled with famous faces ...
2	1	A solid if unremarkable film Matthau as Einste...
3	1	Its a strange feeling to sit alone in a theate...
4	1	You probably all already know this by now but ...

```
0: <review_vector.reviewVec object at 0x7ffc703b3be0>  
> special variables  
correct_label: 'POSITIVE'  
text: 'good good good bad bad '
```



# Data as tables

- **rows:**
  - observations, datapoints
- **columns:**
  - “features”
  - can be many or few!
- The specific **dimensions** are crucial
  - For **any ML** algorithm:
    - need to know very well **how many columns** you have
      - (sometimes also rows, but that’s less important for us)
- In pandas, columns can have **names**
  - which allows convenient querying
  - the “names” row is **ignored**
    - it is not an “observation”/datapoint

	<i>Name</i>	<i>Team</i>	<i>Number</i>	<i>Position</i>	<i>Age</i>
0	Avery Bradley	Boston Celtics	0.0	PG	25.0
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```
label                                     text
0      1  For a movie that gets no respect there sure ar...
1      1  Bizarre horror movie filled with famous faces ...
2      1  A solid if unremarkable film Matthau as Einste...
3      1  Its a strange feeling to sit alone in a theate...
4      1  You probably all already know this by now but ...
```

# pandas demo

# Lecture survey: in the chat