

# Naive Bayes

Supervised Algorithm

START

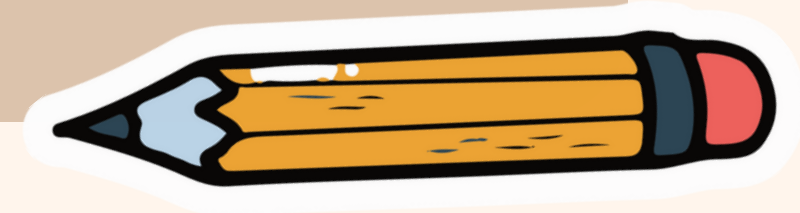
# Background



Naive Bayes is classification algorithm based on Bayes' Theorem.

It assumes that all the features that predict the target value are independent of each other.

It works with huge dataset and mostly used to solve text kinds of data.



# What it does?

**1.**

Calculates probability of each class

**2.**

Picks one with highest probability

# Bayes Theorem

Likelihood of the Evidence  
given that the Hypothesis is  
True

Prior Probability of the  
Hypothesis

$$P(H \setminus E) = \frac{P(E \setminus H) * P(H)}{P(E)}$$

Prior probability of the  
Hypothesis given that the  
Evidence is True

Prior probability that  
the evidence is True

# Problem Statement

**Patient's probability of having liver disease if  
they are an alcoholic.**

If 'A' is the event "Patient has liver disease." Past data tells you that 10% of patients entering your clinic have liver disease.

$$P(A) = 0.10$$

'B' could mean that "Patient is an alcoholic." 5% of the clinic's patients are alcoholics.

$$P(B) = 0.05$$

If you know that among those patients diagnosed with liver disease, 7% are alcoholics.

This is your 'B|A': the probability that a patient is alcoholic, given that they have liver disease, is 7%.

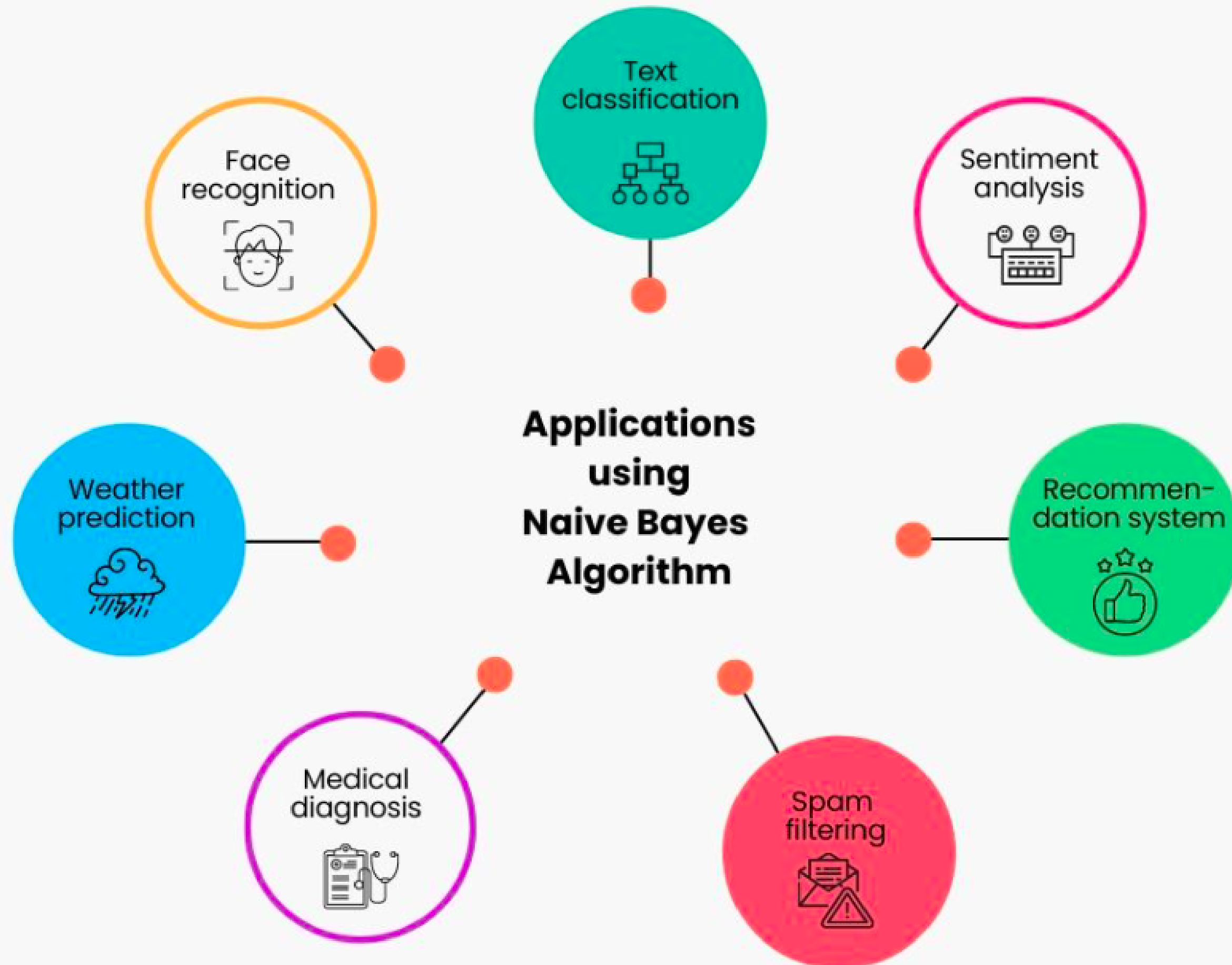
$$P(B|A) = 0.07$$

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

**Bayes' theorem tells you:**

$$P(A|B) = (0.07 * 0.1)/0.05 = 0.14$$

**In other words, if the patient is an alcoholic, their chances of having liver disease is 0.14 (14%).**





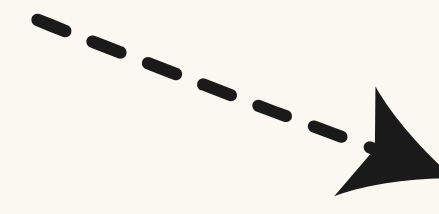
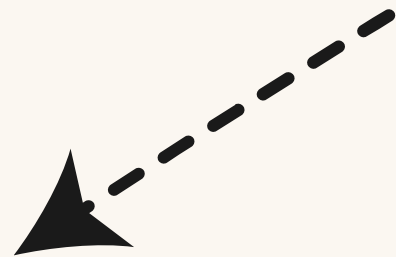
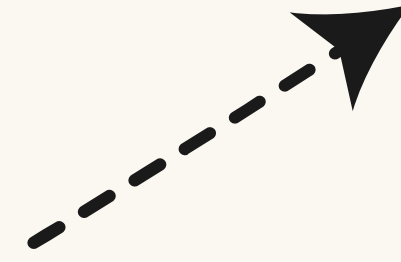
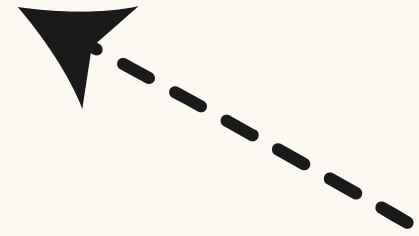
**Gaussian  
Naive Bayes**

**Bernoulli  
Naive Bayes**

# TYPES

**Multinomial  
Naive Bayes**

**Optimal  
Naive Bayes**



# Advantages

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- **Doesn't require larger amounts of training data**
- **Straightforward to implement**
- **Convergence is quicker than other models**
- **Highly scalable**
- **Can handle both continuous and categorical data**