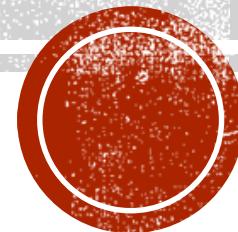


TEXT CLASSIFICATION

BY: Khaled Mohammed

Teaching Assistant at Faculty of Engineering

Khaled.edu.engineer@gmail.com



NAIVE BAYES

- Naïve Bayes is a **probability-based classifier**.
 - Its goal is to find the class **C** (for example: *positive vs negative* sentiment) that is most likely given the words in a document:

$$\hat{C} = \arg \max_c P(c \mid \text{words})$$

- Using Bayes' theorem:

$$P(c \mid w_1, w_2, \dots, w_n) = \frac{P(c)P(w_1, w_2, \dots, w_n \mid c)}{P(w_1, w_2, \dots, w_n)}$$

- Since the denominator is the same for all classes, we compare only:

$$P(c) \cdot P(w_1, w_2, \dots, w_n \mid c)$$



THE NAÏVE ASSUMPTION

- We assume **every word is independent of the others** given the class, So:

$$P(w_1, w_2, \dots, w_n | c) = \prod_{i=1}^n P(w_i | c)$$

- This assumption is **not true in real language**, but it works extremely well because:
 - It simplifies the math
 - Works well in high-dimensional sparse text
 - Word frequencies capture a lot of signal (ex: “good”, “love”, “great” → positive)



HOW NAÏVE BAYES WORKS FOR TEXT

- Given a document with words:
 - I love this movie
- We want to compute:

$$P(\text{positive} \mid \text{document}) \propto P(\text{positive}) \times P(I|\text{pos}) \times P(\text{love}|\text{pos}) \times P(\text{this}|\text{pos}) \times P(\text{movie}|\text{pos})$$

- And the same for negative.
- The class with the **higher** probability is selected.



WHICH NAÏVE BAYES IS USED FOR TEXT?

- We use **Multinomial Naïve Bayes** because it counts words:

$$P(w|c) = \frac{\text{count}(w, c) + 1}{\text{total words in class } c + |V|}$$

- Why Laplace smoothing (+1)?
 - To avoid zeros
 - If a word never appears in a class → its probability becomes 0 → whole product becomes 0
 - Laplace smoothing solves it



DETAILED EXAMPLE

- We have 4 training documents, each belonging to class **c** or **j**:

Doc	Words	Class
1	Chinese Beijing Chinese	c
2	Chinese Chinese Shanghai	c
3	Chinese Macao	c
4	Tokyo Japan Chinese	j

- Test document (Doc 5):
 - Chinese Chinese Chinese Tokyo Japan
- We want to decide: **Is it class c or class j?**



STEP 1: PRIORS

- Because:
 - 3 documents are class **c**
 - 1 document is class **j**

$$\hat{P}(c) = \frac{3}{4}, \quad \hat{P}(j) = \frac{1}{4}$$



STEP 2: COUNT WORDS WITH LAPLACE SMOOTHING

- **Vocabulary:**

- Chinese, Beijing, Shanghai, Macao, Tokyo, Japan
- $|V| = 6$

- **Count total words per class:**

- **For class c**

- Docs 1, 2, 3:
 - Doc1: Chinese Beijing Chinese → 3 words
 - Doc2: Chinese Chinese Shanghai → 3 words
 - Doc3: Chinese Macao → 2 words

Total words = 8

- **For class j**

- Doc 4:
 - Tokyo Japan Chinese → 3 words

Total words = 3



STEP 3: COMPUTE CONDITIONAL PROBABILITIES

$$P(w|c) = \frac{\text{count}(w, c) + 1}{\text{total words in } c + |V|}$$

- For class **c**:

- $\text{count}(\text{Chinese}, c) = 5$
- $\text{count}(\text{Tokyo}, c) = 0$
- $\text{count}(\text{Japan}, c) = 0$

- So:

$$P(\text{Chinese}|c) = \frac{5 + 1}{8 + 6} = \frac{6}{14} = \frac{3}{7}$$

$$P(\text{Tokyo}|c) = \frac{0 + 1}{8 + 6} = \frac{1}{14}$$

$$P(\text{Japan}|c) = \frac{0 + 1}{8 + 6} = \frac{1}{14}$$



CONT.

- For class j:
- Counts from Doc 4:
 - Tokyo: 1
 - Japan: 1
 - Chinese: 1
- Total = 3

$$P(\text{Chinese}|j) = \frac{1+1}{3+6} = \frac{2}{9}$$

$$P(\text{Tokyo}|j) = \frac{1+1}{3+6} = \frac{2}{9}$$

$$P(\text{Japan}|j) = \frac{1+1}{3+6} = \frac{2}{9}$$



STEP 4: COMPUTE PROBABILITY OF CLASS FOR DOC 5

- Document 5:
 - Chinese Chinese Chinese Tokyo Japan
- Number of each word:
 - Chinese ×3
 - Tokyo ×1
 - Japan ×1

- **For class c**

$$P(c|d_5) \propto P(c) \cdot P(\text{Chinese}|c)^3 \cdot P(\text{Tokyo}|c) \cdot P(\text{Japan}|c)$$

- Substitute:

$$= \frac{3}{4} \cdot \left(\frac{3}{7}\right)^3 \cdot \frac{1}{14} \cdot \frac{1}{14}$$

- Numerically ≈ **0.0003**



CONT.

- **For class j**

$$\begin{aligned} P(j|d_5) &\propto P(j) \cdot P(\text{Chinese}|j)^3 \cdot P(\text{Tokyo}|j) \cdot P(\text{Japan}|j) \\ &= \frac{1}{4} \cdot \left(\frac{2}{9}\right)^3 \cdot \frac{2}{9} \cdot \frac{2}{9} \end{aligned}$$

- Numerically ≈ 0.0001



FINAL DECISION

$$P(c|d_5) > P(j|d_5)$$

So the document is classified as **class c**.

