

# K-Nearest Neighbors

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# Introduction

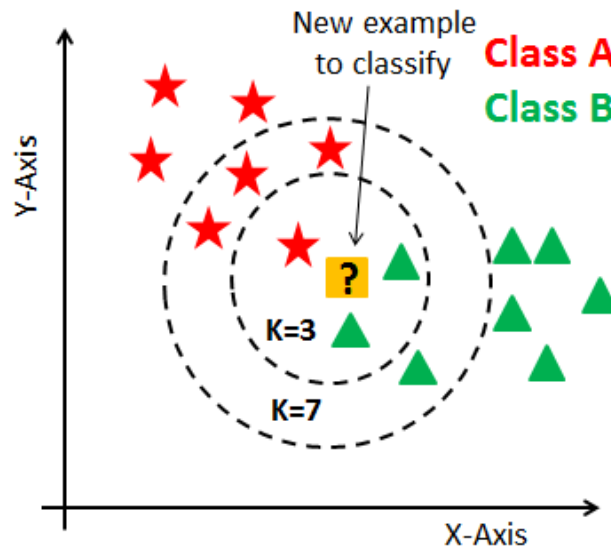
## ❖ Goal

- The HW5 is to implement K-NN model for text classification.
- First, we will briefly explain how the K-NN model works.

# K-NN

## ❖ How The K-NN Model Works?

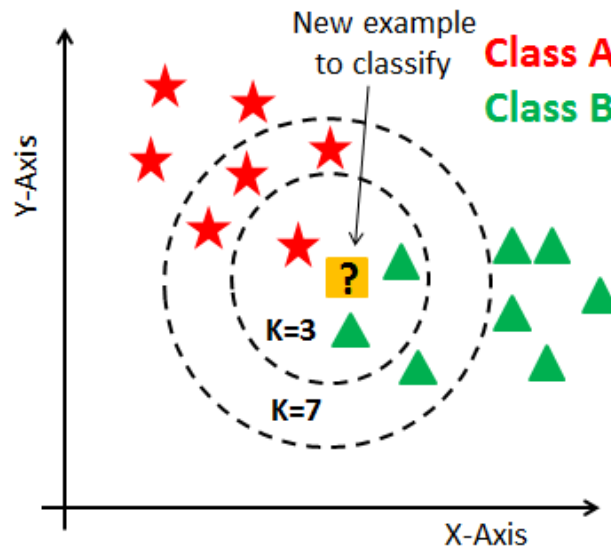
- Assume that there are two classes for the data and that there are already labeled data.
- Let's consider the below figure.



# K-NN

## ❖ How The K-NN Model Works?

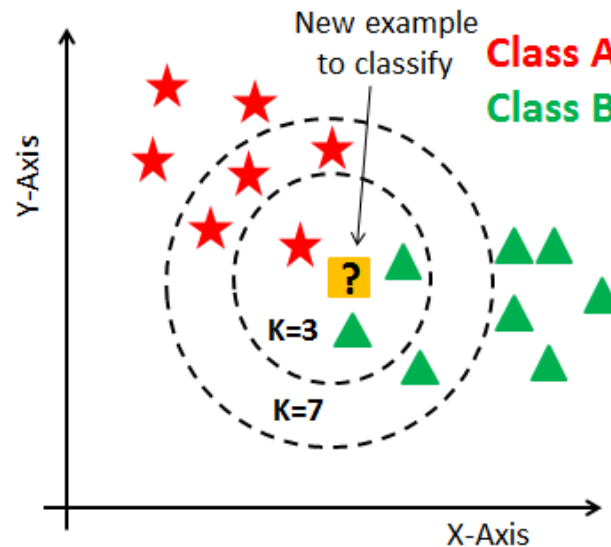
- To assign a label to new data, you have to find K number of nearest data (neighbors).
- The label to be assigned can be different depending on the **K** that you can select.



# K-NN

## ❖ How The K-NN Model Works?

- If you choose 3 as the K, the new data will be assigned as triangle.
- If you choose 7 as the K, the new data will be assigned as star.



## ❖ How The K-NN Model Works?

- Depending on what **distance metric** you want to use, the nearest neighbors can be different and this means that the label to be assigned also can be different.
- There are several metrics for calculating distance between two data points such as Manhattan distance, Euclidean distance, Cosine similarity and so on.

# Assignment

## ❖ Various Possible Inputs

➤ TF.

➤ TF-IDF.

➤ Bag-of-words.

- 1 if a word in document else 0.

- For example,

$doc_a =$  I love dog

$doc_b =$  I like cat

Vocab = { cat, dog, i, like, love }

$vec_a = [ 0, 1, 1, 0, 1 ]$

$vec_b = [ 1, 0, 1, 1, 0 ]$



# Assignment

## ❖ Distance Metrics

- Formula for the distance function
  - Cosine similarity between two vectors:

$$\text{similarity} = \cos(\Theta) = \frac{A \cdot B}{\|A\| \|B\|} = \frac{\sum_{i=1}^n A_i \times B_i}{\sqrt{\sum_{i=1}^n (A_i)^2} \times \sqrt{\sum_{i=1}^n (B_i)^2}}$$

- Euclidean distance between two vectors:

$$D_{\text{euclidean}} = \sqrt{\sum_{i=0}^n (p_i - q_i)^2}$$

where  $p_i \in P, q_i \in Q$  and  $P, Q \in \mathbb{R}^n$ .

# Given Data

## ❖ train.json & test.json

➤ Data Format : a list of dictionaries

➤ Data Example :

```
[  
  {  
    "paragraph" : "Since Game of Throne first aired, ..."  
    "label": "tv",  
    "id": "31"    *this is an article id  
  },  
  ...  
]
```

➤ The number of Labels : 5 (finance, lifestyle, tv, sports, entertainment)

# Assignment

## ❖ Assignment

- Implement the 'K-NN' model for text classification in main.py.
  - Use **TF**, **TF-IDF** and **bag-of-words** vectors as input for the model **respectively**.
    - ✓ You can use the TF-IDF function that you have already implemented for hw3.
  - Use **Euclidean distance** and **cosine similarity** as distance metrics **respectively**.
  - Use **k = 11**.
- See page 13 for specific output format.

# Assignment

## ❖ Submissions

1) StudentName \_StudentID\_main.py (python version 3.x)

- e.g., 홍길동\_2020123123\_hw5\_main.py.  
e.g., MichaelJackson\_2020123123\_hw5\_main.py

2) StudentName \_StudentID.txt

- e.g., 홍길동\_2020123123\_hw5.txt  
e.g., MichaelJackson\_2020123123\_hw5.txt

# Assignment

## ❖ Outlook of the Text File

➤ Be careful of **capitals** and **spellings**.

➤ **Round to second decimal place.**

Metric: Cosine similarity

Input: bag-of-words

Accuracy: 21.21%

Metric: Cosine similarity

Input: TF

Accuracy: 21.21%

*space*

Metric: ☐ Cosine similarity

Input: TF-IDF

Accuracy: 21.21%

*"\n"*

☐

Metric: Eculidean distance

Input: bag-of-words

Accuracy: 21.21%

Metric: Eculidean distance

Input: TF

Accuracy: 21.21%

Metric: Eculidean distance

Input: TF-IDF

Accuracy: 21.21%

# Cautions

## ❖ Cautions

- Use 'Python3' and 'Google Colab'.
- Do not import any library except already imported libraries.
- Copy will be scored 0.

Thank you for your attention!

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<http://nlp.skku.edu/>