# T.C.

# GEBZE TEKNİK ÜNİVERSİTESİ BİLGİSAYAR MÜHENDİSLİĞİ

DATA MINING CSE-454

FINAL PROJECT

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My project TEXT CATEGORIZATION. I separate the text files I have according to the classes I have specified. I have 5 classes. MAGAZINE, ECONOMY, SPORTS, POLITICS, HEALTH. By comparing the input text, it is found out which class it belongs to. I used the K-Nearest Neighbor (KNN) algorithm in my project.

#### **DATASET**

I have a total of 1100 texts for the dataset of this project. There are 230 files in each category. I use 219 of 230 files in each category to fill my vector files. I give the rest of my files as input to the program and find out which category it is in.

### **PREPROCESSING**

```
STOPWORDS = "da,de, ve , vs. , ile , mi , mu | "

idef remove_stopwords(text):
    """custom function to remove the stopwords"""

return " ".join([word for word in str(text).split() if word not in STOPWORDS])
```

Here I clean the data in my text files. The cleaner and simpler my data is, the clearer the results are. I categorize text in my project. Words like "de, da, and, etc." in texts do not mean anything to me. So I remove them from the text and make them get better results.

#### **TFIDF**

I used TFIDF method to convert my words to a numerical value. So my numbers became a numerical value and I managed to aggregate them. All the text files in the 5 categories I have were browsed and the vectors of all the words in them were extracted with TFIDF.

$$\mathrm{idf}(t,D) = \log \frac{N}{|\{d \in D : t \in d\}|} \\ \qquad \mathrm{tf}(t,d) = 0.5 + 0.5 \cdot \frac{f_{t,d}}{\max\{f_{t',d} : t' \in d\}}$$

**TF:** Term Frequency

**IDF:** Inverse Document Frequency

Term frequency; is the division of our selected term by the total number of terms found in the text.

Reverse Document Frequency; It shows how many terms we have in our texts. It is the logarithm of our total text quantity to the text quantity that contains the term.

#### **KNN**

With KNN, basically the closest points to the new point are searched. K represents the amount of the closest neighbors of the unknown point. We choose the quantity k (usually an odd number) of the algorithm to predict the results.

In K-NN classification, the output is class membership. An object is classified by the majority vote of its neighbors; the object is given to the class most common among its closest neighbors (k is a small positive integer). If k = 1, the object is simply assigned to that nearest neighbor's class.

I found the similarities of these vectors I created by putting them into the cosSim operation. Then I put the output values into the sort operation. Then, according to the sorted values, I returned the classification value that has the most in the first k value.

# **POST PROCESSING**

F1 score method was used for postprocessing. I ran my 10 test files. these 10 files have real results in my hand. Then I list the results of this KNN algorithm. I create a Confusion Matrix using Actual and predict results that I have.

Then I read the matrix and find the True Positive, True Negative, False Positive and False negative values.

Then I find the scores of Accuracy, Precision and Recall with these values.

Precision

Recall = 
$$\frac{\text{True Positive}}{\text{(True Positive + False Negative)}} = \frac{8}{8+4} = \frac{8}{12} = 0.67$$

Finally, using all these scores, I get my f1 score.

F1 - score = 2 \* 
$$\frac{\text{Precision * Recall}}{\text{Precision + Recall}} = 2 *  $\frac{0.72 * 0.67}{0.72 + 0.67} = 2 * \frac{0.48}{1.39} = 0.68$$$

f1 - score

Thus, I found the f1 scores of all my classes.

# **TEST**

The following files are my remaining files for testing

```
raw_texts/ekonomi/49.txt 54.txt
raw_texts/ekonomi/33.txt
raw_texts/ekonomi/106.t 177.txt
raw_texts/saglik/442.tx 203.txt
raw_texts/saglik/327.txt
raw_texts/saglik/362.tx 213.txt
raw_texts/saglik/319.tx 227.txt
raw_texts/saglik/322.txt
raw_texts/saglik/320.txt
raw_texts/siyasi/524.txt
raw_texts/siyasi/520.txt
raw_texts/siyasi/565.txt
raw_texts/siyasi/458.txt
raw_texts/siyasi/549.txt
raw texts/siyasi/587.tx 258.txt
raw texts/siyasi/478.txt
raw_texts/siyasi/552.txt
raw_texts/siyasi/476.txt
raw_texts/spor/609.tx 383.txt
raw_texts/spor/632.txt
raw_texts/spor/698.txt
raw_texts/spor/690.txt
raw_texts/spor/749.tx 343.txt
raw_texts/spor/627.tx 384.txt
raw texts/spor/726.txt
raw_texts/magazin/268.txt
raw_texts/magazin/163.tx 144.txt
raw_texts/magazin/168.txt
raw_texts/magazin/178.txt
raw_texts/magazin/185.tx 84.txt
raw_texts/magazin/264.txt
raw_texts/magazin/222.txt
Enter the file name for categorization:264.txt
Enter the method:KNN5
Assigned class is 1
```

input: 49.txt

49.txt is normally in economy class. The result after running with the KNN algorithm is as follows.

```
Enter the file name for categorization:49.txt Enter the method:KNN5
Assigned class is 1
Similarity to (Ekonomi)class 1 : 5
```

input: 327.txt

49.txt is normally in Healthy class. The result after running with the KNN algorithm is as follows.

```
Assigned class is 2
Similarity to (Sağlık)class 2 : 4
Similarity to (Magazin)class 5 : 1
```

Input: 476.txt

476.txt is normally in Politic class. The result after running with the KNN algorithm is as follows.

```
Assigned class is 3

Similarity to (Siyasi)class 3 : 4

Similarity to (Magazin)class 5 : 1
```

input: 698.txt

698.txt is normally in Sport class. The result after running with the KNN algorithm is as follows.

```
Assigned class is 4
Similarity to (Ekonomi)class 1 : 1
Similarity to (Spor)class 4 : 4
```

input 178.txt.

178 .txt is normally in Magazine class. The result after running with the KNN algorithm is as follows.

```
Assigned class is 5 min ()
Similarity to (Magazin)class 5 : 5
```

## **POST PROCESSING TEST**

```
Processing files...raw texts/ekonomi/49.txt
raw texts/saglik/320.txt
raw texts/siyasi/565.txt
raw texts/spor/632.txt
raw texts/magazin/178.txt
raw texts/ekonomi/33.txt
raw texts/saglik/322.txt
578.txt
raw texts/spor/749.txt
raw texts/magazin/185.txt
Real result : 1
              knnRsult : 1
              knnRsult : 2 [1][],i)
Real result : 2
Real result : 3
              knnRsult : 2
Real result : 4
              knnRsult : 4
Real result : 5
              knnRsult : 5
Real result : 1
              knnRsult : 1
Real result : 2
              knnRsult : 1
              knnRsult : 1
Real result : 3
Real result : 4 <sup>C</sup>
              knnRsult : 4
Real result : 5 knnRsult : 5
Conflusion Matrix----KnnDccurrence
20000
1 1 0 0 0
1 1 0 0 0
00020
0 0 0 0 2
Ekonomi ----> Recall : 1 Precision : 0.5 f1Score : 0.666667
Saglik ----> Recall : 0.5 Precision : 0.5 f1Score : 0.5
Siyasi ----> Recall : 0 Precision : 1 f1Score : 0
Spor -----> Recall : 1 Precision : 1 f1Score : 1
Magazain ---> Recall : 1 Precision : 1 f1Score : 1
Program closing...
```