

# Movie Review Classification Assignment

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**Problem Statement:** given a review, I need to develop a model to identify its polarity (either positive or negative)

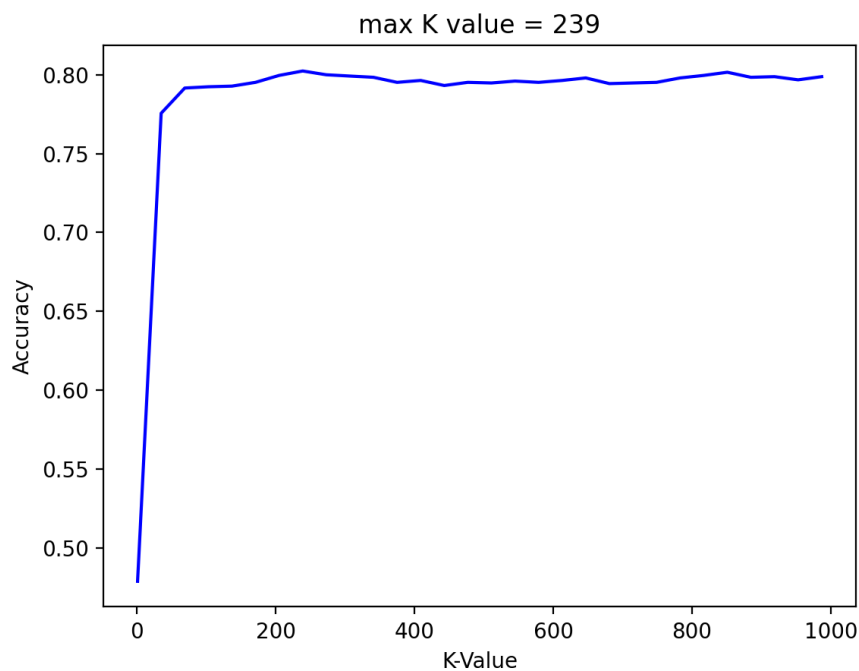
## **Solution:**

1. Preprocess the given reviews
2. convert the given reviews to  $x_q$ .
  - where  $x_q$  is the vector representation of reviews
3. Now for the  $x_q$ , determine its  $y_q$  using K-Nearest-Neighbors algorithm.
  - where  $y_q$  is the polarity
4. Validate the accuracy using cross validation technique with the help of train data

## **Explanation:**

- For preprocessing, I have followed the below steps and implemented the logic in the function named **preProcess()**
  - Removed HTML tags with the help of regular expressions (re library in python)
  - Removed punctuations or some special characters like ` , ~ , ! , @ , # , \$ , % , ^ , & , \* , ( , ) , \_ , - , + , = , { , } , | , \ , : , ; , " , " , < , , , > , . , ? , /
  - Checked for the words not being alphanumeric and have only the English letters
  - Ensured the length of each word is not greater than 2 as there are no adjectives in English with size of word being less than 2
  - Converted the words to lowercase letters
  - Removed stop words ()
  - Finally used the snowball stemmer(porter2) for stemming
- Now that we have the preprocessed data, I have converted the preprocessed data into vectors using **term frequency-inverse document frequency** vectorizer from *sklearn* with the parameters being
  - **min\_df**: it ignores the terms that have a document frequency strictly lower than the given threshold.
  - **max\_features**: only considers the top max\_features ordered by term frequency across the corpus
  - **max\_df**: I have used default value which is 1. It basically ignores terms that have a document frequency strictly higher than the given threshold
- using TFIDF instead of count vectorizer has increased the accuracy and the efficiency of the program to execute fast.
- by using the cosine similarity from *sklearn* library in python on the given test and train data I can get the indices of **k** nearest neighbors and based on the polarities of the neighbors I got, I am able to classify the review either positive or negative

- Finally, for validating the algorithm and to get the K value with high accuracy for KNN I have used K-fold cross validation technique with the help of **KFold** class from **sklearn** and split the given train data with k=10 as it was found to provide good trade-off of low computational cost and low bias in an estimate of model performance
- I have tried cross validating with K in the KNN ranging from 1 to 1001 and found that for K=239 I have got the highest accuracy. Also, I have noticed that this high accuracy is repeated at K around 601
- During this assignment I have learned how cleaning the data impacts the accuracy. I have increased my accuracy by removing punctuations and by using the snowball stemmer(porter2) instead of regular porter algorithm which is known to have better performance and accuracy
- Below is the graph I have plotted for better understanding of the changes in the accuracy with respect to K value



## References:

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