

Assignment No. 4

* Problem Statement :

Consider a suitable text dataset. Remove Stop words, apply Stemming & feature Selection techniques to represent documents as vector, classify documents & evaluate Precision, Recall.

* Objective : Learn how to tokenize & filter a document into its diff. words and then do word count for each word in doc. Apply stemming, feature Selection on doc (text)

* outcome : demonstrated text processing using nltk, understood vectorizing ; removing Stop words.

* Theory :

① NLP is a subfield of linguistic computer science & AI concerned with interaction b/w computer & human language, in particular how to program computer to process & analyse large amt of natural language data

② In computing stop words are words that are filtered out before or after the natural language data (text) are

Processed. While 'stop words' typically refer to the most common words, there is no universal list of stop words.

① **STEMMING** \Rightarrow for grammatical reasons, docs are going to use diff. forms of a word (e.g. organized, organizing, etc). Additionally there are families of derivatively related words with similar meaning like democracy & democratization.

\rightarrow The goal of stemming (and lemmatization) is to reduce inflectional forms & sometimes derivationally linked forms of a word to its common base form

am, are, is \rightarrow be

can, cans, cans', can's \rightarrow can

\rightarrow When applied to a doc, result will be somewhat like this

the boy's cars are diff. color
the boy can be differ color

\rightarrow Stemming is a more crude process, a heuristic that chops off the end of a word in the hope of achieving the goal correctly more of time, and often includes removal of derivational affixes

\rightarrow Feature selection is a process of selecting a subset of the terms occurring

in the remaining set and using only this subset as feature in text classification.

→ It serves 2 purposes. First, it makes training & applying a classifier more efficient by decreasing the size of vocab. this is of particular importance to classifier that are expensive to train.

→ Second, feature selection often increases classification accuracy by eliminating noise features. these features when added to docs representative increases the classification error on new data.

→ Vectorization is a process of converting text into machine readable form. words are represented as 'vectors' (numerically)

→ Count vectorizer (one-hot encoding) involves counting the number of occurrence for each word occurring in the doc.

→ Idea behind it is simple. vector is created having as many dimensions as there are distinct words in text/doc/collection of docs being used. each unique word has a unique dimension & is represented by a 1 in the dimension with 0s everywhere else. This results in huge & sparse vectors that capture no relational data.

→ If IDF vectors are related to one not encoded vectors, but instead of just feature a count, they feature numerical representation, where ~~they~~ words aren't just present or not present instead, they are represented by their term freq. multiplied by inverse doc. freq.

→ In simpler terms, words that occur everywhere should be given very little weight or significance, coz they don't provide a large amt. of value, however if a word appears very little or ~~freq~~ frequently but only in specific place, then they are possible of higher significance.

→ the downside is there is no capture of Semantic relatedness, this is solved with the co-occurrence matrix or a ~~stochastic~~ ~~probabilistic~~ model.
neural Probabilistic

* Conclusion :

Text docs were tokenized, filtered, vectorized, & thus Successfully Processed basics of NLP were understood.