

# **Evaluation of Machine Learning based Text Classifiers**

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## Problem Statement

Evaluating the different methods of Text Classification.

The question that are answered in this project are,

- How much time it took to classify given set of documents ?
- How accurately the documents got classified ?

## Proposed Solution !

The idea here is to classify different text datasets using multiple classifiers. SVM , GNB , LR Classifiers are used to classify the text document.

Experiment with different parameters of the classifier.

Better Feature Extraction Techniques. Like Stemming and Part of Speech tagging.



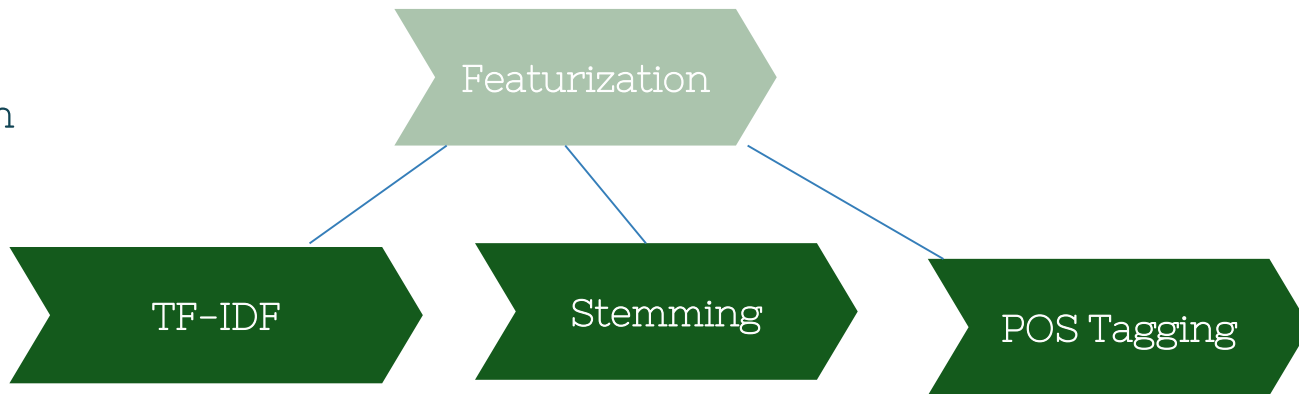


## Implementation

Traditional Approach



Improved Solution



## TF-IDF

Term Frequency – Inverse Document Frequency, Feature matrix is built by measuring, the number of times a term occurred in all documents divided by the number of documents the term occurred in.

This results in huge and sparse feature matrix. [samples X total number of words]

## Stemming

Each word is mapped back to its stem word. This reduces the number of features but doesn't lose any information.

Eg. Amazing and Amazed are mapped to Amaze.

## POS

Part of Speech tagging. Feature matrix contains maximum of words that doesn't affect the classification (Viz.) the, that, and, but, who etc.

The part of speech of every feature is tagged and words which are adjective, modal auxiliary, adverb, pre-determiner, noun, verbs are only selected.

This reduces the size of feature matrix while retaining the useful information.

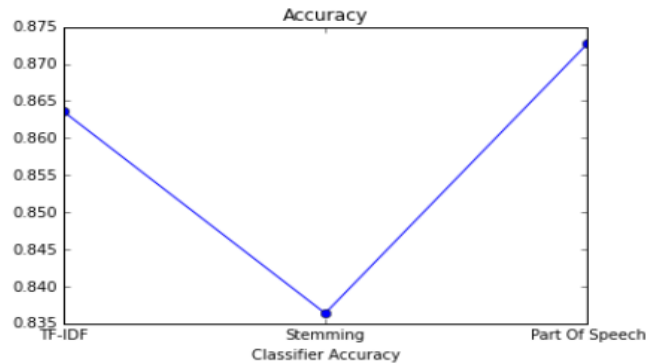
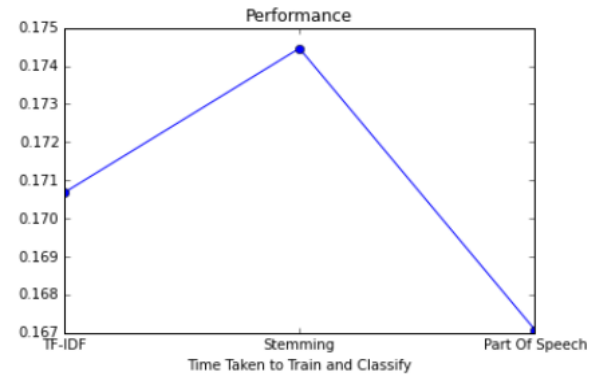
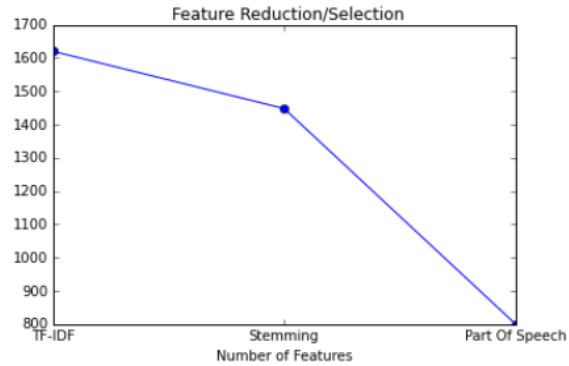


## And tables to compare data

Feature Size	Yelp Data	IMDB Data
TF-IDF	<b>1620</b>	<b>4524</b>
Stemming	<b>1448</b>	<b>3835</b>
POS	<b>801</b>	<b>2224</b>

Yelp Data (SVM)	Accuracy	Time Taken(in s)
TF-IDF	<b>0.863</b>	<b>0.174</b>
Stemming	<b>0.836</b>	<b>0.171</b>
POS	<b>0.872</b>	<b>0.167</b>

## And Graphs to show results



## Top Coefficients

### Positive Coefficients

u'great'

u'excellent'

u'superb'

u'worth'

u'liked'

u'loved'

u'enjoyed'

### Negative Coefficients

u'boring'

u'bad'

u'terrible'

u'poor'

u'awful'

u'waste'

u'almost



# Conclusion

From the experiments we can see the clear effect of feature selection, using POS tagging, on computation time as well as on accuracy of the classification.

We conclude that for classifying text based on the sentiment, SVM with POS tagging performed the best.



## Reference

1. Lillian Lee, Bo Pang, “Thumbs up? Sentiment Classification using Machine Learning Techniques”,

<http://www.cs.cornell.edu/home/llee/papers/sentiment.pdf>

2. Baharum Baharudin, Lam Hong Lee, and Khairullah Khan. A review of machine learning algorithms for text-documents classification. Journal of advances in information technology, 1(1):4–20, 2010

[https://www.researchgate.net/publication/43121576\\_A\\_Review\\_of\\_Machine\\_Learning\\_Algorithms\\_for\\_Text-Documents\\_Classification](https://www.researchgate.net/publication/43121576_A_Review_of_Machine_Learning_Algorithms_for_Text-Documents_Classification)

THANKS!

Any questions?