

Assignment 04

Title: Stemming and feature selection techniques using vectors

Problem Statement: Consider a suitable text. Remove stop words, apply stemming and feature selection techniques to represent documents as vectors, Classify documents and evaluate precision and recall.

Learning Objective: • Implementation of problem using python.
• Remove stop words, applying stemming and feature selection.

Learning Outcome: • Understanding the stemming and feature selection process.
• Learn about precision and recall.

Theory:

1) STOP WORDS.

- In Computing, stop words are words which are filtered out before or after processing of natural language data (text).

- Through "stop words" usually refers to most common values or words in a language, there is no universal list of stop words used by all natural processing tools, and indeed not all tools even use such a list.

- Any group of words can be chosen as the stop words for a given purpose. For some search

engine following are the most common, short function words:- the, is, at, which, and, on, etc.

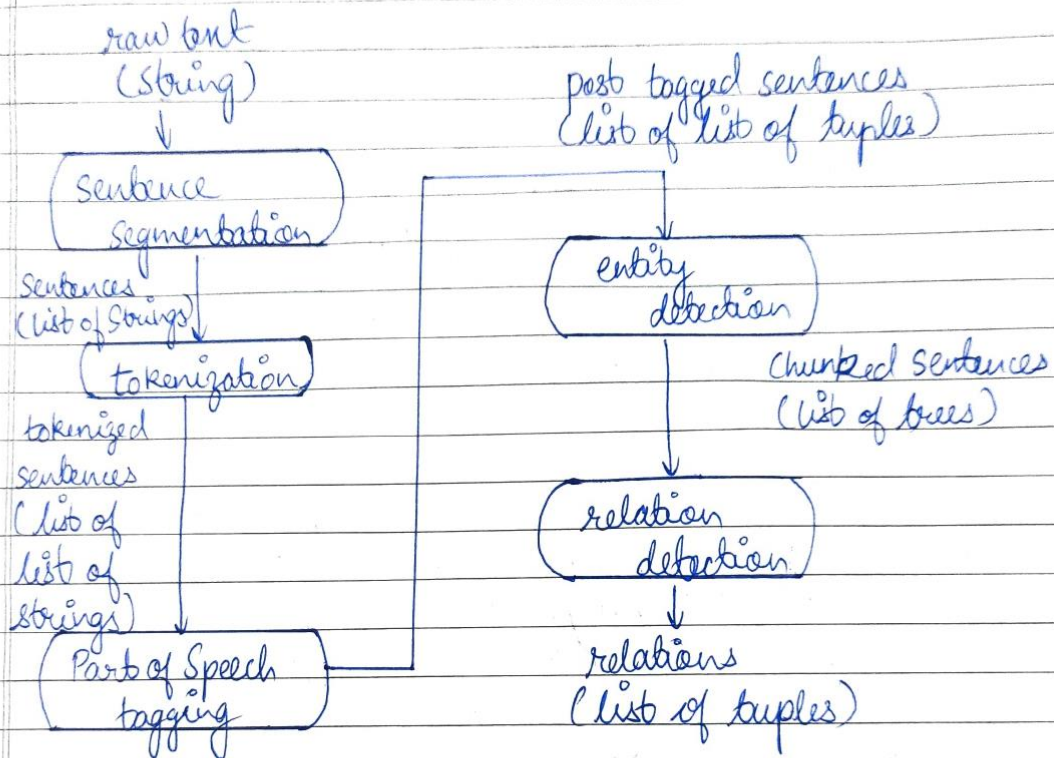
2] STEMMING :-

- Stemming is the process of reducing inflected words to their word stem, base or root form generally a written word form.
- The stem need not be identical to the morphological root of the word, it is usually sufficient that related words map to the same stem, even if the stem is not itself a valid root.
- Many search engines treat words with same stems as synonyms as a kind of query expansion, a process called conflation.
- Suffix-stripping algorithm is famous for stemming.

3] FEATURE SELECTION

- In machine learning and statistics, feature selection, also known as variable subset selection, attribute or variable subset selection, is a process of selecting a subset of relevant features (variables, predictions) for use in model construction.

• Feature extraction Architecture



→ Feature selection techniques are used for four reasons:

1. Simplification of models to make them easier to interpret by researchers/users.
2. Shorter training time.
3. To avoid the curse of dimensionality.
4. Enhanced generalization by reducing over fitting (formally, reduction of variance).

→ Stemming with nltk tool in python module.

```
from nltk.stem import Porter Stemmer
from nltk.tokenize import sent_tokenize, word_tokenize
```

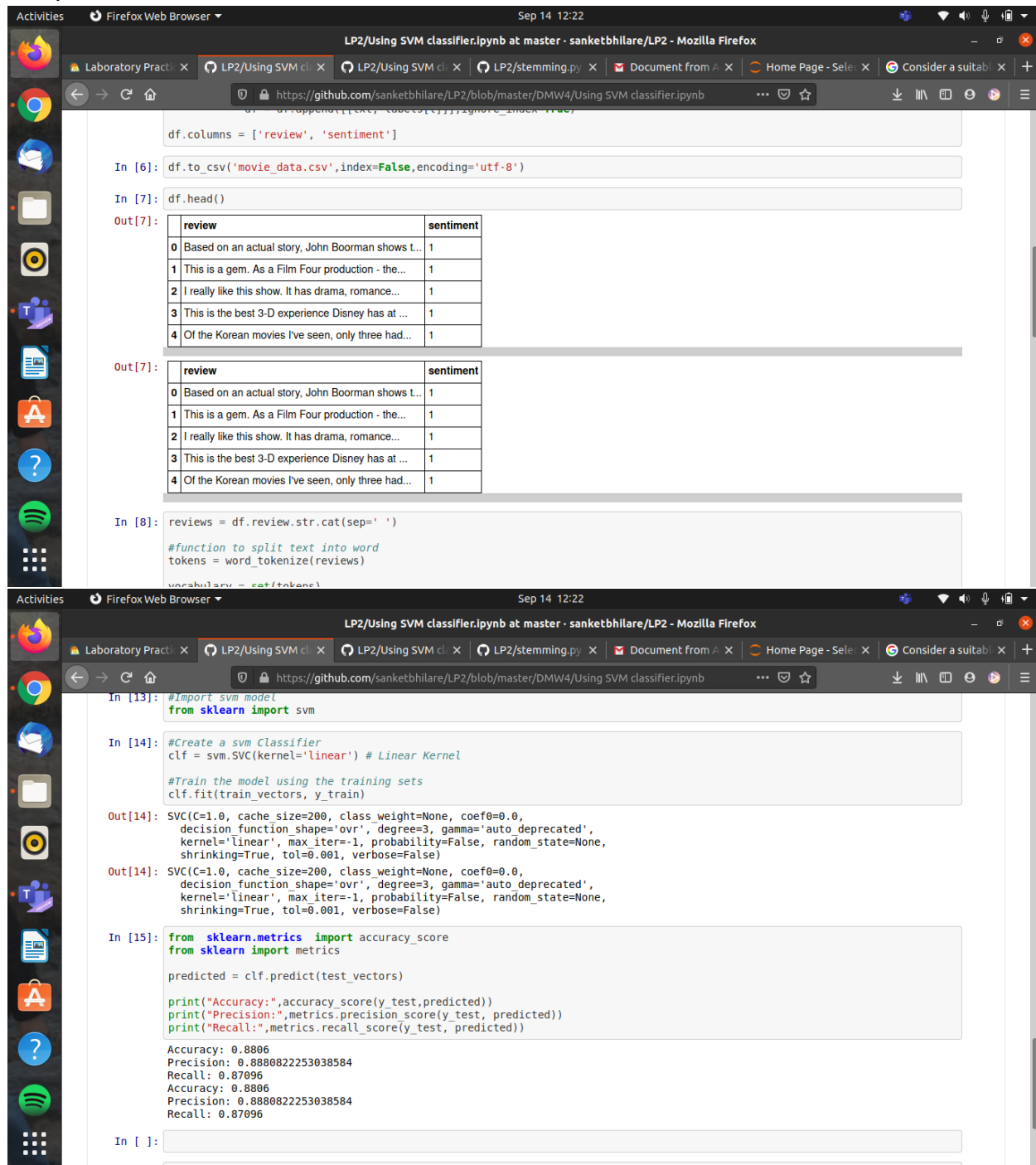
```
ps = PorterStemmer()
example-words = ("python", "pythoner", "Pythoning",
                  "Pythoned", "pythonly")
for w in example-words:
    print(ps.stem(w))
```

Test Cases:

| Sr.no | Description | Expected o/p | Actual o/p |
|-------|--|--------------|------------|
| 1) | Import pandas, os and nltk libraries into jupyter notebook | Success | Success |
| 2) | pre-process the data, and append pos/neg to review and sentiment columns. | Success | Success |
| 3) | Apply suffix-stemming stemming algorithm | Success | Success |
| 4) | Divide data into train and test and obtain accuracy, precision and recall of the model built | Success | Success |

Conclusion: Thus, we have studied to remove stop words, apply stemming and feature extraction techniques to represent documents as vectors.

Output:



The screenshot displays a Jupyter Notebook interface within a Mozilla Firefox browser window. The notebook is titled "LP2/Using SVM classifier.ipynb at master · sanketbhillare/LP2 - Mozilla Firefox". The interface shows a series of input and output cells. The first output cell displays a DataFrame with two columns: "review" and "sentiment". The DataFrame contains five rows of data, each with a review snippet and a sentiment score of 1. The second output cell shows the same DataFrame again. The third output cell displays the code for tokenizing the reviews. The fourth output cell shows the code for importing the SVM model and training it. The fifth output cell shows the code for predicting the sentiment of test vectors. The final output cell displays the accuracy, precision, and recall scores for the SVM classifier.

```
df.columns = ['review', 'sentiment']

In [6]: df.to_csv('movie_data.csv', index=False, encoding='utf-8')

In [7]: df.head()

Out[7]:
```

| | review | sentiment |
|---|---|-----------|
| 0 | Based on an actual story, John Boorman shows t... | 1 |
| 1 | This is a gem. As a Film Four production - the... | 1 |
| 2 | I really like this show. It has drama, romance... | 1 |
| 3 | This is the best 3-D experience Disney has at ... | 1 |
| 4 | Of the Korean movies I've seen, only three had... | 1 |

```
Out[7]:
```

| | review | sentiment |
|---|---|-----------|
| 0 | Based on an actual story, John Boorman shows t... | 1 |
| 1 | This is a gem. As a Film Four production - the... | 1 |
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| 3 | This is the best 3-D experience Disney has at ... | 1 |
| 4 | Of the Korean movies I've seen, only three had... | 1 |

```
In [8]: reviews = df.review.str.cat(sep=' ')

#function to split text into word
tokens = word_tokenize(reviews)

vocabulary = cat(tokens)

In [13]: #Import svm model
from sklearn import svm

In [14]: #Create a svm Classifier
clf = svm.SVC(kernel='linear') # Linear Kernel

#Train the model using the training sets
clf.fit(train_vectors, y_train)

Out[14]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
decision_function_shape='ovr', degree=3, gamma='auto deprecated',
kernel='linear', max_iter=1, probability=False, random_state=None,
shrinking=True, tol=0.001, verbose=False)

Out[14]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
decision_function_shape='ovr', degree=3, gamma='auto deprecated',
kernel='linear', max_iter=1, probability=False, random_state=None,
shrinking=True, tol=0.001, verbose=False)

In [15]: from sklearn.metrics import accuracy_score
from sklearn import metrics

predicted = clf.predict(test_vectors)

print("Accuracy:", accuracy_score(y_test, predicted))
print("Precision:", metrics.precision_score(y_test, predicted))
print("Recall:", metrics.recall_score(y_test, predicted))

Accuracy: 0.8806
Precision: 0.8880822253038584
Recall: 0.87096
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Precision: 0.8880822253038584
Recall: 0.87096

In [ ]:
```