

# CSE6060

## Statistical Natural Language Processing

### Sentiment Analysis

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### Loading Dataset

In [2]:

```
1 import pandas as pd
2 import numpy as np
3 from sklearn.feature_extraction.text import CountVectorizer, TfidfTransformer, TfidfVec
```

In [5]:

```
1 df = pd.read_csv("IMDB Dataset.csv")
```

In [6]:

```
1 # After Loading, Lets introspect this dataset
2 df.head(10)
```

Out[6]:

	review	sentiment
0	One of the other reviewers has mentioned that ...	positive
1	A wonderful little production.   The...	positive
2	I thought this was a wonderful way to spend ti...	positive
3	Basically there's a family where a little boy ...	negative
4	Petter Mattei's "Love in the Time of Money" is...	positive
5	Probably my all-time favorite movie, a story o...	positive
6	I sure would like to see a resurrection of a u...	positive
7	This show was an amazing, fresh & innovative i...	negative
8	Encouraged by the positive comments about this...	negative
9	If you like original gut wrenching laughter yo...	positive

In [7]:

```
1 df['review'][0]
```

Out[7]:

"One of the other reviewers has mentioned that after watching just 1 Oz episode you'll be hooked. They are right, as this is exactly what happened with me.<br /><br />The first thing that struck me about Oz was its brutality and unflinching scenes of violence, which set in right from the word G O. Trust me, this is not a show for the faint hearted or timid. This show pulls no punches with regards to drugs, sex or violence. Its is hardcore, in the classic use of the word.<br /><br />It is called OZ as that is the nickname given to the Oswald Maximum Security State Penitentiary. It focuses mainly on Emerald City, an experimental section of the prison where all the cells have glass fronts and face inwards, so privacy is not high on the agenda. Em City is home to many..Aryans, Muslims, gangstas, Latinos, Christians, Italians, Irish and more....so scuffles, death stares, dodgy dealings and shady agreements are never far away.<br /><br />I would say the main appeal of the show is due to the fact that it goes where other shows wouldn't dare. Forget pretty pictures painted for mainstream audiences, forget charm, forget romance...OZ doesn't mess around. The first episode I ever saw struck me as so nasty it was surreal, I couldn't say I was ready for it. but as I watched more. I developed a taste for Oz. and got accustomed

## Transforming Documents into Feature Vectors

In [8]:

```
1 count = CountVectorizer() # initialize Count Vectorizer
2 docs = np.array(['The sun is shining',
3                 'The weather is sweet',
4                 'The sun is shining, the weather is sweet, and one and one is two'])
5 bag = count.fit_transform(docs)
6 print(count.vocabulary_)
7 print("\n")
8 print(bag.toarray())
```

```
{'the': 6, 'sun': 4, 'is': 1, 'shining': 3, 'weather': 8, 'sweet': 5, 'and': 0, 'one': 2, 'two': 7}
```

```
[[0 1 0 1 1 0 1 0 0]
 [0 1 0 0 0 1 1 0 1]
 [2 3 2 1 1 1 2 1 1]]
```

## Term Frequency and Inverse Document

In [9]:

```
1 tfidf = TfidfTransformer(use_idf = True, norm='l2', smooth_idf=True)
2 np.set_printoptions(precision=2)
3 print(tfidf.fit_transform(count.fit_transform(docs)).toarray())
```

```
[[0.   0.43 0.   0.56 0.56 0.   0.43 0.   0.  ]
 [0.   0.43 0.   0.   0.   0.56 0.43 0.   0.56]
 [0.5  0.45 0.5  0.19 0.19 0.19 0.3  0.25 0.19]]
```

## 1 # Tokenization of Documents

In [10]:

```
1 from nltk.stem.porter import PorterStemmer
2 porter = PorterStemmer()
3
4 def stemmer_tokenize(text):
5     return [porter.stem(word) for word in text.split()]
6
7 stemmer_tokenize('coders like coding and thus they code')
```

Out[10]:

```
['coder', 'like', 'code', 'and', 'thu', 'they', 'code']
```

## Vectorising the Dataset

In [11]:

```
1 tfidf = TfidfVectorizer(strip_accents = None,
2                           lowercase = False,
3                           tokenizer = stemmer_tokenize,
4                           use_idf = True,
5                           norm = 'l2',
6                           smooth_idf = True)
7
8 Y = df.sentiment.values
9 X = tfidf.fit_transform(df.review)
```

## Document Classification Using Logistic regression

In [12]:

```
1 from sklearn.model_selection import train_test_split
2 from sklearn.linear_model import LogisticRegressionCV
3 import pickle
4
5 X_train, X_test, Y_train, Y_test = train_test_split(X, Y, random_state = 1, test_size = 0.2)
6
7 # Model
8 clf = LogisticRegressionCV(cv = 5,
9                             scoring = 'accuracy',
10                             random_state=0,
11                             n_jobs=-3,
12                             verbose=3,
13                             max_iter=300).fit(X_train, Y_train)
14
15 # saving the model
16 saved_model = open('saved_model.sav', 'wb')
17
18 # using the pickle library's dump function to write the trained classifier to the file
19 pickle.dump(clf, saved_model)
20 saved_model.close()
21
```

[Parallel(n\_jobs=-3)]: Using backend LokyBackend with 10 concurrent workers.

[Parallel(n\_jobs=-3)]: Done 2 out of 5 | elapsed: 2.8min remaining: 4.2min

[Parallel(n\_jobs=-3)]: Done 5 out of 5 | elapsed: 3.0min finished

## Model Evaluation

In [13]:

```
1 filename = 'saved_model.sav'
2 saved_clf = pickle.load(open(filename, 'rb'))
3
4 # test the saved model on the test data
5 saved_clf.score(X_test, Y_test)
6 #saved_clf.close()
```

Out[13]:

0.8898

In [ ]:

1