

# TP2 - Sentiment Analysis

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## 1 Sentiment Analysis In Textual Movie Reviews

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```
In [150]: import string

In [151]: import os
          os.chdir('C:\\Users\\Max Tchibozo\\Desktop\\SD-TSIA214\\TP2\\data\\data')

In [152]: # Authors: Alexandre Gramfort
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          # License: BSD Style.
          # TP Cours ML Telecom ParisTech MDI343

          import os.path as op
          import numpy as np

          from sklearn.base import BaseEstimator, ClassifierMixin

          #####
          # Load data
          print("Loading dataset")

          from glob import glob
          filenames_neg = sorted(glob(op.join '..', 'data', 'imdb1', 'neg', '*.txt')))
          filenames_pos = sorted(glob(op.join '..', 'data', 'imdb1', 'pos', '*.txt')))
          texts_neg = [open(f).read() for f in filenames_neg]
          texts_pos = [open(f).read() for f in filenames_pos]
          texts = texts_neg + texts_pos
          y = np.ones(len(texts), dtype=np.int)
          y[:len(texts_neg)] = 0.

          print("%d documents" % len(texts))

Loading dataset
2000 documents
```

```
In [153]: y[999],y[1000]#This is the moment when we go from the negative reviews to the positive
```

```
Out[153]: (0, 1)
```

## 2 Question 1

```
In [154]: import string
          #####
          # Start part to fill in

def count_words(texts):
    """Vectorize text : return count of each word in the text snippets

    Parameters
    -----
    texts : list of str
        The texts

    Returns
    -----
    vocabulary : dict
        A dictionary that points to an index in counts for each word.
    counts : ndarray, shape (n_samples, n_features)
        The counts of each word in each text.
    n_samples == number of documents.
    n_features == number of words in vocabulary.
    """

    punctuation = list(string.punctuation)+['\n'] #We also want to remove the newlines

    total_word_list = [] #will contain the words of all texts without separation
    text_word_list = [] #will separate the words of each text in a different list it

    for text in texts:
        for punct in punctuation:
            text = text.replace(punct, ' ')
        word_list = text.split(' ')
        word_list = [x for x in word_list if x != ''] #We remove the empty strings :

        total_word_list += word_list # We build vocabulary thanks to the list of all
        text_word_list.append(word_list)

    words = list(set(total_word_list))
    vocabulary = {}

    for i in range(len(words)):
        vocabulary[words[i]] = i
```

```

counts = np.zeros((2000,len(words))) #there are 2000 documents

for i in range(len(text_word_list)):
    for j in range(len(text_word_list[i])):
        index = vocabulary[text_word_list[i][j]] #This is the index of the word
        counts[i][index] += 1

return vocabulary, counts

count_words(texts)

```

```

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```

'humourous': 995,
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       [0., 0., 0., ..., 0., 0., 0.],
       ...,
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.]])

```

```
In [155]: vocabulary , counts = count_words(texts)
```

There are 39443 different words in the vocabulary of the IMDB Movie Review database.

It is important to realise that some of these words do not hold much semantic meaning because of the pre-processing we have done:

Composite words (i.e "Jean-Claude") and more generally words with any type of punctuation (i.e: O-M-G) are split into several individual sub-words ("Jean","Claude" and "O","M","G").

### 3 Question 2

The class is attributed to reviews is recognized through the first explicit and identifiable rating.

These ratings are specified through star and numerical values i.e : "8/10", "four out of five", and "OUT OF \*\*\*\*. \*\*\*"

There is one noteworthy specificity, which is that in the case where the identified rating is a 5 star rating with decimal points (i.e 2.5 stars, 3.5 stars), the associated rating will be the truncated value of the rating (resp. 2 stars, 3 stars).

The authors explain that this is not problematic, given that the class output of  $\pm 1$  will be erroneous only when the rating was 2.5. And even then, it could be said that average reviews are negative reviews.

### 4 Question 3

```

In [157]: class NB(BaseEstimator, ClassifierMixin):
           def __init__(self):
               pass

           def fit(self, X, y): #This corresponds to TrainMultinomialNB
               #X = counts
               #y = vecteur de {0,1} label de la classe
               N = X.shape[0]
               self.prior = np.zeros(2)
               counts_neg = X[:N//2][:]
               counts_pos = X[N//2:][:]

```



```

counts = [counts_neg, counts_pos]
self.condprob = np.zeros((X.shape[1], 2))
for c in range(2): #for each class (0 or 1)
    Nc = len(y)/2 #half of the docs are positive, the other half are negative
    self.prior[c] = Nc/N
    current_counts = counts[c] #This is the counts matrix of the given class
    c_total_counts = 0

    T = np.zeros((2, X.shape[1]))
    sums = np.zeros(2)
    for t in range(current_counts.shape[1]):
        T[c][t] = np.sum(list(current_counts[:, t]))
        sums[c] += T[c][t]

    for t in range(current_counts.shape[1]):
        self.condprob[t][c] = (T[c][t] + 1) / (sums[c] + current_counts.shape[1])
return self

def predict(self, X):

    n = X.shape[0]
    predictions = np.zeros(n, dtype=int)

    for i in range(X.shape[0]):
        W = np.argwhere(X[i] != 0).flatten() #Where X points are != 0

        score = np.zeros(2)
        for c in range(2):
            score[c] = np.log(self.prior[c])
            for t in W:
                score[c] += np.log(self.condprob[t][c])

        predictions[i] = np.argmax(score)

    return predictions

def score(self, X, y):
    return np.mean(self.predict(X) == y)

# Count words in text
vocabulary, X = count_words(texts)

# Try to fit, predict and score
nb = NB()
nb.fit(X[:, :2], y[:, :2])
print('The score on the complete X dataset is : ' + str(nb.score(X[1::2], y[1::2])))

```

The score on the complete X dataset is : 0.82

## 5 Cross-Validation 5-Folds

```
In [158]: from sklearn.model_selection import cross_val_score
```

```
print('The 5-fold cross-validation score is : '+str(cross_val_score(NB(), X, y, cv=5))
```

The 5-fold cross-validation score is : 0.82550000000000001

## 6 Stop-Words

```
In [159]: with open ('english.stop','r') as f:
          lines = f.readlines()
          stop_words = [x[:-1] for x in lines]
          stop_words
```

```
Out[159]: ['a',
           "a's",
           'able',
           'about',
           'above',
           'according',
           'accordingly',
           'across',
           'actually',
           'after',
           'afterwards',
           'again',
           'against',
           "ain't",
           'all',
           'allow',
           'allows',
           'almost',
           'alone',
           'along',
           'already',
           'also',
           'although',
           'always',
           'am',
           'among',
           'amongst',
           'an',
           'and',
```

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'without',  
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```
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"you'll",
"you're",
"you've",
'your',
'yours',
'yourself',
'yourselves',
'z',
'zero']
```

```
In [160]: def count_words(texts):
           """Vectorize text : return count of each word in the text snippets

           Parameters
           -----
           texts : list of str
               The texts

           Returns
           -----
           vocabulary : dict
               A dictionary that points to an index in counts for each word.
           counts : ndarray, shape (n_samples, n_features)
               The counts of each word in each text.
           n_samples == number of documents.
           n_features == number of words in vocabulary.
           """

           punctuation = list(string.punctuation)+['\n']+stop_words #We now remove the stop

           total_word_list = [] #will contain the words of all texts without separation
           text_word_list = [] #will separate the words of each text in a different list it

           for text in texts:
               for punct in punctuation:
                   text = text.replace(punct, ' ')

               word_list = text.split(' ')
               word_list = [x for x in word_list if x != ''] #We remove the empty strings :

               total_word_list += word_list # We build vocabulary thanks to the list of all
               text_word_list.append(word_list)
```

```

words = list(set(total_word_list))
vocabulary = {}

for i in range(len(words)):
    vocabulary[words[i]] = i

counts = np.zeros((2000, len(words))) #there are 2000 documents

for i in range(len(text_word_list)):
    for j in range(len(text_word_list[i])):
        index = vocabulary[text_word_list[i][j]] #This is the index of the word
        counts[i][index] += 1

return vocabulary, counts

count_words(texts)

```

```

Out[160]: ({'1935': 0,
            '1912': 1,
            '300': 2,
            '50000': 3,
            '1865': 4,
            '1980': 5,
            '44': 6,
            '209': 7,
            '122': 8,
            '1956': 9,
            '65': 10,
            '640': 11,
            '118': 12,
            '1986': 13,
            '\x13': 14,
            '1952': 15,
            '1400': 16,
            '230': 17,
            '20': 18,
            '\x12': 19,
            '1938': 20,
            '700': 21,
            '1942': 22,
            '75': 23,
            '1925': 24,
            '\x05\x05': 25,
            '67': 26,
            '1972': 27,
            '1871': 28,

```

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```

```
In [161]: from sklearn.model_selection import cross_val_score
          vocabulary, X = count_words(texts)
          nb = NB()

          print('The 5-fold cross-validation score WITHOUT the stop-words is : '+str(cross_val_
```

The 5-fold cross-validation score WITHOUT the stop-words is : 0.5405

Removing the stop-words worsens the performance.

## 7 Scikit-Learn Use

```
In [162]: from sklearn.naive_bayes import MultinomialNB
          from sklearn.feature_extraction.text import CountVectorizer
          from sklearn.pipeline import Pipeline

In [163]: vectorizer = CountVectorizer()
          X = vectorizer.fit_transform(texts)
          nb = MultinomialNB()
          print('The 5-fold cross-validation score of the sklearn function WITHOUT the stop-words')
```

The 5-fold cross-validation score of the sklearn function WITHOUT the stop-words is : 0.8145

The default scikitlearn libraries yield a much better score than our hand-made estimator.

## 8 LinearSVC

```
In [164]: from sklearn.svm import LinearSVC
          clf = LinearSVC()
          clf.fit(X,y)
          print('The 5-fold cross-validation score of the Linear SVC WITHOUT the stop-words is
```

The 5-fold cross-validation score of the Linear SVC WITHOUT the stop-words is : 0.832500000000

LinearSVC yields a similar score to the Sklearn Bayesian approach. However, this score is slightly higher than our hand-made Bayesian estimator.

## 9 NLTK Stemming

```
In [165]: from nltk import SnowballStemmer

In [166]: stemmer = SnowballStemmer(language="english")
           stemmer.stem('congratulations')
```

```
Out[166]: 'congratul'
```

```
In [167]: def count_words(texts):
          """Vectorize text : return count of each word in the text snippets

          Parameters
          -----
          texts : list of str
                  The texts

          Returns
          -----
          vocabulary : dict
                      A dictionary that points to an index in counts for each word.
          counts : ndarray, shape (n_samples, n_features)
                  The counts of each word in each text.
                  n_samples == number of documents.
                  n_features == number of words in vocabulary.
          """

          punctuation = list(string.punctuation)+['\n'] #We also want to remove the newlin

          total_word_list = [] #will contain the words of all texts without separation
          text_word_list = [] #will separate the words of each text in a different list it

          for text in texts:
              for punct in punctuation:
                  text = text.replace(punct, ' ')
              word_list = text.split(' ')
              word_list = [stemmer.stem(x) for x in word_list if x != ''] #We remove the e

              total_word_list += word_list # We build vocabulary thanks to the list of all
              text_word_list.append(word_list)

          words = list(set(total_word_list))
          vocabulary = {}

          for i in range(len(words)):
              vocabulary[words[i]] = i

          counts = np.zeros((2000, len(words))) #there are 2000 documents

          for i in range(len(text_word_list)):
              for j in range(len(text_word_list[i])):
                  index = vocabulary[text_word_list[i][j]] #This is the index of the word
                  counts[i][index] += 1

          return vocabulary, counts
```



```
count_words(texts)
```

```
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```

```
...,
[0., 0., 0., ..., 0., 0., 0.],
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[0., 0., 0., ..., 0., 0., 0.]])])
```

```
In [168]: vocabulary, X = count_words(texts)
```

```
# Try to fit, predict and score
nb = NB()
print('The 5-fold cross-validation score of our hand-made estimator after stemming is
```

```
The 5-fold cross-validation score of our hand-made estimator after stemming is : 0.821000000000
```

Conclusion :

We observe that both the stemmed and non-stemming lead to comparable results : ~0.82 on a 5-fold cross-validation. This is extremely interesting, as it means that we analyze a much smaller text dataset.

The stemmed text dataset contains only partial words, and all words which have the same root will become identical. Our set of words is much smaller as it only contains those very roots.

Stemming amounts to compressing the information, meaning we can process large amounts of text in a shorter amount of time, with comparable results.

## 10 Part of Speech

```
In [169]: import nltk
          nltk.download('punkt')
          nltk.download('averaged_perceptron_tagger')
          nltk.download('universal_tagset')
```

```
[nltk_data] Downloading package punkt to C:\Users\Max
[nltk_data]       Tchibo\\AppData\Roaming\nltk_data...
[nltk_data]   Package punkt is already up-to-date!
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]       C:\Users\Max Tchibo\\AppData\Roaming\nltk_data...
[nltk_data]   Package averaged_perceptron_tagger is already up-to-
[nltk_data]       date!
[nltk_data] Downloading package universal_tagset to C:\Users\Max
[nltk_data]       Tchibo\\AppData\Roaming\nltk_data...
[nltk_data]   Package universal_tagset is already up-to-date!
```

```
Out[169]: True
```

```
In [170]: from nltk import pos_tag, word_tokenize
```

```
In [171]: pos_tag(word_tokenize("John's big idea isn't all that bad very strong. To eating, to
```

```

Out[171]: [('John', 'NOUN'),
            (''s', 'PRT'),
            ('big', 'ADJ'),
            ('idea', 'NOUN'),
            ('is', 'VERB'),
            ('n't', 'ADV'),
            ('all', 'DET'),
            ('that', 'ADP'),
            ('bad', 'ADJ'),
            ('very', 'ADV'),
            ('strong', 'ADJ'),
            ('.', '.'),
            ('To', 'PRT'),
            ('eating', 'VERB'),
            (',', '.'),
            ('to', 'PRT'),
            ('be', 'VERB'),
            ('or', 'CONJ'),
            ('not', 'ADV'),
            ('to', 'PRT'),
            ('be', 'VERB')]

```

```

In [172]: accepted_words = ['NOUN', 'VERB', 'ADV', 'ADJ']

```

```

In [173]: def count_words(texts):
            """Vectorize text : return count of each word in the text snippets

            Parameters
            -----
            texts : list of str
                    The texts

            Returns
            -----
            vocabulary : dict
                    A dictionary that points to an index in counts for each word.
            counts : ndarray, shape (n_samples, n_features)
                    The counts of each word in each text.
            n_samples == number of documents.
            n_features == number of words in vocabulary.
            """

            total_word_list = [] #will contain the words of all texts without separation
            text_word_list = [] #will separate the words of each text in a different list it

            for text in texts:
                sentence = pos_tag(word_tokenize(text), tagset='universal')
                word_list = [x[0] for x in sentence if x[1] in accepted_words] #We retain on
                total_word_list += word_list # We build vocabulary thanks to the list of all

```

```

        text_word_list.append(word_list)

words = list(set(total_word_list))
vocabulary = {}

for i in range(len(words)):
    vocabulary[words[i]] = i

counts = np.zeros((2000,len(words))) #there are 2000 documents

for i in range(len(text_word_list)):
    for j in range(len(text_word_list[i])):
        index = vocabulary[text_word_list[i][j]] #This is the index of the word
        counts[i][index] += 1

return vocabulary, counts

count_words(texts)

```

```

Out[173]: ({'o'barr': 0,
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'correspondence': 994,
'schoolteacher': 995,
'necessarily': 996,
'glossing': 997,
'movie-you': 998,
'uncharismatic': 999,
...},
array([[0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
       ...,
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.],
       [0., 0., 0., ..., 0., 0., 0.]])

```

```
In [174]: vocabulary, X = count_words(texts)
```

```

# Try to fit, predict and score
nb = NB()
print('The 5-fold cross-validation score of our hand-made estimator after stemming is

```

The 5-fold cross-validation score of our hand-made estimator after stemming is : 0.8355

Using the nltk library and keeping only nouns, verbs, adverbs and adjectives yields the best results!

One might still prefer the stemming approach for computational and compression reasons.