Distant Supervision - The End

January 23, 2015

1 Import

```
In [830]: import os, csv, tweepy, ucto, time
          import numpy as np
          import pandas as pd
          from sklearn.ensemble import ExtraTreesClassifier
          from sklearn.feature_extraction.text import TfidfTransformer
          from sklearn.feature_extraction.text import CountVectorizer
          from sklearn.linear_model import SGDClassifier, LogisticRegression
          from sklearn.cross_decomposition import PLSRegression
          from sklearn import tree
          from sklearn.svm import LinearSVC, NuSVC
          from sklearn.ensemble import BaggingClassifier
          from sklearn import metrics
          from sklearn.grid_search import GridSearchCV
          from sklearn.pipeline import FeatureUnion
          from pprint import pprint
          from sklearn.base import BaseEstimator
          import pickle
          from scipy import sparse
          from sklearn.feature_selection import SelectKBest, chi2
          from sklearn.linear_model import RidgeClassifier
          from sklearn.linear_model import SGDClassifier
          from sklearn.linear_model import Perceptron
          from sklearn.linear_model import PassiveAggressiveClassifier
          from sklearn.naive_bayes import MultinomialNB
          from sklearn.neighbors import NearestCentroid
          import logging
          from tweepy.streaming import StreamListener
          from tweepy import Stream
          from pynlpl.clients.frogclient import FrogClient
```

1.1 Settings

```
In [831]: os.chdir("/home/pold/PycharmProjects/twitter-tutorial/")
```

1.2 General Classifier

1.3 Read Aggressive Tweets from TSV files

```
In [878]: def create_key(x):
              '''', Create a file name from a number'''
              return "tweets/fold_" + str(x) + ".txt"
          # Create a list of file names of the aggressive tweets.
          files = [create_key(x) for x in xrange(0,10)]
          # Store everything in a pandas data frame.
          list = []
          for file in files:
              df = pd.read_table(file,
                                 names = ('category', 'user', 'date', 'time', 'message'),
                                 header = None,
                                 index_col = None)
              list.append(df)
          frame = pd.concat(list, axis=0, keys = files)
1.4 Use Dutch tokeniser rules
In [879]: settingsfile = "/etc/ucto/tokconfig-nl-twitter"
          tokenizer = ucto.Tokenizer(settingsfile)
          def remove_non_ascii(s):
              return "".join(i for i in s if ord(i) < 128)</pre>
          def ucto_tokenizer(str):
              remove_non_ascii(str)
              tokenizer.process(unicode(str))
              tokens = [unicode(token).encode('utf-8') for token in tokenizer]
              for pos, token in enumerate(tokens):
                  if token.startswith('http:'): tokens[pos] = 'url'
                  if token.startswith('0'): tokens[pos] = 'user'
              return tokens
          ucto_tokenizer("Was geht #leute #heute @POld hey? http://tinyurl.com/dasd")
Out[879]: ['Was', 'geht', '#', 'leute', '#', 'heute', 'user', 'hey', '?', 'url']
     Aggressive Hashtags - Wordlist
1.5
In [880]: # The wordlist is a dictionary of profanity words.
          wordlists = {}
          # for name in ["illness", "religion", "sexuality", "slurs", "misc"]:
          for name in ["selection"]:
              with open("wordlist_" + name + ".txt") as f:
                  words = f.read().splitlines()
                  wordlists[name] = words
```

2 Twitter

2.1 Authentication

```
In [881]: consumer_key = 'zdprSqld7JxkSjaeqQppemW5y'
          consumer_secret = 'QMZCPjMZnqJjkXQxXerHGCP11PGXIr17iXHOHnrlLUjwNVOQdY'
          access_token = '1612711080-64a9w2e4B0TV4id4HTqLpza08Z6ag0pNVE9axZv'
          access_token_secret = 'iGBJ7jTo7LjqdMWyU8sJT5xX6CllWX0YoeQScYxpbGjMr'
2.2 Connect to Twitter
In [882]: auth = tweepy.OAuthHandler(consumer_key, consumer_secret)
          auth.set_access_token(access_token, access_token_secret)
          api = tweepy.API(auth)
2.3 Create a Query
In [883]: def create_query(wordlist):
             return " OR ".join(wordlist)
          def create_negate_query(wordlist):
              return " ".join(["-" + word for word in wordlist])
2.4 Collect Tweets
In [838]: results_aggressive = []
         results_non_aggressive = []
         flattened_wordlist = [item for sublist in wordlists.values() for item in sublist]
          query_pos = create_query(flattened_wordlist)
          query_neg = create_negate_query(flattened_wordlist)
          # Aggressive Tweets
         for tweet in tweepy.Cursor(api.search, q = query_pos, lang = "nl").items(1500):
             results_aggressive.append(tweet)
          time.sleep(60)
          # Non-aggressive Tweets
         for tweet in tweepy.Cursor(api.search, q = query_neg, lang = "nl").items(len(results_aggress
             results_non_aggressive.append(tweet)
     Count Hashtags
In [885]: flattened_wordlist = [item for sublist in wordlists.values() for item in sublist]
         hash_dict = { }
         for word in flattened_wordlist:
             hash_dict[word] = 0
```

for tweet in results_aggressive:

for word in flattened_wordlist:

```
if word in tweet.text: hash_dict[word] += 1
          print(len(results_aggressive))
          print(len(results_non_aggressive))
          print(hash_dict)
593
593
{'#idiot': 12, '#fuckyou': 19, '#gvd': 15, '#sukkel': 42, '#grrr': 60, '#idioot': 48, '#grr': 75}
2.6
     Additional Feature Extractor
In [842]: # The wordlist is a dictionary of profanity words.
          badwords = []
          for name in ["illness", "religion", "sexuality", "slurs", "misc"]:
              with open("wordlist_" + name + ".txt") as f:
                  word = f.read().splitlines()
                  badwords.append(word)
          class MyWordCounter(BaseEstimator):
              def __init__(self):
                  with open("my_badlist.txt") as f:
                      badwords = [1.strip() for 1 in f.readlines()]
                  self.badwords_ = badwords
              def fit(self, documents, y=None):
                  return self
              def transform(self, documents):
                  n_words = [len(ucto_tokenizer(remove_non_ascii(tweet))) for tweet in documents]
                  n_chars = [len(tweet) for tweet in documents]
                  # number of uppercase words
                  allcaps = [np.sum([w.isupper() for w in comment.split()])
                         for comment in documents]
                  # longest word
                  max_word_len = [np.max([len(w) for w in c.split()]) for c in documents]
                  # average word length
                  mean_word_len = [np.mean([len(w) for w in c.split()])
                                                      for c in documents]
                  # number of badwords:
                  n_bad = [np.sum([c.lower().count(w) for w in self.badwords_])
                                                          for c in documents]
                  exclamation = [c.count("!") for c in documents]
                  addressing = [c.count("0") for c in documents]
                  spaces = [c.count(" ") for c in documents]
                  allcaps_ratio = np.array(allcaps) / np.array(n_words, dtype=np.float)
                  bad_ratio = np.array(n_bad) / np.array(n_words, dtype=np.float)
                  return np.array([n_words
                                   , n_chars
```

```
, allcaps
, max_word_len
, mean_word_len
, exclamation
, addressing
, spaces
, bad_ratio
, n_bad
, allcaps_ratio
```

]).T

2.7 Define Pipeline

2.8 Leave-one-out cross-validation

```
In [914]: accuracy_per_fold = []

for x in xrange(0, 10):
    train = frame.ix[[create_key(y) for y in range(0,10) if x != y]]

# test = frame.ix[create_key(x)]

messages_train = train['message']
    categories_train = train['category']

messages_test = test['message']
    categories_test = test['category']

# train classifier
    pipeline = pipeline.fit(messages_train, categories_train)
```

```
predicted = pipeline.predict(messages_test)
              accuracy = np.mean(predicted == categories_test)
              accuracy_per_fold.append(accuracy)
             print(accuracy)
         print("Total average accuracy: ", np.mean(accuracy_per_fold))
0.879365079365
0.879365079365
0.8888888889
0.88253968254
0.87619047619
0.879365079365
0.873015873016
0.87619047619
0.869841269841
0.806349206349
('Total average accuracy: ', 0.871111111111111)
     Run Classification and Evaluation
2.9
In [915]: aggressive_train = [tweet.text for tweet in results_aggressive]
         non_aggressive_train = [tweet.text for tweet in results_non_aggressive]
         messages_train = aggressive_train + non_aggressive_train
          categories_train = ['aggressive'] * len(aggressive_train) \
                           + ['non_aggressive'] * len(non_aggressive_train)
         pipeline = pipeline.fit(messages_train, categories_train)
          accuracy_per_fold = []
          for x in xrange(0, 10):
             test = frame.ix[create_key(x)]
             messages_test = test['message']
              categories_test = test['category']
             predicted = pipeline.predict(messages_test)
              accuracy = np.mean(predicted == categories_test)
              accuracy_per_fold.append(accuracy)
              print(metrics.classification_report(categories_test, predicted))
         print("Total average accuracy: ", np.mean(accuracy_per_fold))
precision
            recall f1-score
                                support
 aggressive
                  0.68
                            0.78
                                      0.73
                                                 158
                              0.64
                                        0.68
                                                   157
non_aggressive
                    0.74
                  0.71
                            0.71
                                      0.71
                                                 315
avg / total
            precision recall f1-score
                                            support
```

aggressive non_aggressive				
avg / total	0.75	0.75	0.75	315
pr	ecision	recall	f1-score	support
aggressive non_aggressive				
avg / total	0.77	0.77	0.77	315
pr	ecision	recall	f1-score	support
aggressive non_aggressive				
avg / total	0.72	0.72	0.72	315
pr	ecision	recall	f1-score	support
aggressive non_aggressive				
avg / total	0.73	0.73	0.73	315
pr	ecision	recall	f1-score	support
aggressive non_aggressive				
avg / total	0.71	0.71	0.71	315
pr	ecision	recall	f1-score	support
aggressive non_aggressive				
avg / total	0.69	0.69	0.69	315
pr	ecision	recall	f1-score	support
aggressive non_aggressive				
avg / total	0.71	0.71	0.71	268
pr	ecision	recall	f1-score	support
aggressive non_aggressive				
avg / total	0.70	0.70	0.70	312

```
precision
                         recall f1-score
                                             support
                  0.70
                            0.74
                                      0.72
                                                 157
 aggressive
non_aggressive
                    0.73
                              0.69
                                        0.71
                                                    158
avg / total
                  0.71
                            0.71
                                      0.71
                                                 315
('Total average accuracy: ', 0.71949747599001335)
In [924]: interactive_test = [raw_input('Please type message that should be categorized: ')]
          train = frame.ix[[create_key(y) for y in range(0,10)]]
         messages_train = train['message']
          categories_train = train['category']
          # train classifier
         pipeline = pipeline.fit(messages_train, categories_train)
         predicted = pipeline.predict(interactive_test)
         print(predicted)
Please type message that should be categorized: eikel
['non_aggressive']
2.10 Inspect Bag of Words
In [851]: matrix = count_vect.fit_transform(messages_train, categories_train)
         freqs = [(word, matrix.getcol(idx).sum()) for word, idx in count_vect.vocabulary_.items()]
          #sort from largest to smallest
         print sorted (freqs, key = lambda x: -x[1])
[('user', 814), ('url', 517), ('...', 199), ('fuck', 156), ('(', 155), (')', 154), ('kleur', 152), ('ja
2.11 Pickle Load
In [642]: # Getting back the objects:
          with open('results_854.pickle') as f:
              results_aggressive, results_non_aggressive = pickle.load(f)
In [94]: # Saving the objects:
         with open('results_479.pickle', 'w') as f:
             pickle.dump([results_aggressive, results_non_aggressive], f)
```