sad-python3

March 22, 2017

```
In [1]: import logging
        import sys
        log = logging.getLogger("P/R/ROC")
        log.setLevel(logging.INFO)
        ch = logging.StreamHandler()
        ch.setLevel(logging.INFO)
        formatter = logging.Formatter("%(asctime)s %(levelname)s:%(name)s %(message)s")
        ch.setFormatter(formatter)
        log.addHandler(ch)
        log.info("Now logging")
2017-03-21 20:01:57,470 INFO:P/R/ROC Now logging
In [2]: # MAX_TWEETS = 1_578_628
        MAX_TWEETS = 100_001
        # http://www.cs.ou.edu/~cgrant/teaching/cs5970sp16/sad/sad.html
        import io
        feature_list = []
        with io.open("sad.thorn", "r", encoding='utf-8') as source:
            labels = []
            for i,line in enumerate(source.readlines()):
                if i == MAX_TWEETS: break
                elif i == 0:
                    labels = line.split('b')
                else:
                    data = line.split('b')
                    feature_list.append({'id': int(data[0]),
                                          'label': '+' if (data[1] == '1') else '-',
                                          'text' : data[2]})
        #print(feature_list[0:10])
```

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pos_tweets = [x for x in feature_list if x['label'] == '+']
        neg_tweets = [x for x in feature_list if x['label'] == '-']
        log.info("pos: {}, neg: {}".format(len(pos_tweets), len(neg_tweets)))
2017-03-21 20:01:58,524 INFO:P/R/ROC pos: 56462, neg: 43538
In [3]: # Create big [({feature}, label)]
        import nltk
        from nltk import word_tokenize
        import collections
        import nltk.util
        from nltk.util import ngrams
        GRAM_SIZE = 1
        # postive feature list
        pos_features = []
        for item in pos_tweets:
            for words in ngrams(word_tokenize(item['text']), GRAM_SIZE):
                word = ' '.join(words)
                pos_features.append({'feature': word})
        neg_features = []
        for item in neg_tweets:
            for words in ngrams(word_tokenize(item['text']), GRAM_SIZE):
                word = ' '.join(words)
                neg_features.append({'feature': word})
        # Remove stop words
        from nltk.corpus import stopwords
        stopwords = stopwords.words('english')
        #print(stopwords)
        # TODO filter tweets by stopwords
        pos_features = [x for x in pos_features if x["feature"] not in stopwords]
        neg_features = [x for x in neg_features if x["feature"] not in stopwords]
In [4]: import random
        from random import shuffle
        SPLIT = 0.75
        shuffle(pos_features)
        shuffle(neg_features)
```

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POS_TRAIN_SIZE = int(len(pos_features) * SPLIT)
        NEG_TRAIN_SIZE = int(len(neg_features) * SPLIT)
        train_set = [(feat, '+') for feat in pos_features[:POS_TRAIN_SIZE]]
        train_set += [(feat, '-') for feat in neg_features[:NEG_TRAIN_SIZE]]
        test_set = [(feat, '+') for feat in pos_features[POS_TRAIN_SIZE:]]
        test_set += [(feat, '-') for feat in neg_features[NEG_TRAIN_SIZE:]]
        shuffle(test_set)
        shuffle(train_set)
0.0.1 Naive Bayes Classifier
In [5]: from nltk.classify import NaiveBayesClassifier
        log.info("Training NB")
        nb_classifier = NaiveBayesClassifier.train(train_set)
        log.info("Finished Training NB. classes: {}".format(nb_classifier.labels()))
2017-03-21 20:02:31,957 INFO:P/R/ROC Training NB
2017-03-21 20:02:34,154 INFO:P/R/ROC Finished Training NB. classes: ['+', '-']
0.0.2 Decision Tree Classifier
In [6]: from nltk.classify import DecisionTreeClassifier
        log.info("Training DT")
        dt_classifier = DecisionTreeClassifier.train(train_set)
        log.info("Finished Training DT. classes: {}".format(dt_classifier.labels()))
2017-03-21 20:02:34,161 INFO:P/R/ROC Training DT
        KeyboardInterrupt
                                                  Traceback (most recent call last)
        <ipython-input-6-2ed8c7eef75b> in <module>()
          3 log.info("Training DT")
    ----> 4 dt_classifier = DecisionTreeClassifier.train(train_set)
          5 log.info("Finished Training DT. classes: {}".format(dt_classifier.labels()))
        /usr/local/lib/python3.6/site-packages/nltk/classify/decisiontree.py in train(labeled_fe
        159
                    # Refine the stump.
```

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160
                tree.refine(labeled_featuresets, entropy_cutoff, depth_cutoff-1,
--> 161
                             support_cutoff, binary, feature_values, verbose)
    162
    163
                # Return it
    /usr/local/lib/python3.6/site-packages/nltk/classify/decisiontree.py in refine(self, lab
                for fval in self._decisions:
    193
    194
                    fval_featuresets = [(featureset, label) for (featureset, label)
--> 195
                                         in labeled_featuresets
    196
                                         if featureset.get(self._fname) == fval]
    197
    /usr/local/lib/python3.6/site-packages/nltk/classify/decisiontree.py in <listcomp>(.0)
                if depth_cutoff <= 0: return</pre>
    192
    193
                for fval in self._decisions:
--> 194
                    fval_featuresets = [(featureset, label) for (featureset, label)
    195
                                         in labeled_featuresets
    196
                                         if featureset.get(self._fname) == fval]
```

${\tt KeyboardInterrupt:}$

0.0.3 MaxentClassifier

==> Training (100 iterations)

Iteration	Log Likelihood	Accuracy
1	-0.69315	0.438
2	-0.59092	0.654
3	-0.57451	0.654
4	-0.56479	0.654
5	-0.55836	0.654
6	-0.55380	0.654
7	-0.55039	0.654
8	-0.54774	0.654

9	-0.54563	0.654
10	-0.54391	0.654
11	-0.54248	0.654
12	-0.54127	0.654
13	-0.54023	0.654
14	-0.53933	0.654
15	-0.53854	0.654
16	-0.53785	0.654
17	-0.53724	0.654
18	-0.53 <i>f</i> 24 -0.53669	
19		0.654
	-0.53619	0.654
20	-0.53574	0.654
21	-0.53534	0.654
22	-0.53497	0.654
23	-0.53463	0.654
24	-0.53431	0.654
25	-0.53403	0.654
26	-0.53376	0.654
27	-0.53351	0.654
28	-0.53328	0.654
29	-0.53306	0.654
30	-0.53286	0.654
31	-0.53267	0.654
32	-0.53250	0.654
33	-0.53233	0.654
34	-0.53217	0.654
35	-0.53202	0.654
36	-0.53188	0.654
37	-0.53175	0.654
38	-0.53162	0.654
39	-0.53150	0.654
40	-0.53139	0.654
41	-0.53128	0.654
42	-0.53118	0.654
43	-0.53108	0.654
44	-0.53098	0.654
45	-0.53089	0.654
46	-0.53081	0.654
47	-0.53072	0.654
48	-0.53065	0.654
49	-0.53057	0.654
50	-0.53050	0.654
51	-0.53043	0.654
52	-0.53036	0.654
53	-0.53029	0.654
54	-0.53023	0.654
55	-0.53017	0.654
56	-0.53011	0.654
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90 -0.52889 0.654 91 -0.52886 0.654 92 -0.52884 0.654 93 -0.52882 0.654 94 -0.52880 0.654 95 -0.52878 0.654 96 -0.52876 0.654 97 -0.52874 0.654 98 -0.52872 0.654			
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60 -0.52990 0.654 61 -0.52985 0.654 62 -0.52980 0.654 63 -0.52975 0.654 64 -0.52971 0.654 65 -0.52966 0.654 66 -0.52962 0.654 67 -0.52958 0.654 68 -0.52954 0.654 69 -0.52950 0.654 70 -0.52946 0.654 71 -0.52943 0.654 72 -0.52939 0.654 73 -0.52939 0.654 74 -0.52932 0.654 75 -0.52929 0.654 77 -0.52923 0.654 78 -0.52920 0.654 79 -0.52917 0.654 81 -0.52917 0.654 82 -0.52908 0.654 83 -0.52901 0.654 84 -0.52903 0.654 85 -0.52903 0.654 86 -0.52898 0.654	58	-0.53000	0.654
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90 -0.52889 0.654 91 -0.52886 0.654 92 -0.52884 0.654 93 -0.52882 0.654 94 -0.52880 0.654 95 -0.52878 0.654 96 -0.52876 0.654 97 -0.52874 0.654 98 -0.52872 0.654	88	-0.52893	0.654
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95 -0.52878 0.654 96 -0.52876 0.654 97 -0.52874 0.654 98 -0.52872 0.654	93	-0.52882	0.654
96 -0.52876 0.654 97 -0.52874 0.654 98 -0.52872 0.654	94	-0.52880	0.654
97 -0.52874 0.654 98 -0.52872 0.654	95	-0.52878	0.654
98 -0.52872 0.654	96	-0.52876	0.654
	97	-0.52874	0.654
99 -0 52870 0 654	98	-0.52872	0.654
0.02010 0.004	99	-0.52870	0.654

2017-03-22 11:18:43,912 INFO:P/R/ROC Finished Training M. classes: ['-', '+']

Final -0.52868 0.654

0.0.4 Get precision recall f1 measure of each (http://scikit-learn.org/stable/modules/generated/sklearn.metrics.precision_recall_fscore_support.html)

```
In [8]: import sklearn
    import sklearn.metrics
    from sklearn.metrics import precision_recall_fscore_support

import numpy as np
    from operator import itemgetter

def prfs(classify, test_set):
        truth = np.array([b for (a,b) in test_set])
        prediction = np.array([classify(a) for (a,b) in test_set])
        return precision_recall_fscore_support(truth, prediction)
```

0.0.5 Run the Naive Bayes Classier

0.0.6 Plotting the ROC curve (http://blog.yhat.com/posts/roc-curves.html)

```
In [10]: from sklearn.metrics import roc_curve
    from sklearn.metrics import auc
    import pandas as pd
    import ggplot
    from ggplot import aes
    from ggplot import geom_abline
    from ggplot import geom_line
    from ggplot import ggplot
    from ggplot import ggtitle
    import matplotlib.pyplot as plt

# An iPython directive for inline graphs
    %matplotlib inline

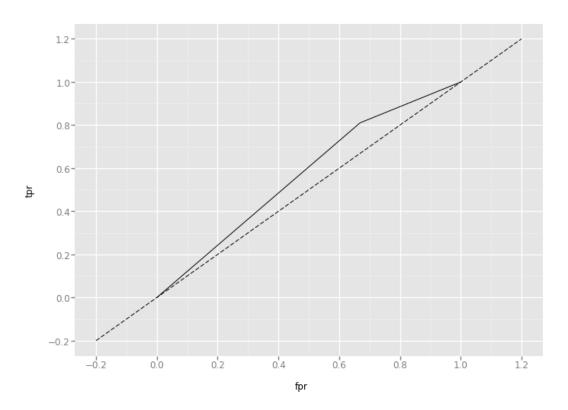
def plus2one(x): return 1 if x == '+' else 0

def plot(classify, test_set):
    truth = [plus2one(b) for (a,b) in test_set]
```

```
data = [plus2one(classify(a)) for (a,b) in test_set]
    fpr, tpr, thresholds = roc_curve(truth, data, drop_intermediate=False)
    roc_auc = auc(fpr, tpr)
   df = pd.DataFrame(dict(fpr=fpr, tpr=tpr))
   g = ggplot(df, aes(x='fpr', y='tpr')) + geom_line() + geom_abline(linetype='dashed
    g += ggtitle("ROC Curve: AUC({})".format(roc_auc))
    return g
def matplot(classify, test_set):
    truth = [plus2one(b) for (a,b) in test_set]
    data = [plus2one(classify(a)) for (a,b) in test_set]
    fpr, tpr, thresholds = roc_curve(truth, data, drop_intermediate=False)
    roc_auc = auc(fpr, tpr)
   plt.plot(fpr, tpr, label='ROC curve (area = %0.2f)' % roc_auc)
   plt.xlabel('False Positive Rate')
   plt.ylabel('True Positive Rate')
    plt.legend(loc="lower right")
   return plt
```

In [13]: plot(nb_classifier.classify, test_set)

ROC Curve: AUC(0.5708424327103738)



```
Out[13]: <ggplot: (320277708)>
```

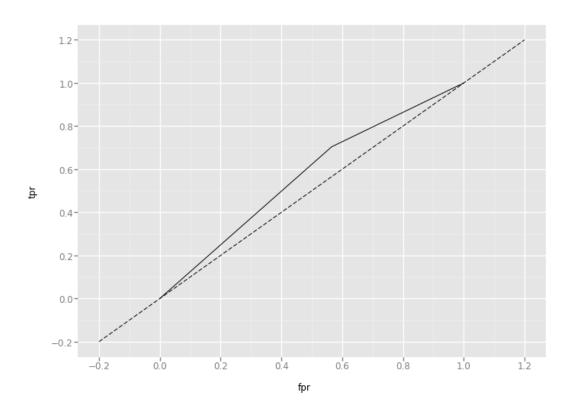
0.0.7 Run the Decision Tree Classifier

In []: plot(dt_classifier.classify, test_set)

0.0.8 Run the Maxent Classifier

In [12]: plot(m_classifier.classify, test_set)

ROC Curve: AUC(0.5693038554050354)



```
Out[12]: <ggplot: (-9223372036538259452)>
In []:
```