GBDT and Random Forests bow tfidf[M]

June 2, 2018

In [19]: !pip install gensim !pip install imblearn !pip install xgboost

Requirement already satisfied: gensim in /usr/local/lib/python3.6/dist-packages (3.4.0) Requirement already satisfied: smart-open>=1.2.1 in /usr/local/lib/python3.6/dist-packages (from gensim) (1.5.7) Requirement already satisfied: scipy>=0.18.1 in /usr/local/lib/python3.6/dist-packages (from gensim) (0.19.1) Requirement already satisfied: numpy>=1.11.3 in /usr/local/lib/python3.6/dist-packages (from gensim) (1.14.3) Requirement already satisfied: six>=1.5.0 in /usr/local/lib/python3.6/dist-packages (from gensim) (1.11.0) Requirement already satisfied: bz2file in /usr/local/lib/python3.6/dist-packages (from smart-open>=1.2.1->gensir Requirement already satisfied: requests in /usr/local/lib/python3.6/dist-packages (from smart-open>=1.2.1->gen Requirement already satisfied: boto>=2.32 in /usr/local/lib/python3.6/dist-packages (from smart-open>=1.2.1->g Requirement already satisfied: boto3 in /usr/local/lib/python3.6/dist-packages (from smart-open>=1.2.1->gensim Requirement already satisfied: idna<2.7,>=2.5 in /usr/local/lib/python3.6/dist-packages (from requests->smart-o Requirement already satisfied: urllib3<1.23,>=1.21.1 in /usr/local/lib/python3.6/dist-packages (from requests->sr Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.6/dist-packages (from requests->smar Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist-packages (from requests->s Requirement already satisfied: botocore<1.11.0,>=1.10.31 in /usr/local/lib/python3.6/dist-packages (from boto3-Requirement already satisfied: s3transfer<0.2.0,>=0.1.10 in /usr/local/lib/python3.6/dist-packages (from boto3-> Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in /usr/local/lib/python3.6/dist-packages (from boto3->sn Requirement already satisfied: python-dateutil<3.0.0,>=2.1; python version >= "2.7" in /usr/local/lib/python3.6/d Requirement already satisfied: docutils>=0.10 in /usr/local/lib/python3.6/dist-packages (from botocore<1.11.0,>= Requirement already satisfied: imblearn in /usr/local/lib/python3.6/dist-packages (0.0) Requirement already satisfied: imbalanced-learn in /usr/local/lib/python3.6/dist-packages (from imblearn) (0.3.3) Requirement already satisfied: numpy in /usr/local/lib/python3.6/dist-packages (from imbalanced-learn->imblear Requirement already satisfied: scipy in /usr/local/lib/python3.6/dist-packages (from imbalanced-learn->imblearn) Requirement already satisfied: scikit-learn in /usr/local/lib/python3.6/dist-packages (from imbalanced-learn->imb Requirement already satisfied: xgboost in /usr/local/lib/python3.6/dist-packages (0.7.post4) Requirement already satisfied: numpy in /usr/local/lib/python3.6/dist-packages (from xgboost) (1.14.3) Requirement already satisfied: scipy in /usr/local/lib/python3.6/dist-packages (from xgboost) (0.19.1)

In [0]: from sklearn.model_selection import train_test_split

from sklearn.grid_search import GridSearchCV from sklearn.grid_search import RandomizedSearchCV from scipy.stats import randint

```
from imblearn.over_sampling import SMOTE
         import sqlite3
         import pandas as pd
         import numpy as np
         from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.feature extraction.text import CountVectorizer
         import gensim
         from sklearn.metrics import classification report, accuracy score, confusion matrix
         from sklearn.ensemble import RandomForestClassifier
         from xgboost.sklearn import XGBClassifier
In [21]: !pip install PyDrive
          from pydrive.auth import GoogleAuth
          from pydrive.drive import GoogleDrive
          from google.colab import auth
          from oauth2client.client import GoogleCredentials
          # 1. Authenticate and create the PyDrive client.
          auth.authenticate user()
          gauth = GoogleAuth()
          gauth.credentials = GoogleCredentials.get application default()
          drive = GoogleDrive(gauth)
          file list = drive.ListFile({'q': "'1pbLvjcsi6UtFm3sPciCJGbCG4NK3uyuS' in parents and trashed=false"}
          for file1 in file list:
             print('title: %s, id: %s' % (file1['title'], file1['id']))
          sql = drive.CreateFile({'id': '1OzLc3k6-T55I-XRMq47ERyCbQbVw4caF'})
          sql.GetContentFile('final.sqlite')
Requirement already satisfied: PyDrive in /usr/local/lib/python3.6/dist-packages (1.3.1)
Requirement already satisfied: google-api-python-client>=1.2 in /usr/local/lib/python3.6/dist-packages (from PyD
Requirement already satisfied: PyYAML>=3.0 in /usr/local/lib/python3.6/dist-packages (from PyDrive) (3.12)
Requirement already satisfied: oauth2client>=4.0.0 in /usr/local/lib/python3.6/dist-packages (from PyDrive) (4.1.
Requirement already satisfied: httplib2<1dev,>=0.9.2 in /usr/local/lib/python3.6/dist-packages (from google-api-packages)
Requirement already satisfied: uritemplate<4dev,>=3.0.0 in /usr/local/lib/python3.6/dist-packages (from google-a
Requirement already satisfied: six<2dev,>=1.6.1 in /usr/local/lib/python3.6/dist-packages (from google-api-python3.6/dist-packages)
```

title: GBDT and Random Forests [M].ipynb, id: 1NauX7hmr_HdwByih8sRQlekzLugLSkll title: Apply SVM to Amazon reviews data set avg_w2vec [M].ipynb, id: 1ElWunFgWZPb1lq6w4ZMmqoBSVuPt00 title: Apply Logistic regression to Amazon reviews data set. [M].ipynb, id: 1Es1wP2edJ0vrKasA5wYJEO-zeZvrqe

Requirement already satisfied: rsa>=3.1.4 in /usr/local/lib/python3.6/dist-packages (from oauth2client>=4.0.0->F Requirement already satisfied: pyasn1-modules>=0.0.5 in /usr/local/lib/python3.6/dist-packages (from oauth2client>=4.0. Requirement already satisfied: pyasn1>=0.1.7 in /usr/local/lib/python3.6/dist-packages (from oauth2client>=4.0.

```
title: Apply Naive Bayes to Amazon reviews [M].ipynb, id: 1qPxAZeYQUM-eqaKnOSM5ubK2IPIVmdyo
title: clean final.sqlite, id: 1T0HyUqaVFyD8HflQEM6WN8jF8SpEOsAo
title: KNN on Credit Card fraud detection.ipynb, id: 1CkA-RBfXqvubKkQrpnjbYUKVsC7VHITI
title: creditcard.csv, id: 1VpegIS0IPVrlzIMIgvQTzc3Pno Cj4SV
title: creditcard.csv, id: 1bnZktEq3N 5wjoCH85oIXHxNwXUW jx-
title: Untitled, id: 1K0wwkizWx3WO8d-zw-YewWIUrPdINYmp
title: final.sqlite, id: 10zLc3k6-T55I-XRMq47ERyCbQbVw4caF
title: HeavyComputations.ipynb, id: 1aBORe3ggeFY-iNhzMtr-TlkzEyEvFxcG
title: LogisticRegression.ipynb, id: 1WcVTklMZBMu9VTClWeupOK0r2aYbHk8p
In [0]: con = sqlite3.connect('final.sqlite') # this is cleaned dataset
         final = pd.read sql query("""
         SELECT Score, Text not included
         FROM reviews
         "", con)[:2000]
         for i, seq in enumerate(final['Text not included']):
           final['Text not included'][i]=final['Text not included'][i].decode('UTF-8')
         X_train, X_test, y_train, y_test = train_test_split(final['Text_not_included'], final['Score'], test_size=0.2,
In [0]: # Generate count BoW
         count vect = CountVectorizer(ngram range=(1,2))
         count vect.fit(X train)
         bow train=count vect.transform(X train)
         bow test=count vect.transform(X test)
         # Generate tf idf
         tf idf vect=TfidfVectorizer(ngram_range=(1,2), min_df=10, dtype=float)
         tf idf vect.fit(X train)
         tf idf train=tf idf vect.transform(X train)
         tf idf test=tf idf vect.transform(X test)
         # Generate average word2vec
         sentences=[]
         for review in X train:
           sentence=[]
           for word in review.split():
             sentence.append(word)
           sentences.append(sentence)
         w2vec model=gensim.models.word2vec.Word2Vec(sentences, min count=10)
         avg w2vec train=np.zeros(shape=(len(X train), 100), dtype=float)
         for i, sentence in enumerate(sentences):
           for word in sentence:
             try:
```

```
avg w2vec train[i]+=w2vec model.wv[word]
    except KeyError:
       pass
  avg w2vec train[i]/=len(sentence)
sentences=[]
for review in X test:
  sentence=[]
  for word in review.split():
    sentence.append(word)
  sentences.append(sentence)
avg_w2vec_test=np.zeros(shape=(len(X_test), 100), dtype=float)
for i, sentence in enumerate(sentences):
  for word in sentence:
    try:
      avg w2vec test[i]+=w2vec model.wv[word]
    except KeyError:
      pass
  avg w2vec test[i]/=len(sentence)
# Generate tf idf weighted word2vec
sentences=[]
for review in X train:
  sentence=[]
  for word in review.split():
    sentence.append(word)
  sentences.append(sentence)
tf idf w2vec train=np.zeros((len(X train), 100), dtype=float)
feat=tf idf vect.get feature names()
for i, sentence in enumerate(sentences):
  tf idf sum=0
  for word in sentence:
    try:
      tf idf w2vec train[i]+=w2vec model.wv[word]*tf idf train[i, feat.index(word)]
      tf idf sum+=tf idf train[i, feat.index(word)]
    except KeyError:
      pass
    except ValueError:
  tf_idf_w2vec_train[i]/=tf_idf_sum
```

```
for review in X_test:
           sentence=[]
           for word in review.split():
             sentence.append(word)
           sentences.append(sentence)
         tf_idf_w2vec_test=np.zeros((len(X_test), 100), dtype=float)
         for i, sentence in enumerate(sentences):
           tf idf sum=0
           for word in sentence:
             try:
                tf_idf_w2vec_test[i]+=w2vec_model.wv[word]*tf_idf_test[i, feat.index(word)]
                tf_idf_sum+=tf_idf_test[i, feat.index(word)]
             except KeyError:
                pass
             except ValueError:
                pass
           tf idf w2vec test[i]/=tf idf sum
In [0]: # Upsampling minority class
         over sampler = SMOTE(ratio='minority')
         bow_train_resampled, y_train_resampled = over_sampler.fit_sample(bow_train, y_train)
         tf idf train resampled, y train resampled = over sampler.fit sample(tf idf train, y train)
         avg_w2vec_train_resampled, y_train_resampled = over_sampler.fit_sample(avg_w2vec_train, y_train)
         tf idf w2vec train resampled, y train resampled = over sampler.fit sample(tf idf w2vec train, y tra
    Classification using RandomForest
In [0]: tuned_parameters = {'n_estimators': np.arange(1,100,1)}
         gscv = GridSearchCV(RandomForestClassifier(n_jobs=-1), tuned_parameters, scoring = 'accuracy', cv
         tuned parameters = {'n estimators' : randint(low=1, high=101)}
         rscv = RandomizedSearchCV(RandomForestClassifier(n_jobs=-1), tuned_parameters, scoring = 'accu
1.0.1 Bow
In [16]: gscv.fit(bow_train_resampled, y_train_resampled)
          predictions=gscv.best_estimator_.predict(bow_test)
          print(classification_report(y_test, predictions))
          print(confusion_matrix(y_test, predictions).T)
          tn, fp, fn, tp = confusion matrix(y test, predictions).ravel()
          print("TPR = {} n TNR = {} n FPR = {} n FNR = {} (fn+tp), tn/(fn+tp), fp/(tn+fp), fn/(fn+tp)))
```

sentences=[]

```
precision
                              recall
                                       f1-score
                                                   support
                    0.00
                                0.00
                                            0.00
                                                         87
   negative
   positive
                    0.78
                                1.00
                                            88.0
                                                        313
avg / total
                    0.61
                                0.78
                                           0.69
                                                        400
\prod
    0
       0]
[87 313]]
TPR = 1.0
 TNR = 0.0
 FPR = 1.0
 FNR = 0.0
/usr/local/lib/python3.6/dist-packages/sklearn/metrics/classification.py:1135: UndefinedMetricWarning: Precision
  'precision', 'predicted', average, warn for)
In [17]: gscv.best_estimator_
Out[17]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                        max_depth=None, max_features='auto', max_leaf_nodes=None,
                        min_impurity_decrease=0.0, min_impurity_split=None,
                        min samples leaf=1, min samples split=2,
                        min_weight_fraction_leaf=0.0, n_estimators=20, n_jobs=-1,
                        oob_score=False, random_state=None, verbose=0,
                        warm start=False)
In [14]: rscv.fit(bow_train_resampled, y_train_resampled)
          predictions=rscv.best_estimator_.predict(bow_test)
          print(classification_report(y_test, predictions))
          print(confusion_matrix(y_test, predictions).T)
          tn, fp, fn, tp = confusion matrix(y test, predictions).ravel()
          print("TPR = {\n TNR = }\n FPR = {\n FNR = {\n FNR = {\n fn/(fn+tp), fn/(tn+fp), fn/(fn+tp))}}
              precision
                              recall
                                       f1-score
                                                   support
                    0.33
                                0.01
                                            0.02
                                                         87
   negative
                                0.99
                                            88.0
   positive
                    0.78
                                                        313
avg / total
                    0.69
                                0.78
                                           0.69
                                                        400
    1
         2]
[[
[ 86 311]]
```

TPR = 0.9936102236421726 TNR = 0.011494252873563218

```
FPR = 0.9885057471264368
 FNR = 0.006389776357827476
In [15]: rscv.best_estimator_
Out[15]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                        max_depth=None, max_features='auto', max_leaf_nodes=None,
                        min_impurity_decrease=0.0, min_impurity_split=None,
                        min samples leaf=1, min samples split=2,
                        min_weight_fraction_leaf=0.0, n_estimators=24, n_jobs=-1,
                        oob score=False, random state=None, verbose=0,
                        warm_start=False)
1.0.2 TF IDF
In [15]: gscv.fit(tf_idf_train_resampled, y_train_resampled)
          predictions=gscv.best_estimator_.predict(tf_idf_test)
          print(classification report(y test, predictions))
          print(confusion_matrix(y_test, predictions).T)
          tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
          print("TPR = {} \ TNR = {} \ FPR = {} \ FNR = {} \ format(tp/(fn+tp), tn/(tn+fp), fp/(tn+fp), fn/(fn+tp)))
                             recall
                                      f1-score
               precision
                                                   support
                                0.11
                                                         87
   negative
                    0.42
                                           0.18
                    0.80
                                0.96
                                           0.87
                                                        313
   positive
                    0.71
                                0.77
                                           0.72
                                                        400
avg / total
[[ 10
        14]
[77 299]]
TPR = 0.9552715654952076
 TNR = 0.11494252873563218
 FPR = 0.8850574712643678
 FNR = 0.04472843450479233
In [16]: rscv.fit(tf_idf_train_resampled, y_train_resampled)
          predictions=rscv.best_estimator_.predict(tf_idf_test)
          print(classification_report(y_test, predictions))
          print(confusion matrix(y test, predictions).T)
          tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
          print("TPR = {} \ TNR = {} \ FNR = {} \ format(tp/(fn+tp), tn/(tn+fp), fp/(tn+fp), fn/(fn+tp)))
```

	precision	recall	f1-score	support
negative positive	0.54 0.80	0.15 0.96	0.23 0.88	87 313
avg / total	0.75	0.79	0.74	400
[[13				

In [17]: gscv.best_estimator_

```
Out[17]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini', max_depth=None, max_features='auto', max_leaf_nodes=None, min_impurity_decrease=0.0, min_impurity_split=None, min_samples_leaf=1, min_samples_split=2, min_weight_fraction_leaf=0.0, n_estimators=95, n_jobs=-1, oob_score=False, random_state=None, verbose=0, warm_start=False)
```

In [18]: rscv.best_estimator

1.1 Classification using GBDT

1.1.1 BoW

In [0]: from scipy.stats import uniform

```
tuned_parameters = {'n_estimators': np.arange(10,60,10),'max_depth' : np.arange(1,5,1), 'learning_rate gscv = GridSearchCV(XGBClassifier(), tuned_parameters, scoring = 'accuracy', cv=5)

tuned_parameters = {'n_estimators': randint(low=10, high=61), 'max_depth' : randint(low=1, high=6), 'learning_rate gscv = RandomizedSearchCV(XGBClassifier(), tuned_parameters, scoring = 'accuracy', cv=5, n_iter=2

In [38]: gscv.fit(bow_train_resampled, y_train_resampled)
```

```
predictions=gscv.best_estimator_.predict(bow_test)
print(classification_report(y_test, predictions))
print(confusion_matrix(y_test, predictions).T)
tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
```

```
print("TPR = {\n TNR = {\n FPR = {\n FNR = {\n FNR = {\n fn/(fn+tp), fn/(tn+fp), fn/(fn+tp))}}}
```

- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value

	precision	recall	f1-score	support
negative positive	0.77 0.80	0.11 0.99	0.20 0.89	87 313
avg / total	0.79	0.80	0.74	400

[[10 3] [77 310]]

TPR = 0.9904153354632588 TNR = 0.11494252873563218 FPR = 0.8850574712643678

FNR = 0.009584664536741214

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

```
In [41]: rscv.fit(bow_train_resampled, y_train_resampled)
```

```
predictions=rscv.best_estimator_.predict(bow_test)
print(classification_report(y_test, predictions))
print(confusion_matrix(y_test, predictions).T)
tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
```

```
print("TPR = {\n TNR = {\n FNR = {
```

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value of different di

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

	precision	recall	f1-score	support
negative positive	0.55 0.79	0.07 0.98	0.12 0.88	87 313
avg / total	0.74	0.79	0.71	400
[[6 5]				

```
[81 308]]
TPR = 0.9840255591054313
 TNR = 0.06896551724137931
FPR = 0.9310344827586207
 FNR = 0.01597444089456869
/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value
  if diff:
In [42]: gscv.best_estimator_
Out[42]: XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1,
                  colsample_bytree=1, gamma=0, learning_rate=0.275, max_delta_step=0,
                  max depth=2, min child weight=1, missing=None, n estimators=50,
                  n_jobs=1, nthread=None, objective='binary:logistic', random_state=0,
                  reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None,
                  silent=True, subsample=1)
In [43]: rscv.best_estimator
Out[43]: XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1,
                  colsample bytree=1, gamma=0, learning rate=0.20249593153909556,
                  max_delta_step=0, max_depth=4, min_child_weight=1, missing=None,
                  n estimators=31, n jobs=1, nthread=None,
                  objective='binary:logistic', random_state=0, reg_alpha=0,
                  reg lambda=1, scale pos weight=1, seed=None, silent=True,
                  subsample=1)
1.1.2 TF IDF
In [44]: gscv.fit(tf idf train resampled, y train resampled)
          predictions=gscv.best_estimator_.predict(tf_idf_test)
          print(classification_report(y_test, predictions))
          print(confusion_matrix(y_test, predictions).T)
          tn, fp, fn, tp = confusion matrix(y test, predictions).ravel()
          print("TPR = {} \ TNR = {} \ FPR = {} \ FNR = {} \ format(tp/(fn+tp), tn/(tn+fp), fp/(tn+fp), fn/(fn+tp)))
```

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

	precision	recall	f1-score	support
negative positive	0.57 0.83	0.32 0.93	0.41 0.88	87 313
avg / total	0.78	0.80	0.78	400

[[28 21] [59 292]]

TPR = 0.9329073482428115

TNR = 0.3218390804597701

FPR = 0.6781609195402298

FNR = 0.0670926517571885

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

```
predictions=rscv.best_estimator_.predict(tf_idf_test)
print(classification_report(y_test, predictions))
print(confusion_matrix(y_test, predictions).T)
```

In [45]: rscv.fit(tf_idf_train_resampled, y_train_resampled)

tn, fp, fn, tp = confusion matrix(y test, predictions).ravel()

```
print("TPR = {}\n TNR = {}\n FPR = {}\n FNR = {}".format(tp/(fn+tp), tn/(tn+fp), fp/(tn+fp), fn/(fn+tp)))
```

- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:
- /usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value

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/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

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/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

	precision	recall	f1-score	support
negative positive	0.60 0.83	0.32 0.94	0.42 0.88	87 313
avg / total	0.78	0.81	0.78	400

[[28 19] [59 294]]

TPR = 0.939297124600639

TNR = 0.3218390804597701

FPR = 0.6781609195402298

FNR = 0.06070287539936102

/usr/local/lib/python3.6/dist-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

In [46]: gscv.best_estimator

Out[46]: XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1, colsample_bytree=1, gamma=0, learning_rate=0.5, max_delta_step=0, max_depth=4, min_child_weight=1, missing=None, n_estimators=50,

```
n_jobs=1, nthread=None, objective='binary:logistic', random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None, silent=True, subsample=1)
```

In [47]: rscv.best_estimator_

Out[47]: XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1, colsample_bytree=1, gamma=0, learning_rate=0.5238871512238155, max_delta_step=0, max_depth=5, min_child_weight=1, missing=None, n_estimators=47, n_jobs=1, nthread=None, objective='binary:logistic', random_state=0, reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None, silent=True, subsample=1)

2 Conclusions

Random forest performance

Using BoW TPR = 0.99 TNR = 0.01

best n_estimators = 93

Using TF IDF TPR = 0.96 TNR = 0.14

best n_estimators = 72

GBDT performance

Using BoW TPR = 0.99 TNR = 0.11

best n_estimators = 50 best max_depth = 2 best eta = 0.275

Using TF IDF TPR = 0.93 TNR = 0.32

best n_estimators = 50 best max_depth = 4 best eta = 0.5

TF IDF representations provide better results with both classifiers. GBDT provides better TNR.

GBDT combined with TF IDF provides optimal value for both TPR and TNR