GBDT and Random Forests w2vec tfidfw2vec[M]

June 2, 2018

In [2]: !pip install gensim !pip install imblearn !pip install xgboost

Requirement already satisfied: gensim in /usr/local/lib/python3.6/site-packages Requirement already satisfied: numpy>=1.11.3 in /usr/local/lib/python3.6/site-packages (from gensim) Requirement already satisfied: scipy>=0.18.1 in /usr/local/lib/python3.6/site-packages (from gensim) Requirement already satisfied: six>=1.5.0 in /usr/local/lib/python3.6/site-packages (from gensim) Requirement already satisfied: smart-open>=1.2.1 in /usr/local/lib/python3.6/site-packages (from gensim) Requirement already satisfied: bz2file in /usr/local/lib/python3.6/site-packages (from smart-open>=1.2.1->gensir Requirement already satisfied: boto3 in /usr/local/lib/python3.6/site-packages (from smart-open>=1.2.1->gensim Requirement already satisfied: boto>=2.32 in /usr/local/lib/python3.6/site-packages (from smart-open>=1.2.1->g Requirement already satisfied: requests in /usr/local/lib/python3.6/site-packages (from smart-open>=1.2.1->gen Requirement already satisfied: s3transfer<0.2.0,>=0.1.10 in /usr/local/lib/python3.6/site-packages (from boto3-> Requirement already satisfied: jmespath<1.0.0,>=0.7.1 in /usr/local/lib/python3.6/site-packages (from boto3->sn Requirement already satisfied: botocore<1.11.0,>=1.10.31 in /usr/local/lib/python3.6/site-packages (from boto3-Requirement already satisfied: python-dateutil<3.0.0,>=2.1; python_version >= "2.7" in /usr/local/lib/python3.6/si Requirement already satisfied: docutils>=0.10 in /usr/local/lib/python3.6/site-packages (from botocore<1.11.0,>= You are using pip version 9.0.1, however version 10.0.1 is available. You should consider upgrading via the 'pip i Requirement already satisfied: imblearn in /usr/local/lib/python3.6/site-packages Requirement already satisfied: imbalanced-learn in /usr/local/lib/python3.6/site-packages (from imblearn) Requirement already satisfied: numpy in /usr/local/lib/python3.6/site-packages (from imbalanced-learn->imblear Requirement already satisfied: scipy in /usr/local/lib/python3.6/site-packages (from imbalanced-learn->imblearn) Requirement already satisfied: scikit-learn in /usr/local/lib/python3.6/site-packages (from imbalanced-learn->imb You are using pip version 9.0.1, however version 10.0.1 is available. You should consider upgrading via the 'pip i Requirement already satisfied: xgboost in /usr/local/lib/python3.6/site-packages/xgboost-0.7-py3.6.egg Requirement already satisfied: numpy in /usr/local/lib/python3.6/site-packages (from xgboost) Requirement already satisfied: scipy in /usr/local/lib/python3.6/site-packages (from xgboost)

You are using pip version 9.0.1, however version 10.0.1 is available. You should consider upgrading via the 'pip i

In [3]: 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 [20]: 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 [21]: # 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:
       pass
  tf idf w2vec train[i]/=tf idf sum
sentences=[]
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
```

/usr/local/lib/python3.6/site-packages/sklearn/feature_extraction/text.py:1089: FutureWarning: Conversion of the if hasattr(X, 'dtype') and np.issubdtype(X.dtype, np.float):

0.1 Upsampling - Decision trees affected by imbalanced dataset

```
In [22]: # 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_train)
```

1 Classification using RandomForest

```
In [25]: tuned_parameters = {'n_estimators': np.arange(1,100,1)}
gscv = GridSearchCV(RandomForestClassifier(n_jobs=-1), tuned_parameters, scoring = 'accuracy', of
tuned_parameters = {'n_estimators' : randint(low=1, high=101)}
rscv = RandomizedSearchCV(RandomForestClassifier(n_jobs=-1), tuned_parameters, scoring = 'accuracy', of
tuned_parameters = {'n_estimators' : randint(low=1, high=101)}
```

```
1.0.1 Word2Vec
In [26]: gscv.fit(avg w2vec train resampled, y train resampled)
                                               predictions=gscv.best estimator .predict(avg w2vec 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 = {
                                                                                                                                        recall
                                                                                                                                                                                 f1-score
                                                                   precision
                                                                                                                                                                                                                                           support
                                                                                                                                                   0.17
                                                                                                                                                                                                      0.24
                                                                                                                                                                                                                                                                     87
               negative
                                                                                              0.38
                                                                                              0.80
                                                                                                                                                   0.92
                                                                                                                                                                                                       0.86
                                                                                                                                                                                                                                                                 313
               positive
                                                                                              0.71
                                                                                                                                                  0.76
                                                                                                                                                                                                      0.72
                                                                                                                                                                                                                                                                 400
avg / total
[[ 15
                                    24]
   [72 289]]
TPR = 0.9233226837060703
    TNR = 0.1724137931034483
    FPR = 0.8275862068965517
    FNR = 0.07667731629392971
In [27]: rscv.fit(avg w2vec train resampled, y train resampled)
                                               predictions=rscv.best estimator .predict(avg w2vec 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 = {
                                                                   precision
                                                                                                                                        recall
                                                                                                                                                                                 f1-score
                                                                                                                                                                                                                                           support
                                                                                                                                                   0.22
                                                                                                                                                                                                       0.29
                                                                                              0.41
                                                                                                                                                                                                                                                                     87
               negative
               positive
                                                                                              0.81
                                                                                                                                                  0.91
                                                                                                                                                                                                       0.86
                                                                                                                                                                                                                                                                 313
```

[[19 27] [68 286]]

avg / total

TPR = 0.9137380191693291

TNR = 0.21839080459770116

0.72

0.76

FPR = 0.7816091954022989

FNR = 0.08626198083067092

0.73

400

```
In [28]: gscv.best_estimator_
Out[28]: 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=99, n_jobs=-1,
                       oob score=False, random state=None, verbose=0,
                       warm_start=False)
In [29]: rscv.best_estimator_
Out[29]: 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=93, n jobs=-1,
                       oob score=False, random state=None, verbose=0,
                       warm start=False)
1.0.2 TF IDF weighted word2Vec
In [30]: gscv.fit(tf idf w2vec_train_resampled, y_train_resampled)
          predictions=gscv.best_estimator_.predict(tf_idf_w2vec_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)))
              precision
                             recall
                                      f1-score
                                                  support
                               0.07
                    0.23
                                           0.11
                                                        87
   negative
   positive
                    0.78
                               0.94
                                          0.85
                                                       313
                               0.75
                                          0.69
                                                       400
avg / total
                    0.66
    6 20]
\prod
[81 293]]
TPR = 0.9361022364217252
 TNR = 0.06896551724137931
 FPR = 0.9310344827586207
 FNR = 0.06389776357827476
In [31]: rscv.fit(tf_idf_w2vec_train_resampled, y_train_resampled)
          predictions=rscv.best estimator .predict(tf idf w2vec 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
                    0.25
                               80.0
                                          0.12
                                                       87
   negative
   positive
                    0.78
                               0.93
                                          0.85
                                                      313
                    0.67
                               0.75
                                          0.69
                                                      400
avg / total
    7 21]
[80 292]]
TPR = 0.9329073482428115
 TNR = 0.08045977011494253
 FPR = 0.9195402298850575
 FNR = 0.0670926517571885
In [32]: gscv.best_estimator_
Out[32]: 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=73, n_jobs=-1,
                       oob_score=False, random_state=None, verbose=0,
                       warm start=False)
In [33]: rscv.best_estimator_
Out[33]: 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=72, n_jobs=-1,
                       oob score=False, random state=None, verbose=0,
                       warm start=False)
```

Classification using GBDT

In [41]: from scipy.stats import uniform

```
tuned_parameters = {'n_estimators': np.arange(10,60,10), 'max_depth' : np.arange(1,5,1), 'learning_ra
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), 'max_dep rscv = RandomizedSearchCV(XGBClassifier(), tuned parameters, scoring = 'accuracy', cv=5, n iter=

2.0.1 Word2Vec

```
In [42]: gscv.fit(avg_w2vec_train_resampled, y_train_resampled)

predictions=gscv.best_estimator_.predict(avg_w2vec_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/site-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

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

	precision	recall	f1-score	support
negative positive	0.36 0.82	0.34 0.83	0.35 0.83	87 313
avg / total	0.72	0.72	0.72	400

[[30 53] [57 260]]

TPR = 0.8306709265175719

TNR = 0.3448275862068966

FPR = 0.6551724137931034

FNR = 0.16932907348242812

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

```
In [37]: rscv.fit(avg_w2vec_train_resampled, y_train_resampled)
```

```
predictions=rscv.best_estimator_.predict(avg_w2vec_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/site-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value if diff:

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

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

	precision	recall	f1-score	support
negative	0.37	0.31	0.34	87
positive	0.82	0.85	0.83	313

```
0.72
                               0.73
                                          0.73
                                                      400
avg / total
[[ 27
       46]
[ 60 267]]
TPR = 0.853035143769968
 TNR = 0.3103448275862069
 FPR = 0.6896551724137931
 FNR = 0.14696485623003194
/usr/local/lib/python3.6/site-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value
  if diff:
In [43]: gscv.best_estimator
Out[43]: 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 [39]: rscv.best_estimator
Out[39]: XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
                 colsample_bytree=1, gamma=0, learning_rate=0.30371216290769487,
                 max_delta_step=0, max_depth=5, min_child_weight=1, missing=None,
                 n_estimators=53, 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)
2.0.2 TF IDF weighted word2Vec
In [44]: gscv.fit(tf_idf_w2vec_train_resampled, y_train_resampled)
          predictions=gscv.best_estimator_.predict(tf_idf_w2vec_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 = {} \ fn/(fn+tp), tn/(tn+fp), fn/(fn+tp)))
/usr/local/lib/python3.6/site-packages/sklearn/preprocessing/label.py:151: DeprecationWarning: The truth value
  if diff:
```

if diff:

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

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

if diff:

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

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

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

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

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

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

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

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

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

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

	precision	recall	f1-score	support
negative positive	0.33 0.80	0.23 0.87	0.27 0.84	87 313
avg / total	0.70	0.73	0.71	400

[[20 40] [67 273]]

TPR = 0.8722044728434505

TNR = 0.22988505747126436

```
FPR = 0.7701149425287356
FNR = 0.12779552715654952
```

In [45]: rscv.fit(tf idf w2vec train resampled, y train resampled)

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

```
predictions=rscv.best_estimator_.predict(tf_idf_w2vec_test)
print(classification_report(y_test, predictions))
print(confusion_matrix(y_test, predictions).T)
tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
```

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

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

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

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

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

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

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

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

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

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

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

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	precision	recall	f1-score	support
negative positive	0.29 0.80	0.23 0.85	0.26 0.82	87 313
avg / total	0.69	0.71	0.70	400

[[20 48]

[67 265]]

TPR = 0.8466453674121406

TNR = 0.22988505747126436

FPR = 0.7701149425287356

FNR = 0.15335463258785942

/usr/local/lib/python3.6/site-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=0.28470023723952187, max_delta_step=0, max_depth=5, min_child_weight=1, missing=None, n_estimators=46, 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)

3 Conclusions

Random forest performance

Using W2vec TPR = 0.91 TNR = 0.22

best n_estimators = 93

Using TF IDF W2vec TPR = 0.93 TNR = 0.08

best n_estimators = 72

GBDT performance

Using W2vec TPR = 0.83 TNR = 0.34

best n_estimators = 93 best max_depth = 4 best eta = 0.5

Using TF IDF W2vec TPR = 0.87 TNR = 0.22

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

W2Vec representations provide better results with both classifiers. GBDT provides better TNR.