RobustTest

September 29, 2019

```
[32]: from sklearn import metrics, ensemble
     from sklearn.model_selection import cross_validate,GridSearchCV,train_test_split
     import xgboost as xgb
     import numpy as np
     import pandas as pd
     import seaborn as sns
     import matplotlib.pyplot as plt
     import matplotlib as mpl
     import warnings
     warnings.filterwarnings('ignore')
     plt.style.use('ggplot')
[33]: print('Reading and Merging Data')
     train = pd.read csv('input/train 1.csv')
     train = train.sample(frac=0.5)
     songs = pd.read csv('input/songs.csv')
     train = pd.merge(train, songs, on='song_id', how='left')
     del songs
     members = pd.read_csv('input/members.csv')
     train = pd.merge(train, members, on='msno', how='left')
     del members
     song_extra_info = pd.read_csv('input/song_extra_info.csv')
     train = pd.merge(train, song extra info, on='song id', how='left')
     del song_extra_info
     print('Reading complete')
     print('Handle Nulls')
     for i in train.select_dtypes(include=['object']).columns:
         train[i][train[i].isnull()] = 'unknown'
     train = train.fillna(value=0)
     print('Flatten the dates')
     train.registration_init_time = pd.to_datetime(train.registration_init_time,_

→format='%Y%m%d', errors='ignore')
     train['registration_init_time_year'] = train['registration_init_time'].dt.year
     train['registration_init_time_month'] = train['registration_init_time'].dt.month
     train['registration_init_time_day'] = train['registration_init_time'].dt.day
```

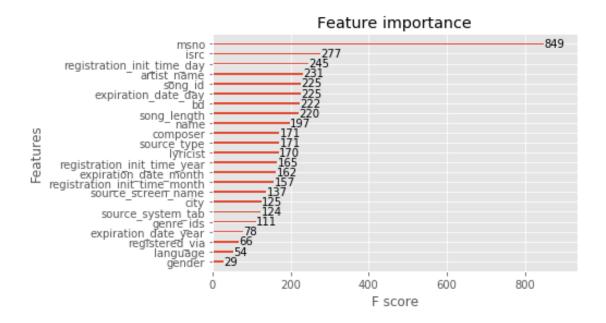
Reading and Merging Data
Reading complete
Handle Nulls
Flatten the dates
Handling the categorical Features
Label Encode the values

```
[97]: print('Training and Calculating Accuracy')
#target = train.pop('target')
train_data, test_data, train_labels, test_labels = train_test_split(train, ustarget, test_size = 0.3)
model = xgb.XGBClassifier(max_depth=3, learning_rate=0.3, n_estimators=300, ustar_jobs=8)
model.fit(train_data, train_labels)
predict_labels = model.predict(test_data)
print(metrics.accuracy_score(test_labels, predict_labels))
```

Training and Calculating Accuracy 0.7370266666666667

```
[66]: xgb.plot_importance(model)
```

[66]: <matplotlib.axes._subplots.AxesSubplot at 0x7f2963d7c860>



```
[98]: graph_to_save = xgb.to_graphviz(model, rankdir='LR')
      graph_to_save.format = 'png'
      graph_to_save.render('xgb_tree')
 [98]: 'xgb_tree.png'
[103]: #=test_data.iloc[0-5]
      predictest=test_data.head(n=5)
[104]: predictest.to_csv('predictest.csv')
      (predictest)
[104]:
                      song_id source_system_tab
                                                    source_screen_name
                                                                         source_type
               msno
      431412 12101
                         8209
                                                0
                                                                     10
                                                                                    5
      1474
               4792
                        41923
                                                6
                                                                      0
                                                                                    0
      81622
                                                 3
                                                                      7
                                                                                    3
               13243
                        28169
                                                 3
                                                                      7
                                                                                    3
      127565
               7959
                        11154
                                                                                    3
      384004
                        70094
                                                 3
                                                                      7
               9401
                                                                                 gender
              song_length genre_ids
                                        artist_name
                                                      composer
                                                                 lyricist
                  326243.0
                                                         21207
      431412
                                   204
                                               14583
                                                                     8046
                                                                                      2
      1474
                  221727.0
                                   204
                                               11979
                                                         23434
                                                                     9664
                                                                                      2
                                   209
                                                                     8735
                                                                                      0
      81622
                  281573.0
                                                3071
                                                         22536
      127565
                  290586.0
                                   209
                                               13073
                                                         27130
                                                                    13715
                                                                                      0
                                   209
      384004
                  262112.0
                                               12786
                                                         17175
                                                                     7229
              registered_via
                                        isrc registration_init_time_year \
                                 name
                               47012 36945
      431412
                                                                       2010
```

1474 81622 127565 384004	7 5270: 9 3749: 9 6171: 3 6344:	3 7481 7 37812		2016 2005 2010 2016
	registration_init_ti	ne_month	registration_i	nit_time_day \
431412		5		29
1474		2		6
81622		7		19
127565		5		23
384004		11		12
	expiration_date_year	expirat	cion_date_month	expiration_date_day
431412	2017		10	1
1474	2018		2	4
81622	2017		9	30
127565	2017		9	26
384004	2016		12	3

[5 rows x 23 columns]

[105]: pred = model.predict(predictest)

[106]: print(pred)

[1 1 1 1 0]