Social network Graph Link Prediction - Facebook Challenge

In [0]:

dtype='object')

```
#Importing Libraries
# please do go through this python notebook:
import warnings
warnings.filterwarnings("ignore")
import csv
import pandas as pd#pandas to create small dataframes
import datetime #Convert to unix time
import time #Convert to unix time
# if numpy is not installed already : pip3 install numpy
import numpy as np #Do aritmetic operations on arrays
# matplotlib: used to plot graphs
import matplotlib
import matplotlib.pylab as plt
import seaborn as sns#Plots
from matplotlib import rcParams#Size of plots
from sklearn.cluster import MiniBatchKMeans, KMeans#Clustering
import math
import pickle
import os
# to install xgboost: pip3 install xgboost
import xgboost as xgb
import warnings
import networkx as nx
import pdb
import pickle
from pandas import HDFStore, DataFrame
from pandas import read hdf
from scipy.sparse.linalg import svds, eigs
import qc
from tqdm import tqdm
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import fl score
In [0]:
#reading
from pandas import read hdf
df final train = read hdf('data/fea sample/storage sample stage4.h5', 'train df', mode='r'
df final test = read hdf('data/fea sample/storage sample stage4.h5', 'test df', mode='r')
In [0]:
df final train.columns
Out[0]:
Index(['source node', 'destination node', 'indicator link',
       'jaccard_followers', 'jaccard_followees', 'cosine followers',
       'cosine_followees', 'num_followers_s', 'num_followees_s',
'num_followees_d', 'inter_followers', 'inter_followees', 'adar_index',
       'follows back', 'same comp', 'shortest path', 'weight in', 'weight out',
       'weight_f1', 'weight_f2', 'weight_f3', 'weight_f4', 'page_rank_s',
       'page_rank_d', 'katz_s', 'katz_d', 'hubs_s', 'hubs_d', 'authorities_s',
       'authorities_d', 'svd_u_s_1', 'svd_u_s_2', 'svd_u_s_3', 'svd_u_s_4',
       'svd_u_s_5', 'svd_u_s_6', 'svd_u_d_1', 'svd_u_d_2', 'svd_u_d_3',
       'svd_u_d_4', 'svd_u_d_5', 'svd_u_d_6', 'svd_v_s_1', 'svd_v_s_2',
       'svd_v_s_3', 'svd_v_s_4', 'svd_v_s_5', 'svd_v_s_6', 'svd_v_d_1',
       'svd v d 2', 'svd v d 3', 'svd v d 4', 'svd v d 5', 'svd v d 6'],
```

In [0]: y_train = df_final_train.indicator_link y_test = df_final_test.indicator_link

In [0]:

```
df_final_train.drop(['source_node', 'destination_node','indicator_link'],axis=1,inplace=T
rue)
df_final_test.drop(['source_node', 'destination_node','indicator_link'],axis=1,inplace=Tr
ue)
```

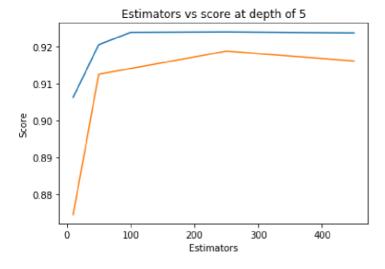
In [0]:

```
estimators = [10, 50, 100, 250, 450]
train scores = []
test scores = []
for i in estimators:
    clf = RandomForestClassifier(bootstrap=True, class weight=None, criterion='gini',
            max_depth=5, max_features='auto', max_leaf_nodes=None,
            min impurity decrease=0.0, min impurity split=None,
            min samples leaf=52, min samples split=120,
            min weight fraction leaf=0.0, n estimators=i, n jobs=-1, random state=25, verb
ose=0, warm start=False)
   clf.fit(df final_train,y_train)
    train_sc = f1_score(y_train,clf.predict(df_final_train))
   test sc = f1 score(y test, clf.predict(df final test))
    test scores.append(test sc)
    train scores.append(train sc)
    print('Estimators = ',i,'Train Score',train sc,'test Score',test sc)
plt.plot(estimators, train scores, label='Train Score')
plt.plot(estimators, test scores, label='Test Score')
plt.xlabel('Estimators')
plt.ylabel('Score')
plt.title('Estimators vs score at depth of 5')
```

Estimators = 10 Train Score 0.9063252121775113 test Score 0.8745605278006858
Estimators = 50 Train Score 0.9205725512208812 test Score 0.9125653355634538
Estimators = 100 Train Score 0.9238690848446947 test Score 0.9141199714153599
Estimators = 250 Train Score 0.9239789348046863 test Score 0.9188007232664732
Estimators = 450 Train Score 0.9237190618658074 test Score 0.9161507685828595

Out[0]:

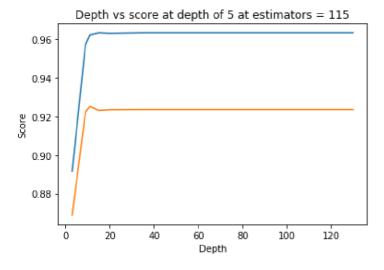
Text(0.5,1,'Estimators vs score at depth of 5')



In [0]:

```
min_impurity_decrease=0.0, min_impurity_split=None,
            min_samples_leaf=52, min_samples_split=120,
            min_weight_fraction_leaf=0.0, n_estimators=115, n_jobs=-1,random_state=25,ve
rbose=0, warm start=False)
   clf.fit(df_final_train,y_train)
    train sc = f1 score(y train, clf.predict(df final train))
    test sc = f1 score(y test, clf.predict(df final test))
    test scores.append(test sc)
    train scores.append(train sc)
    print('depth = ',i,'Train Score',train sc,'test Score',test sc)
plt.plot(depths, train scores, label='Train Score')
plt.plot(depths, test_scores, label='Test Score')
plt.xlabel('Depth')
plt.ylabel('Score')
plt.title('Depth vs score at depth of 5 at estimators = 115')
plt.show()
```

```
depth = 3 Train Score 0.8916120853581238 test Score 0.8687934859875491
depth = 9 Train Score 0.9572226298198419 test Score 0.9222953031452904
depth = 11 Train Score 0.9623451340902863 test Score 0.9252318758281279
depth = 15 Train Score 0.9634267621927706 test Score 0.9231288356496615
depth = 20 Train Score 0.9631629153051491 test Score 0.9235051024711141
depth = 35 Train Score 0.9634333127085721 test Score 0.9235601652753184
depth = 70 Train Score 0.9634333127085721 test Score 0.9235601652753184
depth = 130 Train Score 0.9634333127085721 test Score 0.9235601652753184
```



In [0]:

```
from sklearn.metrics import f1 score
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import f1 score
from sklearn.model selection import RandomizedSearchCV
from scipy.stats import randint as sp randint
from scipy.stats import uniform
param_dist = {"n_estimators":sp_randint(105,125),
              "max depth": sp randint(10,15),
              "min_samples_split": sp_randint(110,190),
              "min_samples_leaf": sp_randint(25,65)}
clf = RandomForestClassifier(random state=25, n jobs=-1)
rf random = RandomizedSearchCV(clf, param distributions=param dist,
                                   n iter=5, cv=10, scoring='f1', random state=25)
rf random.fit(df final train, y train)
print('mean test scores',rf_random.cv_results ['mean_test_score'])
print('mean train scores',rf random.cv results ['mean train score'])
```

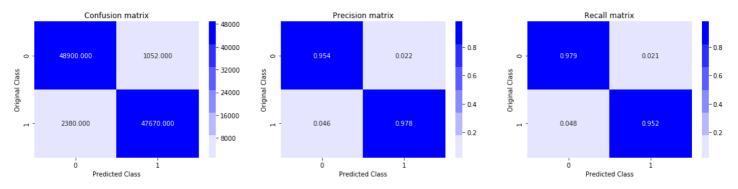
mean test scores [0.96225043 0.96215493 0.96057081 0.96194015 0.96330005] mean train scores [0.96294922 0.96266735 0.96115674 0.96263457 0.96430539]

```
print(rf random.best estimator )
RandomForestClassifier(bootstrap=True, class weight=None, criterion='gini',
            max depth=14, max features='auto', max leaf nodes=None,
            min impurity decrease=0.0, min impurity split=None,
            min_samples_leaf=28, min_samples_split=111,
            min_weight_fraction_leaf=0.0, n_estimators=121, n_jobs=-1,
            oob score=False, random state=25, verbose=0, warm start=False)
In [0]:
clf = RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
            max depth=14, max features='auto', max leaf nodes=None,
            min impurity decrease=0.0, min impurity split=None,
            min samples leaf=28, min samples split=111,
            min weight fraction leaf=0.0, n estimators=121, n jobs=-1,
            oob score=False, random state=25, verbose=0, warm start=False)
In [0]:
clf.fit(df final train, y train)
y_train_pred = clf.predict(df final train)
y test pred = clf.predict(df final test)
In [0]:
from sklearn.metrics import fl score
print('Train f1 score', f1 score(y train, y train pred))
print('Test f1 score', f1 score(y test, y test pred))
Train f1 score 0.9652533106548414
Test f1 score 0.9241678239279553
In [0]:
from sklearn.metrics import confusion matrix
def plot confusion matrix(test y, predict y):
    C = confusion matrix(test y, predict y)
   A = (((C.T) / (C.sum(axis=1))).T)
    B = (C/C.sum(axis=0))
    plt.figure(figsize=(20,4))
    labels = [0,1]
    # representing A in heatmap format
    cmap=sns.light palette("blue")
   plt.subplot(1, 3, 1)
   sns.heatmap(C, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=lab
els)
   plt.xlabel('Predicted Class')
   plt.ylabel('Original Class')
    plt.title("Confusion matrix")
    plt.subplot(1, 3, 2)
    sns.heatmap(B, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=lab
els)
    plt.xlabel('Predicted Class')
    plt.ylabel('Original Class')
   plt.title("Precision matrix")
    plt.subplot(1, 3, 3)
    # representing B in heatmap format
    sns.heatmap(A, annot=True, cmap=cmap, fmt=".3f", xticklabels=labels, yticklabels=lab
els)
   plt.xlabel('Predicted Class')
   plt.ylabel('Original Class')
   plt.title("Recall matrix")
    plt.show()
```

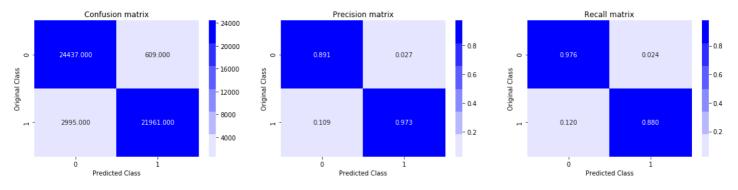
In [0]:

```
print('Train confusion_matrix')
plot_confusion_matrix(y_train,y_train_pred)
print('Test confusion_matrix')
plot_confusion_matrix(y_test,y_test_pred)
```

Train confusion_matrix

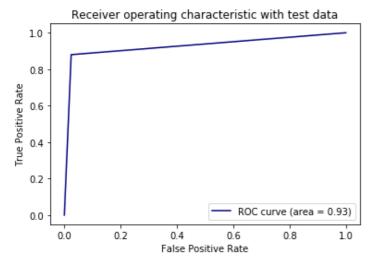


Test confusion_matrix



In [0]:

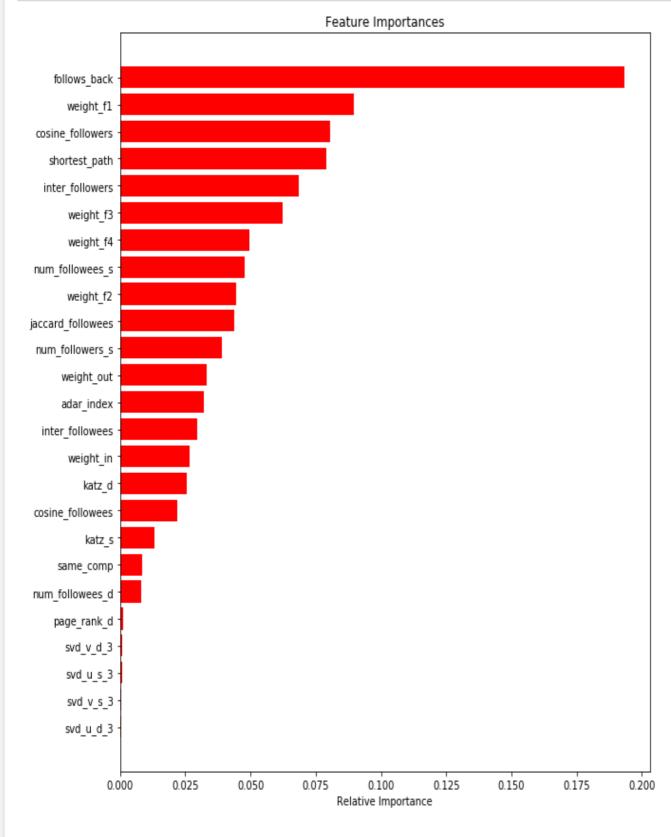
```
from sklearn.metrics import roc_curve, auc
fpr,tpr,ths = roc_curve(y_test,y_test_pred)
auc_sc = auc(fpr, tpr)
plt.plot(fpr, tpr, color='navy',label='ROC curve (area = %0.2f)' % auc_sc)
plt.xlabel('False Positive Rate')
plt.ylabel('True Positive Rate')
plt.title('Receiver operating characteristic with test data')
plt.legend()
plt.show()
```



In [0]:

```
features = df_final_train.columns
importances = clf.feature_importances_
indices = (np.argsort(importances))[-25:]
plt.figure(figsize=(10,12))
plt.title('Feature Importances')
plt.barh(range(len(indices)), importances[indices], color='r', align='center')
```

```
plt.yticks(range(len(indices)), [features[i] for i in indices])
plt.xlabel('Relative Importance')
plt.show()
```



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