Social network Graph Link Prediction - Facebook Challenge

```
In [0]:
#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 gc
\textbf{from } \textbf{tqdm} \textbf{ import} \textbf{ tqdm}
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import fl score
In [0]:
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'],
       dtype='object')
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=True)
df_final_test.drop(['source_node', 'destination_node','indicator_link'],axis=1,inplace=True)
```

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, verbose=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

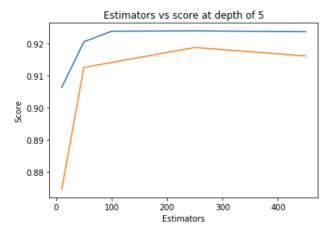
Estimators = 100 Train Score 0.9238690848446947 test Score 0.9141199714153599

Estimators = 250 Train Score 0.9239789348046863 test Score 0.9188007232664732

Estimators = 450 Train Score 0.9237190618658074 test Score 0.9161507685828595

Out[0]:

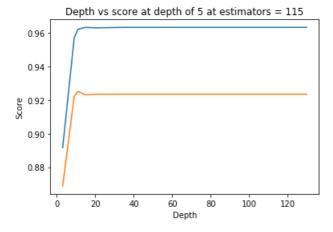
Text(0.5,1,'Estimators vs score at depth of 5')



In [0]:

```
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 = 50 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 fl_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]

In [0]:

```
print(rf_random.best_estimator_)
```

```
In [0]:
```

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 f1_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

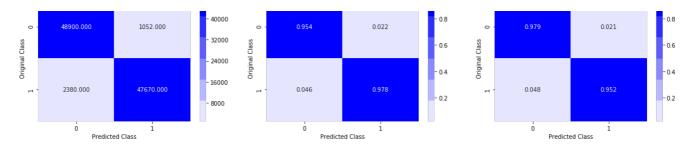
In [0]:

```
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=labels)
    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=labels)
    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=labels)
    plt.xlabel('Predicted Class')
    plt.ylabel('Original Class')
    plt.title("Recall matrix")
    plt.show()
```

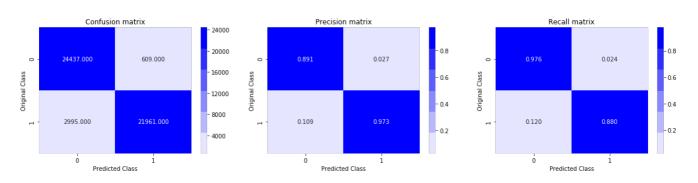
In [0]:

```
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

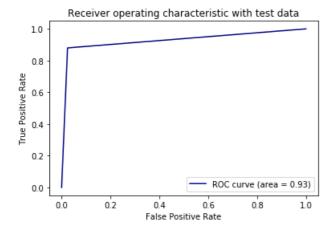


Test confusion matrix



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()
```

