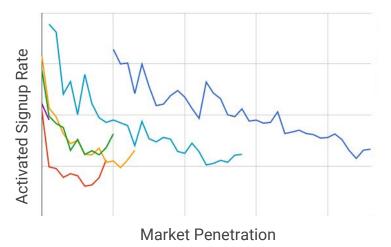
# **Activated Signups Prediction**

### Goal

Identify the model that predicts signup or activated signup\*



- \* Activated Signup : users who was active in their 4th week after signup.
- \* Activated Signup Rate = Activated Signups / Signups

### Data

(102295, 7)

	refer_type	page_type	search_type	country	category	duration	activated_signup
0	GOOGLE		unknown	GB	HAIR_BEAUTY	over_5min	1
2	GOOGLE		unknown	TH	ART	1min_to_3min	0
7	GOOGLE		unknown	US	FOOD_DRINK	1min_to_3min	0
8	GOOGLE		unknown	US	DIY_CRAFTS	over_5min	0
9	DIRECT		unknown	US	WOMENS_FASHION	over_5min	0

Signup sessions data on 7/1/2016 activated\_signup = 1, if the use has actions between 7/22 and 7/28

### Data

	activated_signup
count	102295.000000
mean	0.132255
std	0.338769
min	0.000000
25%	0.000000
50%	0.000000
75%	0.000000
max	1.000000

Conversion Rate = 0.13

```
print SignupData.shape
```

(102295, 268)

# Logistic Regression

#### CV score by changing C

```
0.867745247456 5.60613107681

0.867745247456 12.0417211056

0.867745247456 27.1350998878

0.867745247456 94.7220349312

0.867745248411 197.150823116

0.867618167365 310.803071976

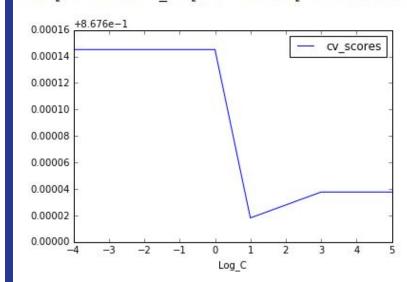
0.867627943491 430.512274027

0.867637718662 541.536948919

0.867637718662 653.438175917

0.867637718662 758.160063028
```

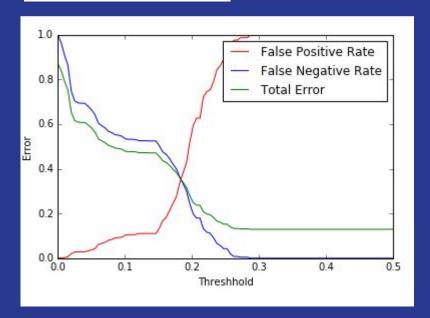
<matplotlib.axes.\_subplots.AxesSubplot at 0x11817a710>



## Threshold

#### **Confusion Matrix**

```
array([[88766, 0], [13527, 2]])
```



#### Threshold = 0.25

[[86219 2547] [12551 978]]

#### Threshold = 0.15

[[46133 42633] [ 2479 11050]]

# Logistic Regression

#### Im = LogisticRegression(C = 1, penalty = "I1")

```
(0.35284489159093213, 'category EDUCATION'),
(0.35751096435273061, 'category DESIGN'),
(0.36389969511080822, 'country RE'),
(0.36502488626054069,
                      'country BO'),
(0.37006237653409335, 'country VE'),
(0.38974819836277375, 'country AO'),
(0.40844718442187061, 'country SD'),
(0.40904374080139561,
                      'country CV'),
(0.46445886518871965,
                      'country OM'),
(0.52056906602024611,
                      'country GY'),
(0.63416031543126816,
                      'country BA'),
(0.64140467799845124,
                      'country AF'),
(0.68384037365509087, 'country FJ'),
(0.77829772847371148, 'country ZW'),
(0.96141049796668021, 'country NC'),
(1.0819972609960393, 'country KG'),
(1.0846816689285714, 'country BI'),
(1.1321740215854936, 'duration 3min to 5min'),
(1.1984547499346265, 'country BJ'),
(2.3207343281713468, 'duration over 5min')
```

```
(-2.0155859736136073, 'duration_30s_to_1min'),
(-1.1738456781808744, 'duration_10s_to_30s'),
(-1.1390789879885599, 'country_MO'),
(-1.0138877999755382, 'country_JO'),
(-0.72538182679051399, 'country_KW')]
```

## Random Forest

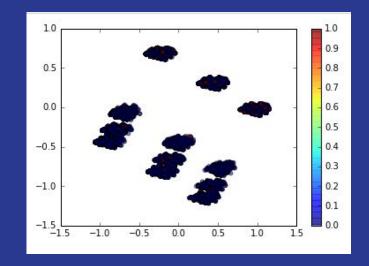
```
Model 2 : Random Forest ###
from sklearn.ensemble import RandomForestClassifier
from sklearn.cross validation import cross val score
RFClass = RandomForestClassifier(n estimators = 1000,
                                 max features = 4, # You
                                 min samples leaf = 10,
                                 oob score = True,
                                 random state = 1,
                                 n jobs = -1)
RFClass.fit(X, y)
print("Out of Bag Accuracy = %f" %RFClass.oob score )
scores = cross_val_score(RFClass, X, y, cv = 10)
print("Cross-validation Accuracy = %f" %scores.mean())
Out of Bag Accuracy = 0.867745
Cross-validation Accuracy = 0.867745
```

## Random Forest

```
Features = range(2,8)
oob score RF = []
for i in Features:
         RFClass = RandomForestClassifier(n estimators = 1000,
                             max features = i,
                                                      #How many fea
                             min samples leaf = 5, #Minimum numb
                              oob score = True,
                             random state = 1,
                              n \text{ jobs} = -1)
        RFClass.fit(X,y)
        oob score RF.append(RFClass.oob score )
plt.plot(Features, oob_score RF)
plt.xlabel("Number of Features")
plt.ylabel("Out of Bag Accuracy")
plt.show()
  0.92
  0.90
Out of Bag Accuracy
  0.88
  0.86
  0.84
  0.82
                      Number of Features
```

## PCA & KNN

0.168942693584



```
from sklearn import neighbors, metrics
misclass = []

for i in range(1, 41, 10):
    clf = PCA(i)
    X_trans = clf.fit_transform(X)
    knn = neighbors.KNeighborsClassifier(n_neighbors = 3, weights='uniform')
    knn.fit(X_trans,y)
    MisClassificationError = 1 - (cross_val_score(knn, X_trans, y, cv=10).mean())
    misclass.append(MisClassificationError)

print(min(misclass))
```