jupyter Mobile Popularity Prediction Project Last Checkpoint: 04/23/2022 (autosaved)

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Run Code

In [1]:
```
#Import all necessary functions and tools.
import pandas as pd
import numpy as np
import sklearn.ensemble
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn import metrics
from sklearn.ensemble import RandomForestClassifier
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import normalize
import scipy.cluster.hierarchy as shc

# rfc = RandomForestClassifier(random_state=0)
# sns.set(style="whitegrid")
```

In [2]:
```
#Read the dataset.
df = pd.read_csv("train.csv")
```

In [3]:
```
#Display a sample of the data.
#df.head(5)
```

In [4]:
```
#Here, I will split the 'installs' column into two columns and discard the 'plus' symbols.
df[['install_nums','plus']] = df.installs.str.split("+",expand=True,)
#df
```

In [5]:
```
#Drop the 'plus' and 'installs' columns.
df.drop(['installs','plus'],axis=1,inplace=True)
```

In [6]:
```
#Display results.
#df.head()
```

In [7]:
```
df.drop(['size'],axis=1,inplace=True)
```

In [8]:
```
#df.head()
```

In [9]:
```
#List all unique values of the 'category' column.
df.category.unique()
```

Out[9]:
```
array(['PERSONALIZATION', 'GAME', 'FAMILY', 'DATING', 'PARENTING',
       'PHOTOGRAPHY', 'TOOLS', 'MEDICAL', 'SOCIAL', 'BOOKS_AND_REFERENCE',
       'FOOD_AND_DRINK', 'AUTO_AND_VEHICLES', 'FINANCE',
       'HEALTH_AND_FITNESS', 'SHOPPING', 'LIFESTYLE', 'EDUCATION',
       'SPORTS', 'NEWS_AND_MAGAZINES', 'COMMUNICATION', 'BUSINESS',
       'PRODUCTIVITY', 'VIDEO_PLAYERS', 'BEAUTY', 'TRAVEL_AND_LOCAL',
       'WEATHER', 'MAPS_AND_NAVIGATION', 'ART_AND_DESIGN', 'EVENTS',
       'COMICS', 'HOUSE_AND_HOME', 'ENTERTAINMENT', 'LIBRARIES_AND_DEMO'],
      dtype=object)
```

In [10]:
```
#Find the number of unique values of the 'category' columns.
numTypes = df.category.unique()

i = 0

for each in numTypes:
    i = i + 1
print(i)
```

```
33
```

In [11]:
```
df.drop(['last_update'],axis=1,inplace=True)
#df.head()
```

In [12]:
```
df.suitable_for.unique()
```

Out[12]:
```
array(['Everyone', 'Teen', 'Mature 17+', 'Everyone 10+',
       'Adults only 18+'], dtype=object)
```

In [13]:
```
#df['category'][0]
```

In [14]:
```
# for each in df['category']:
#     df['category_values'] = np.where(df['category'][each]=='PERSON',True,False)
```

In [15]:
```
#Create a list of our conditions for category conversion.
conditions = [
    (df['category'] == 'PERSONALIZATION'),
    (df['category'] == 'GAME'),
    (df['category'] == 'FAMILY'),
    (df['category'] == 'DATING'),
    (df['category'] == 'PARENTING'),
    (df['category'] == 'PHOTOGRAPHY'),
    (df['category'] == 'TOOLS'),
    (df['category'] == 'MEDICAL'),
    (df['category'] == 'SOCIAL'),
    (df['category'] == 'BOOKS_AND_REFERENCE'),
    (df['category'] == 'FOOD_AND_DRINK'),
    (df['category'] == 'AUTO_AND_VEHICLES'),
    (df['category'] == 'FINANCE'),
    (df['category'] == 'HEALTH_AND_FITNESS'),
    (df['category'] == 'SHOPPING'),
    (df['category'] == 'LIFESTYLE'),
    (df['category'] == 'EDUCATION'),
    (df['category'] == 'SPORTS'),
    (df['category'] == 'NEWS_AND_MAGAZINES'),
    (df['category'] == 'COMMUNICATION'),
    (df['category'] == 'BUSINESS'),
    (df['category'] == 'PRODUCTIVITY'),
    (df['category'] == 'VIDEO_PLAYERS'),
    (df['category'] == 'BEAUTY'),
    (df['category'] == 'TRAVEL_AND_LOCAL'),
    (df['category'] == 'WEATHER'),
    (df['category'] == 'MAPS_AND_NAVIGATION'),
    (df['category'] == 'ART_AND_DESIGN'),
    (df['category'] == 'EVENTS'),
    (df['category'] == 'COMICS'),
    (df['category'] == 'HOUSE_AND_HOME'),
    (df['category'] == 'ENTERTAINMENT'),
    (df['category'] == 'LIBRARIES_AND_DEMO'),
    ]
```

In [16]:
```
values = list(range(1,34))
#print(values)
```

In [17]:
```
df['category_values'] = np.select(conditions, values)
```

In [18]:
```
#df.head(50)
```

In [19]:
```
df.drop(['category'],axis=1,inplace=True)
```

In [20]:
```
#df.head(5)
```

In [21]:
```
df.suitable_for.unique()
```

Out[21]:
```
array(['Everyone', 'Teen', 'Mature 17+', 'Everyone 10+',
       'Adults only 18+'], dtype=object)
```

In [22]:
```
#Create a list of our conditions for category conversion.
conditions_suitable = [
    (df['suitable_for'] == 'Everyone'),
```

```
    (df['suitable_for'] == 'Teen'),
    (df['suitable_for'] == 'Mature 17+'),
    (df['suitable_for'] == 'Everyone 10+'),
    (df['suitable_for'] == 'Adults only 18+')
    ]
```

In [23]:
```
values_suitable = list(range(1,6))
```

In [24]:
```
df['suitable_values'] = np.select(conditions_suitable, values_suitable)
```

In [25]:
```
#df.head()
```

In [26]:
```
#Here, I will split the 'price' column into two columns and discard the 'dollar' symbols.
df[['dollar','price_real']] = df.price.str.split("$",expand=True,)
#df.head()
```

In [27]:
```
df.drop(['price','dollar'],axis=1,inplace=True)
```

In [28]:
```
#df.head(5)
```

In [29]:
```
df.drop(['suitable_for'],axis=1,inplace=True)
```

In [30]:
```
#df.head(5)
```

In [31]:
```
df['prices'] = np.where(df['price_real'] == 'None', '0',df['price_real'])
#df.head()
```

In [32]:
```
df.drop(['price_real','prices'],axis=1,inplace=True)
```

In [33]:
```
#df.head()
```

In [34]:
```
df.drop(['latest_ver'],axis=1,inplace=True)
```

In [35]:
```
#df.head()
```

In [36]:
```
#Create a list of our conditions for category conversion.
conditions_pop = [
    (df['popularity'] == 'High'),
    (df['popularity'] == 'Low')
]
```

In [37]:
```
values_popularity = list(range(0,2))
```

In [38]:
```
df['popularity_val'] = np.select(conditions_pop, values_popularity)
```

In [39]:
```
#df.head(50)
```

In [40]:
```
df.drop(['popularity'],axis=1,inplace=True)
```

In [41]:
```
#df.head()
```

In [ ]:
```

```

In [42]:
```
df.drop(['install_nums'],axis=1,inplace=True)
# = df['install_nums'].astype(int)
```

In [43]:
```
x=df.drop(['app_id','popularity_val'],axis=1)
y=df.popularity_val
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.35, random_state=12340)
model = RandomForestClassifier(n_estimators=20)
model.fit(x_train, y_train)
model.score(x_test, y_test)
y_pred = model.predict(x_test)
cnf_matrix = metrics.confusion_matrix(y_test, y_pred)
print(cnf_matrix)
```

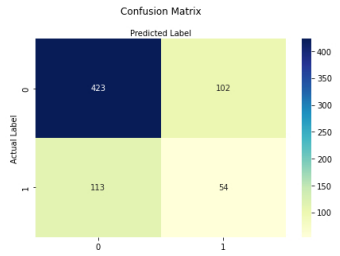
```
[[423 102]
 [113  54]]
```

In [44]:
```
print("Accuracy:",metrics.accuracy_score(y_test, y_pred))
print("Precision:",metrics.precision_score(y_test, y_pred))
print("Recall:",metrics.recall_score(y_test, y_pred))
```

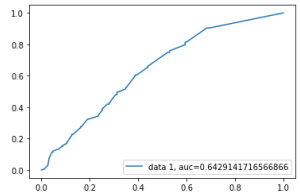
```
Accuracy: 0.6893063583815029
Precision: 0.34615384615384615
Recall: 0.32335329341317365
```

In [45]:
```
class_names=[0,1] # name  of classes
fig, ax = plt.subplots()
tick_marks = np.arange(len(class_names))
plt.xticks(tick_marks, class_names)
plt.yticks(tick_marks, class_names)
# create heatmap
sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap="YlGnBu" ,fmt='g')
ax.xaxis.set_label_position("top")
plt.tight_layout()
plt.title('Confusion Matrix', y=1.1)
plt.ylabel('Actual Label')
plt.xlabel('Predicted Label')
```

Out[45]:
```
Text(0.5, 257.44, 'Predicted Label')
```



In [46]:
```
y_pred_proba = model.predict_proba(x_test)[::,1]
fpr, tpr, _ = metrics.roc_curve(y_test,  y_pred_proba)
auc = metrics.roc_auc_score(y_test, y_pred_proba)
plt.plot(fpr,tpr,label="data 1, auc="+str(auc))
plt.legend(loc=4)
plt.show()
```



In [ ]:
```

```