In [1]:

```
import numpy as np # linear algebra
from numpy import *
import numpy
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import os
import sys
import ast
from sklearn.metrics import accuracy_score,roc_auc_score
from sklearn import metrics
#from catboost import Pool
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
import catboost
from catboost import CatBoostClassifier
print(catboost.__version__)
import seaborn as sns
import statsmodels.formula.api as smfrmla
import statsmodels.api as sm
import warnings
```

0.26

Data reading, informatin checking and visualization

In [2]:

```
tokens_df = pd.read_csv('C:/Users/nasrin/Desktop/AGE PREDICTION/archive/bundles_desc_tokens
tokens_df.info()
tokens_df.head()
```

dtypes: object(3)
memory usage: 9.0+ MB

Out[2]:

	tokens	genre	genres
id			
492765	['king', 'deliveri', 'game', 'awad', 'abu', 's	Games	['Games', 'Racing', 'Casual', 'Entertainment']
687458	['guid', 'jurass', 'winner', 'world', 'tip', '	Books & Reference	
876577	['car', 'photo', 'frame', 'car', 'photo', 'fra	Photography	0
1405997	['short', 'tale', 'toy', 'size', 'room', 'esca	Puzzle	0
64074	['super', 'hero', 'citi', 'rescu', 'crime', 'f	Games	['Games', 'Role Playing', 'Action']

In [3]:

```
desc_df = pd.read_csv('C:/Users/nasrin/Desktop/AGE PREDICTION/archive/bundles_desc.csv', in
desc_df.info()
desc_df.head()
```

memory usage: 4.5+ MB

Out[3]:

description

id	
492765	لعبة سباق سيارات لت\nملك التوصيل - عوض أبو شفة
687458	Kiat untuk panduan dan trik Lego Jurassic Worl
876577	Car photo editor-photo frames is an elite pho
1405997	A Short Tale is a first person point and click
64074	This crime fighter rescue is all about a super

In [4]:

prop_df = pd.read_csv('C:/Users/nasrin/Desktop/AGE PREDICTION/archive/bundles_prop.csv', in prop_df.head()

Out[4]:

	store_os	bundle_released_at	bundle_updated_at	updated_at
id				
492765	ios	2014-06-18 07:00:00.000	2016-07- 13T12:03:29.000+00:00	2021-01-14 06:41:41.706
687458	android	2020-02-15 00:00:00.000	2020-02- 16T07:20:30.000+00:00	2021-01-14 06:41:46.056
876577	android	2017-12-13 00:00:00.000	2020-12- 18T04:19:20.000+00:00	2021-01-14 11:35:25.963
1405997	android	2016-02-10 00:00:00.000	2020-08- 14T15:05:48.000+00:00	2021-01-14 19:28:52.177
64074	ios	2017-10-30 14:18:03.000	2020-05- 28T19:21:48.000+00:00	2021-01-14 06:41:51.201

In [5]:

```
summary_df = pd.read_csv('C:/Users/nasrin/Desktop/AGE PREDICTION/archive/bundles_summary.cs
summary_df.info()
summary_df.head()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 293392 entries, 492765 to 80282
Data columns (total 6 columns):
    Column
               Non-Null Count
                               Dtype
               -----
    -----
    summary
               232372 non-null object
0
1
    rating
               293392 non-null float64
               293392 non-null int64
2
    reviews
               293392 non-null float64
    score
4
    languages 293392 non-null object
5
    is_free
               293392 non-null bool
dtypes: bool(1), float64(2), int64(1), object(2)
```

Out[5]:

memory usage: 13.7+ MB

	summary	rating	reviews	score	languages	is_free
id						
492765	NaN	0.0	2707	4.48356	['EN']	False
687458	Tips for guides and tricks for Lego Jurassic W	407.0	280	2.79000	0	False
876577	Car Photo Editor-Photo Frames with Background	161.0	49	3.66000	0	False
1405997	Decipher clues, solve puzzles, and escape from	171.0	80	4.71000	0	True
64074	NaN	0.0	32	4.03125	['EN']	False

In [6]:

```
bundles_gender= pd.read_csv('C:/Users/nasrin/Desktop/AGE PREDICTION/archive/bundles_gender.
df = tokens_df.join(desc_df).join(prop_df).join(summary_df).join(bundles_gender)
df.info()
df.head()
```

<class 'pandas.core.frame.DataFrame'> Int64Index: 293392 entries, 492765 to 80282 Data columns (total 17 columns):

#	Column	Non-Null Count	Dtype
0	tokens	293392 non-null	object
1	genre	293392 non-null	object
2	genres	293392 non-null	object
3	description	293392 non-null	object
4	store_os	293392 non-null	object
5	<pre>bundle_released_at</pre>	285312 non-null	object
6	<pre>bundle_updated_at</pre>	293392 non-null	object
7	updated_at	293392 non-null	object
8	summary	232372 non-null	object
9	rating	293392 non-null	float64
10	reviews	293392 non-null	int64
11	score	293392 non-null	float64
12	languages	293392 non-null	object
13	is_free	293392 non-null	bool
14	cnt	171109 non-null	float64
15	М	171109 non-null	float64
16	F	171109 non-null	float64
dtyp	es: bool(1), float64	(5), int64(1), ob	ject(10)

dtypes: bool(1),

memory usage: 48.3+ MB

Out[6]:

	tokens	genre	genres	genres description store_os bundle_releas		bundle_released_at	
id							
492765	['king', 'deliveri', 'game', 'awad', 'abu', 's	Games	['Games', 'Racing', 'Casual', 'Entertainment']	عوض أبو	ios	2014-06-18 07:00:00.000	1
687458	['guid', 'jurass', 'winner', 'world', 'tip', '	Books & Reference	0	Kiat untuk panduan dan trik Lego Jurassic Worl	android	2020-02-15 00:00:00.000	1
876577	['car', 'photo', 'frame', 'car', 'photo', 'fra	Photography	0	Car photo editor-photo frames is an elite pho	android	2017-12-13 00:00:00.000	1

	tokens	genre	genres	description	store_os	bundle_released_at	
id							
1405997	['short', 'tale', 'toy', 'size', 'room', 'esca	Puzzle	П	A Short Tale is a first person point and click	android	2016-02-10 00:00:00.000	1
64074	['super', 'hero', 'citi', 'rescu', 'crime', 'f	Games	['Games', 'Role Playing', 'Action']	This crime fighter rescue is all about a super	ios	2017-10-30 14:18:03.000	2

In []:

In [7]:

```
df1=df.drop(columns=['tokens','genres','description','bundle_released_at','bundle_updated_a
print(df1.head())
```

		genre	store_os	rating	reviews	score	is_free	
cnt \								
id								
492765		Games	ios	0.0	2707	4.48356	False	
NaN								
687458	Books & R	eference	android	407.0	280	2.79000	False	
4.0								
876577	Pho	tography	android	161.0	49	3.66000	False	
1.0								
1405997		Puzzle	android	171.0	80	4.71000	True	
NaN								
64074		Games	ios	0.0	32	4.03125	False	2
1.0								
	М	F						
id								
492765	NaN	NaN						
687458	0.000000	1.000000						
876577	1.000000	0.000000						
1405997	NaN	NaN						
64074	0.571429	0.428571						

In [8]:

```
print("Shape-" ,df1.shape)
```

```
In [9]:
```

```
print(df1.info())
<class 'pandas.core.frame.DataFrame'>
Int64Index: 293392 entries, 492765 to 80282
Data columns (total 9 columns):
    Column Non-Null Count
                             Dtype
--- -----
             -----
                             ----
    genre
             293392 non-null object
0
1 store_os 293392 non-null object
2 rating 293392 non-null float64
  reviews 293392 non-null int64
3
4
   score 293392 non-null float64
5 is free 293392 non-null bool
6 cnt
            171109 non-null float64
             171109 non-null float64
7
    F
8
             171109 non-null float64
dtypes: bool(1), float64(5), int64(1), object(2)
memory usage: 30.4+ MB
None
```

In [10]:

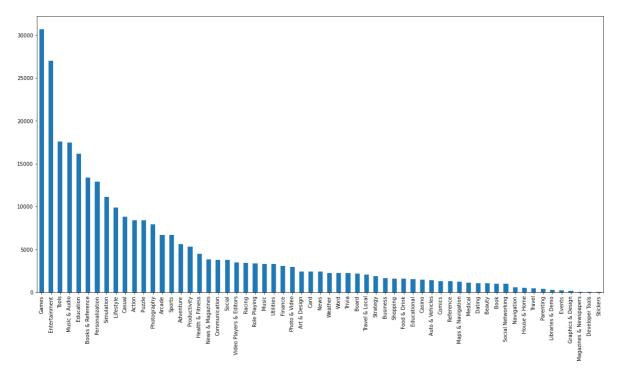
```
print(df1.genre.value_counts())
Games
                           30716
Entertainment
                           27020
Tools
                           17582
Music & Audio
                           17489
Education
                           16176
Events
                            199
Graphics & Design
                             126
Magazines & Newspapers
                               8
Developer Tools
Stickers
Name: genre, Length: 61, dtype: int64
```

Bar graph for no of different geners app:

In [11]:

```
df.head()
print(df.genre.value_counts().plot(kind="bar",figsize=(20, 10)))
```

AxesSubplot(0.125,0.125;0.775x0.755)



Histogram of number of particular gender user for a given range of probability

In [12]:

print(bundles_gender)

	cnt	М	F
id			
26550	79	0.430380	0.569620
22488	236	0.525424	0.474576
203745	6	0.500000	0.500000
101327	5	0.200000	0.800000
354773	1	0.000000	1.000000
• • •			
13899	365	0.430137	0.569863
370068	11	0.636364	0.363636
1385065	2	0.000000	1.000000
1096415	29	0.034483	0.965517
295193	48	0.416667	0.583333

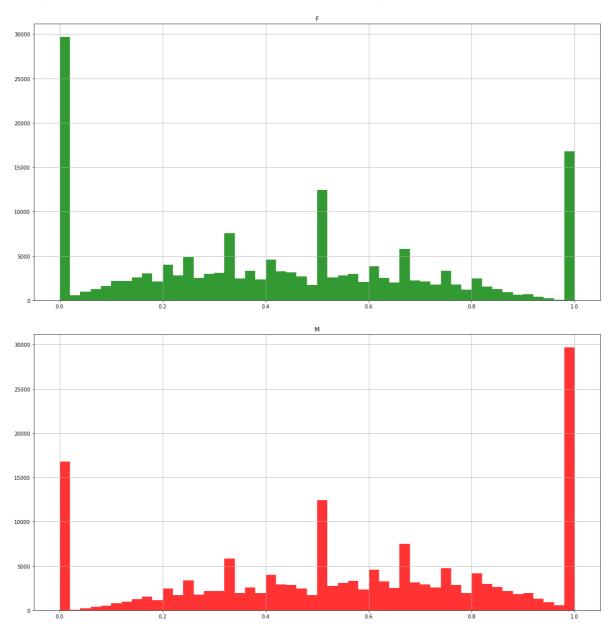
[171272 rows x 3 columns]

In [13]:

```
bundles_gender.hist(column='F',alpha = .8, color= 'g',bins=50,figsize=(20, 10))
bundles_gender.hist(column='M',alpha = 0.8, color= 'r',bins=50,figsize=(20, 10))
```

Out[13]:

array([[<AxesSubplot:title={'center':'M'}>]], dtype=object)



Users data reading, informatin checking and visualization

```
In [14]:
users_df = pd.read_csv('C:/Users/nasrin/Desktop/AGE PREDICTION/archive/users.csv', index_co
print(users_df.head())
                                                                      ids gend
uid
12881473748306291261
                                   [1550,112062,2838,54980,64759,993066]
                                                                             Μ
5871496169617046171
                              [40391,2190,1371978,2023,39200,3516,1634]
                                                                             Μ
13595464671590588595
                      [9728,1314190,979199,2552,1479,1449,976774,131...
                                                                             Μ
12650219932966072351
                                          [4564, 284734, 16370, 3044, 10801]
                                                                             Μ
14238267784075812558
                                                     [6834,4149,1408540]
                                                                             Μ
In [15]:
print("Shape-" ,users_df.shape,"\n\n INFO-" )
print(users_df.info())
Shape- (627761, 2)
INFO-
<class 'pandas.core.frame.DataFrame'>
UInt64Index: 627761 entries, 12881473748306291261 to 7574564252383497711
Data columns (total 2 columns):
     Column Non-Null Count
             627761 non-null object
0
     ids
             627761 non-null object
1
     gend
dtypes: object(2)
memory usage: 14.4+ MB
None
In [16]:
i=len(str.split(users_df.iloc[0,0],','))
print(i)
6
In [17]:
print(users df.gend.value counts())
     418845
Μ
     208916
```

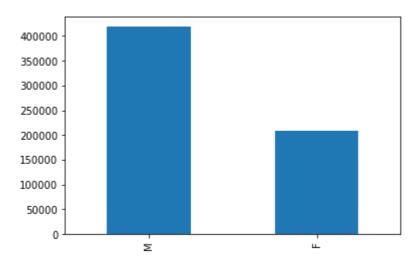
Name: gend, dtype: int64

So, This is a unbalanced dataset.

In [18]:

```
print(users_df.gend.value_counts().plot(kind="bar"))
```

AxesSubplot(0.125,0.125;0.775x0.755)



In [19]:

```
print(users_df.ids.value_counts())
```

```
[1569,2594,5510]
85
[1487,1569,1482]
53
[1468,11166,4149]
39
[1450,1900,1471]
35
[2594,1569,5510]
33
...
[58801,1449,1583,49560,5435,3086,4356,1371991]
1
[8906,1041829,1370225,1408949]
1
[4825,1476,2661,1498,1450,1779,8626,67154,1637,14085,9386,1375247,1372211]
1
[1453,2487,33658,5925]
1
[2371,4356,1450,1533]
1
Name: ids, Length: 619397, dtype: int64
```

In [20]:

```
users_df['apps_count'] = users_df['ids'].str.split(',').apply(len)
users_df.groupby('gend')['apps_count'].describe()
```

Out[20]:

	count	mean	std	min	25%	50%	75%	max
gend								
F	208916.0	15.166067	21.847228	3.0	5.0	8.0	16.0	685.0
М	418845.0	12.445571	17.637542	3.0	5.0	7.0	13.0	772.0

So, average no of apps used by women more than the male.

In [21]:

```
bundles_gender[(bundles_gender['F']>=0.3325) & (bundles_gender['F']<=0.3375)].describe()</pre>
```

Out[21]:

	cnt	М	F
count	6151.000000	6151.000000	6151.000000
mean	171.156235	0.666571	0.333429
std	2760.011403	0.000529	0.000529
min	3.000000	0.662500	0.332512
25%	3.000000	0.666667	0.333333
50%	3.000000	0.666667	0.333333
75%	9.000000	0.666667	0.333333
max	113392.000000	0.667488	0.337500

Using mean of probability of each installed app

#Accuracy and AUC based on probability value given

```
Not working code:
g_dict = bundles_gender['F'].to_dict()
users_df['F_prob'] = users_df['ids'].apply(
    lambda x: np.mean(
        list(filter(None.__ne__, list(map(g_dict.get, x))))
    )
)
Output:RED ALART
    C:\Users\nasrin\anaconda3\lib\site-packages\numpy\core\fromnumeric.py:3372:
RuntimeWarning: Mean of empty slice.
    return _methods._mean(a, axis=axis, dtype=dtype,
```

```
In [22]:
```

```
g_dict = bundles_gender['F'].to_dict()
```

In [23]:

```
def meanprob(a):
   summ =0
   mean=0
   i=0
   count=0
   count2=0
   for i in range (len(a)):
       try:
            summ = summ + g_dict.get(int(str(a[i])))
        except (RuntimeError, TypeError, NameError ,ZeroDivisionError,IndexError ):
            count= count+1
            pass
   try:
        mean = summ / (len(a)-count)
        return mean
   except (RuntimeError, TypeError, NameError, ZeroDivisionError, IndexError):
        count2=count2 +1
        return ("None")
        pass
   if count2>0:
        print("No of missing value",count2)
```

In [24]:

```
F_prob=[]
for k in range (0,6):
    p=k*100000
    q=(k+1)*100000

for j in range (p,q):
        b=list((users_df.iloc[j][0])[1:-1].split(","))
        F_prob.append(meanprob(b))

for j in range (600000,627761):
    b=list((users_df.iloc[j][0])[1:-1].split(","))
    F_prob.append(meanprob(b))
users_df['F_prob'] = np.array(F_prob)
```

In [25]:

```
print(users_df)
```

```
ids gend
\
uid
12881473748306291261
                                   [1550,112062,2838,54980,64759,993066]
                                                                              Μ
5871496169617046171
                               [40391,2190,1371978,2023,39200,3516,1634]
                                                                              Μ
13595464671590588595
                      [9728,1314190,979199,2552,1479,1449,976774,131...
                                                                              Μ
12650219932966072351
                                          [4564, 284734, 16370, 3044, 10801]
                                                                              Μ
14238267784075812558
                                                      [6834,4149,1408540]
                                                                              Μ
2105809415949843792
                                                  [1528, 35250, 15078, 2439]
                                                                              F
12814429133630927125
                                                        [17776,1550,6562]
                                                                              Μ
13688574997457355644
                                                      [1449,165261,10768]
                                                                              Μ
5830984436669696438
                       [1900, 1374218, 17248, 6841, 1860, 6787, 1412486, 147...
                                                                              F
7574564252383497711
                       [1511,42869,1833,10183,152191,16036,27304,1077...
                                                                              Μ
                       apps_count
                                     F_prob
uid
12881473748306291261
                                6 0.238108
5871496169617046171
                                7
                                   0.392768
13595464671590588595
                               39 0.581965
12650219932966072351
                                5 0.435303
14238267784075812558
                                3
                                   0.345937
. . .
                              . . .
2105809415949843792
                                4 0.614698
12814429133630927125
                                3 0.141977
13688574997457355644
                                3 0.186081
5830984436669696438
                               13 0.755751
7574564252383497711
                               39 0.244649
```

[627761 rows x 4 columns]

In [26]:

```
print(f"Accuracy: {accuracy_score(users_df['gend'].astype('category').cat.codes, users_df['
```

Accuracy: 0.740925288445762

In [27]:
<pre>print(f"AUC: {1 - roc_auc_score(users_df['gend'].astype('category').cat.codes, users_df['F_</pre>
AUC: 0.7793767184317941
In [28]:
<pre>np.corrcoef(users_df['F_prob'],users_df['gend'].astype('category').cat.codes)[0,1]</pre>
Out[28]:
-0.46602945129982837
Logistic Regression:
T. [].
In []:
<pre>In []:</pre>