

Game_Analysis

March 15, 2018

Observed Trends: 1. Males not only the make up over 80% of the players of this game, the are also responsible for over 80% of the revenue. 2. Out of the 573 players, each player has spend under 20 dollars on items. 3. The 20-24 age bracket more revenue than any other bracket, but the 25-29 age group, on average, purchases more expensive items. 4. The Retribution Axe not only appears in the most popular item list, it is priced almost 2 dollars more than other popular items on that list and is the item that has generated the most revenue. 5. Other than the Retribution Axe, the most popular items list generates items that are priced below the average purchahse price.

NOTE: 1. The age bins are divided in 10 instead of 4, so that data trends can be analyzed more accurately. 2. To show the normalized total, i have used 0-1 scale (feature scaling method) because when it comes to real time data visualization and to visualize the data where the difference of the value could be 1000 to 20000 then 0-1 scale would be more scalable than the regular normalization. Also prepared supporting documentation from google search.

```
In [1]: import pandas as pd
import numpy as np
import csv
```

```
In [2]: # Reading input file to dataframe
filepath = ("purchase_data.json")
df = pd.read_json(filepath)
#df.to_csv("mainjson output.csv", sep=',', encoding='utf-8') - added this step to ana
#df.head()
game_df = pd.DataFrame(df)
game_df.head()
```

```
Out[2]:
```

	Age	Gender	Item ID	Item Name	Price	\
0	38	Male	165	Bone Crushing Silver Skewer	3.37	
1	21	Male	119	Stormbringer, Dark Blade of Ending Misery	2.32	
2	34	Male	174	Primitive Blade	2.46	
3	21	Male	92	Final Critic	1.36	
4	23	Male	63	Stormfury Mace	1.27	

	SN
0	Aelalis34
1	Eolo46
2	Assastnya25
3	Pheusrical25
4	Aela59

```
In [3]: #Total Players Count
players_count = len(game_df["SN"].unique())
total_players_pd = pd.DataFrame({"Total Players Count": [players_count]})
total_players_pd
```

```
Out[3]:      Total Players Count
0                573
```

```
In [4]: #Purchasing Analysis
purchases_unique = len(game_df["Item ID"].unique())
purchases_total = game_df["Price"].count()
purchases_revenue = game_df["Price"].sum()
purchases_average = round(game_df["Price"].sum()/purchases_total,2)

purc_analysis_pd = pd.DataFrame({"Number of Unique Items":purchases_unique,"Average Purchase Price":purchases_average,
                                "Total Number of Purchases":purchases_total, "Total Revenue":purchases_revenue,
                                columns=["Number of Unique Items","Average Purchase Price","Total Number of Purchases","Total Revenue"])

purc_analysis_pd["Average Purchase Price"] = purc_analysis_pd["Average Purchase Price"].round(2)
purc_analysis_pd["Total Revenue"] = purc_analysis_pd["Total Revenue"].map("${:.2f}".format)
purc_analysis_pd
```

```
Out[4]:      Number of Unique Items  Average Purchase Price  Total Number of Purchases  \
0                183                $2.93                780

      Total Revenue
0      $2286.33
```

```
In [5]: #Gender Demographics
gender_df = game_df.drop_duplicates(subset=['SN'],keep='first')
gender_df = gender_df.groupby("Gender")
gender_count = gender_df["Age"].count()
gender_percentage = round((gender_count/players_count)*100,2)
gender_analysis = pd.DataFrame({"Players count":gender_count,"Percentage of Players":gender_percentage})
gender_analysis.sort_values(["Players count"],ascending = False)
gender_analysis["Percentage of Players"] = gender_analysis["Percentage of Players"].map("${:.2f}%".format)
gender_analysis
```

```
Out[5]:      Percentage of Players  Players count
Gender
Female                17.45%                100
Male                 81.15%                465
Other / Non-Disclosed    1.40%                 8
```

```
In [6]: #Purchasing Analysis - Gender based
purc_gender_df = game_df.groupby("Gender")
gender_purchases_count = purc_gender_df["Price"].count()
gender_purchases_average = round(purc_gender_df["Price"].mean(),2)
gender_purchases_total = purc_gender_df["Price"].sum()
```

```

gender_purchases_max = gender_purchases_total.max()
gender_purchases_min = gender_purchases_total.min()
gender_normalize_total = (gender_purchases_total - gender_purchases_min)/(gender_purchases_max - gender_purchases_min)
gender_purchase_analysis = pd.DataFrame({"Total Gender Purchases":gender_purchases_total,
                                         "Total Purchase Value": gender_purchases_total,
                                         "Average Gender Purchases": gender_normalize_total,
                                         columns=["Total Gender Purchases","Average Gender Purchases"]})
gender_purchase_analysis

```

```

Out [6]:

```

	Total Gender Purchases	Average Gender Purchases
Gender		
Female	136	2.82
Male	633	2.95
Other / Non-Disclosed	11	3.25

	Total Purchase Value	Normalized Totals
Gender		
Female	382.91	0.189509
Male	1867.68	1.000000
Other / Non-Disclosed	35.74	0.000000

```

In [7]: #Age Analysis
bins = [0, 9, 14, 19, 24, 29, 34, 39, 100]
range_names = ['< 10', '10 - 14', '15 -19', '20 - 24', '25 - 29', '30 - 34', '35 - 39', '40+']
game_new_df = game_df
game_new_df["Age Range"] = pd.cut(game_new_df["Age"], bins, labels=range_names)
age_df = game_new_df.drop_duplicates(subset=['SN'],keep='first')
age_analysis_df = age_df.groupby("Age Range")
age_players_count = age_analysis_df["Age"].count()
age_percentage_players = round((age_analysis_df["Age"].count()/players_count)*100,2)
age_analysis_pd = pd.DataFrame({"Players Count":age_players_count,
                               "Percentage of Players": age_percentage_players})
age_analysis_pd["Percentage of Players"] = age_analysis_pd["Percentage of Players"].map(lambda x: f'{x:.2f}%')
age_analysis_pd

```

```

Out [7]:

```

Age Range	Percentage of Players	Players Count
< 10	3.32%	19
10 - 14	4.01%	23
15 -19	17.45%	100
20 - 24	45.20%	259
25 - 29	15.18%	87
30 - 34	8.20%	47
35 - 39	4.71%	27
40+	1.92%	11

```

In [8]: #Age Purchase Analysis
age_purc_df = game_new_df.groupby("Age Range")
age_purchases_count = age_purc_df["Price"].count()
age_purchases_average = round(age_purc_df["Price"].mean(),2)

```

```

age_purchases_total = age_purc_df["Price"].sum()
age_purchase_max = age_purchases_total.max()
age_purchase_min = age_purchases_total.min()
age_normalize_total = (age_purchases_total-age_purchase_min)/(age_purchase_max - age_p
age_purchase_analysis_pd = pd.DataFrame({"Purchase Count":age_purchases_count,"Average
                                "Total Purchase Value": age_purchases_total, "Normaliz
                                columns =["Purchase Count","Average Purchase Price","T

age_purchase_analysis_pd["Average Purchase Price"] = age_purchase_analysis_pd["Average
age_purchase_analysis_pd["Total Purchase Value"] = age_purchase_analysis_pd["Total Pur
age_purchase_analysis_pd

```

Out[8]:

	Purchase Count	Average Purchase Price	Total Purchase Value \
--	----------------	------------------------	------------------------

Age Range	Purchase Count	Average Purchase Price	Total Purchase Value \
< 10	28	\$2.98	\$83.46
10 - 14	35	\$2.77	\$96.95
15 -19	133	\$2.91	\$386.42
20 - 24	336	\$2.91	\$978.77
25 - 29	125	\$2.96	\$370.33
30 - 34	64	\$3.08	\$197.25
35 - 39	42	\$2.84	\$119.40
40+	17	\$3.16	\$53.75

	Normalize Total
Age Range	
< 10	0.032118
10 - 14	0.046702
15 -19	0.359635
20 - 24	1.000000
25 - 29	0.342241
30 - 34	0.155132
35 - 39	0.070971
40+	0.000000

In [9]: *#Top 5 Spenders*

```

spenders_df = game_df.groupby("SN")
spenders_count = spenders_df["Price"].count()
spenders_average_price = spenders_df["Price"].mean()
spenders_total = spenders_df["Price"].sum()

spender_analysis_pd = pd.DataFrame({"Purchase Count":spenders_count,"Average Purchase I
                                "Total Purchase Value": spenders_total},
                                columns =["Purchase Count","Average Purchase Price","T

total_spender_analysis_pd = spender_analysis_pd.sort_values("Total Purchase Value", as
top_spender_analysis_pd = total_spender_analysis_pd.head(5)
top_spender_analysis_pd["Average Purchase Price"] = top_spender_analysis_pd["Average P
top_spender_analysis_pd["Total Purchase Value"] = top_spender_analysis_pd["Total Purcha

```

```
top_spender_analysis_pd
```

```
C:\Users\santo\Anaconda3\lib\site-packages\ipykernel_launcher.py:13: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

```
See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html
del sys.path[0]
```

```
C:\Users\santo\Anaconda3\lib\site-packages\ipykernel_launcher.py:14: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

```
See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html
```

```
Out[9]:
```

	Purchase Count	Average Purchase Price	Total Purchase Value
SN			
Undirrala66	5	\$3.41	\$17.06
Saedue76	4	\$3.39	\$13.56
Mindimnya67	4	\$3.18	\$12.74
Haellysu29	3	\$4.24	\$12.73
Eoda93	3	\$3.86	\$11.58

```
In [10]: #Most Popular Items
```

```
item_df = game_df.groupby(["Item ID", "Item Name"])
item_count = item_df["Price"].count()
item_total = item_df["Price"].sum()
item_price = item_df["Price"].unique()
item_analysis_pd = pd.DataFrame({"Purchase Count": item_count, "Purchase Price": item_price})
item_price_pd = item_analysis_pd["Purchase Price"].values.astype(float)
item_analysis_pd["Price"] = item_price_pd
new_item_analysis_pd = item_analysis_pd[["Purchase Count", "Price", "Total Purchase Value"]]
total_item_analysis_pd = new_item_analysis_pd.sort_values("Purchase Count", ascending=False)
top_item_analysis_pd = total_item_analysis_pd.head(6)

top_item_analysis_pd["Price"] = top_item_analysis_pd["Price"].map("${:.2f}".format)
top_item_analysis_pd["Total Purchase Value"] = top_item_analysis_pd["Total Purchase Value"].map("${:.2f}".format)

top_item_analysis_pd
```

```
C:\Users\santo\Anaconda3\lib\site-packages\ipykernel_launcher.py:13: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
```

```
See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html
del sys.path[0]
```

```
C:\Users\santo\Anaconda3\lib\site-packages\ipykernel_launcher.py:14: SettingWithCopyWarning:
```

A value is trying to be set on a copy of a slice from a DataFrame.
Try using `.loc[row_indexer,col_indexer] = value` instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>

```
Out[10]:
```

		Purchase Count	Price \
Item ID	Item Name		
39	Betrayal, Whisper of Grieving Widows	11	\$2.35
84	Arcane Gem	11	\$2.23
31	Trickster	9	\$2.07
175	Woeful Adamantite Claymore	9	\$1.24
13	Serenity	9	\$1.49
34	Retribution Axe	9	\$4.14

		Total Purchase Value
Item ID	Item Name	
39	Betrayal, Whisper of Grieving Widows	\$25.85
84	Arcane Gem	\$24.53
31	Trickster	\$18.63
175	Woeful Adamantite Claymore	\$11.16
13	Serenity	\$13.41
34	Retribution Axe	\$37.26

```
In [11]: #Most Profitable Items
total_item_analysis_purc_pd = total_item_analysis_pd.sort_values("Total Purchase Value")
top_item_analysis_purc_pd = total_item_analysis_purc_pd.head(5)
top_item_analysis_purc_pd["Price"] = top_item_analysis_purc_pd["Price"].map("${:.2f}")
top_item_analysis_purc_pd["Total Purchase Value"] = top_item_analysis_purc_pd["Total Purchase Value"].map("${:.2f}")
top_item_analysis_purc_pd
```

C:\Users\santo\Anaconda3\lib\site-packages\ipykernel_launcher.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using `.loc[row_indexer,col_indexer] = value` instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>
after removing the cwd from sys.path.

C:\Users\santo\Anaconda3\lib\site-packages\ipykernel_launcher.py:5: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using `.loc[row_indexer,col_indexer] = value` instead

See the caveats in the documentation: <http://pandas.pydata.org/pandas-docs/stable/indexing.html>
"""

```
Out[11]:
```

		Purchase Count	Price	Total Purchase Value
Item ID	Item Name			
34	Retribution Axe	9	\$4.14	\$37.26

115	Spectral Diamond Doomblade	7	\$4.25	\$29.75
32	Orenmir	6	\$4.95	\$29.70
103	Singed Scalpel	6	\$4.87	\$29.22
107	Splitter, Foe Of Subtlety	8	\$3.61	\$28.88

In []: