### **Heroes Of Pymoli Data Analysis**

- Of the 1163 active players, the vast majority are male (84%). There also exists, a smaller, but notable proportion of female players (14%).
- Our peak age demographic falls between 20-24 (44.8%) with secondary groups falling between 15-19 (18.60%) and 25-29 (13.4%).

### **Note**

• Instructions have been included for each segment. You do not have to follow them exactly, but they are included to help you think through the steps.

```
In [52]: # Dependencies and Setup
import pandas as pd
import numpy as np

# Raw data file
file_to_load = "Resources/purchase_data.csv"

# Read purchasing file and store into pandas data frame
purchase_data = pd.read_csv(file_to_load)
purchase_data.head()
```

### Out[52]:

	Purchase ID	SN	Age	Gender	Item ID	Item Name	Price
0	0	Lisim78	20	Male	108	Extraction, Quickblade Of Trembling Hands	3.53
1	1	Lisovynya38	40	Male	143	Frenzied Scimitar	1.56
2	2	Ithergue48	24	Male	92	Final Critic	4.88
3	3	Chamassasya86	24	Male	100	Blindscythe	3.27
4	4	Iskosia90	23	Male	131	Fury	1.44

## ## Player Count

```
* Display the total number of players
```

```
In [31]: players=purchase_data["SN"].unique()
    total_players=len(players)
    print("total players:"+str(total_players))
```

total players:576

# ## Purchasing Analysis (Total)

```
* Run basic calculations to obtain number of unique items, average price, etc.

* Create a summary data frame to hold the results

* Optional: give the displayed data cleaner formatting

* Display the summary data frame
```

#### Out[54]:

# Number of Unique Items Average Price Number of Purchases Total Revenue 183 \$3.05 780 \$2379.77

### ## Gender Demographics

```
* Run basic calculations to obtain number of unique items, average price, etc.

* Create a summary data frame to hold the results

* Optional: give the displayed data cleaner formatting

* Display the summary data frame
```

#### Out[196]:

		Iotal Count	Percentage of Players
0	Male	652	1.13%
1	Female	113	0.20%
2	Others/Non-disclosed	15	0.03%

# **Purchasing Analysis (Gender)**

- Run basic calculations to obtain purchase count, avg. purchase price, etc. by gender
- For normalized purchasing, divide total purchase value by purchase count, by gender
- Create a summary data frame to hold the results
- Optional: give the displayed data cleaner formatting
- · Display the summary data frame

```
In [203]:
```

### Out[203]:

	Purchase Count	Average Purchase Price	Total Purchase Value	Normalized Totals
Gender				
Female	113	\$3.20	\$361.94	\$3.20
Male	652	\$3.02	\$1967.64	\$3.02
Other / Non- Disclosed	15	\$3.35	\$50.19	\$3.35

## **Age Demographics**

```
* Establish bins for ages

* Categorize the existing players using the age bins. Hint: use pd.cut()

* Calculate the numbers and percentages by age group

* Create a summary data frame to hold the results

* Optional: round the percentage column to two decimal points

* Display Age Demographics Table
```

#### Out[110]:

#### Percentage of Players Total Count

age_bin		
<10	3.99	23
10-14	4.86	28
15-19	23.61	136
20-24	63.37	365
25-29	17.53	101
30-34	12.67	73
35-39	7.12	41
40+	2.26	13

# ## Purchasing Analysis (Age)

```
* Bin the purchase_data data frame by age

* Run basic calculations to obtain purchase count, avg. purchase price, etc. in the table below

* Calculate Normalized Purchasing

* Create a summary data frame to hold the results

* Optional: give the displayed data cleaner formatting

* Display the summary data frame
```

#### Out[116]:

age_bin				
<10	23	\$3.35	\$77.13	\$3.35
10-14	28	\$2.96	\$82.78	\$2.96
15-19	136	\$3.04	\$412.89	\$3.04
20-24	365	\$3.05	\$1114.06	\$3.05
25-29	101	\$2.90	\$293.00	\$2.90
30-34	73	\$2.93	\$214.00	\$2.93
35-39	41	\$3.60	\$147.67	\$3.60
40+	13	\$2.94	\$38.24	\$2.94

### **Top Spenders**

- \* Run basic calculations to obtain the results in the table below
- \* Create a summary data frame to hold the results
- \* Sort the total purchase value column in descending order
- \* Optional: give the displayed data cleaner formatting
- \* Display a preview of the summary data frame

#### Out[129]:

	Purchase Count	Average Purchase Price	Total Purchase Value
SN			
Lisosia93	5	3.792000	18.96
Idastidru52	4	3.862500	15.45
Chamjask73	3	4.610000	13.83
Iral74	4	3.405000	13.62
Iskadarya95	3	4.366667	13.10

# **Most Popular Items**

- Retrieve the Item ID, Item Name, and Item Price columns
- Group by Item ID and Item Name. Perform calculations to obtain purchase count, item price, and total purchase value
- Create a summary data frame to hold the results
- Sort the purchase count column in descending order
- Optional: give the displayed data cleaner formatting
- Display a preview of the summary data frame

#### Out[160]:

		Purchase Count	Item Price	Total Purchase Value
Item ID	Item Name			
178	Oathbreaker, Last Hope of the Breaking Storm	12	\$4.23	\$50.76
145	Fiery Glass Crusader	9	\$4.58	\$41.22
108	Extraction, Quickblade Of Trembling Hands	9	\$3.53	\$31.77
82	Nirvana	9	\$4.90	\$44.10
19	Pursuit, Cudgel of Necromancy	8	\$1.02	\$8.16

### **Most Profitable Items**

- Sort the above table by total purchase value in descending order
- Optional: give the displayed data cleaner formatting
- Display a preview of the data frame

In [161]: sorted\_item\_prof=df\_grouped\_byItem.sort\_values(by='Total Purchase Value', as
 sorted\_item\_prof["Item Price"] = sorted\_item\_prof["Item Price"].map("\${:.2f}]
 sorted\_item\_prof["Total Purchase Value"] = sorted\_item\_prof["Total Purchase
 sorted\_item\_prof.head()

### Out[161]:

		Purchase Count	Item Price	Total Purchase Value
Item ID	Item Name			
178	Oathbreaker, Last Hope of the Breaking Storm	12	\$4.23	\$50.76
82	Nirvana	9	\$4.90	\$44.10
145	Fiery Glass Crusader	9	\$4.58	\$41.22
92	Final Critic	8	\$4.88	\$39.04
103	Singed Scalpel	8	\$4.35	\$34.80

In [ ]: