

In [13]:

```
# Heroes of Pymoli
# By: Jack Cohen

# Import Dependencies and Setup
import pandas as pd

# File to Load
file = 'Resources/purchase_data.csv'

# Read Purchasing File and store into Pandas data frame
purchase_data = pd.read_csv(file)
```

In [14]:

```
# Player Count
player_list = purchase_data.SN.unique()
total_num_players = len(player_list)
print(f"There are {total_num_players} total players.")
player = pd.DataFrame([{'Total Players':total_num_players}])
player
```

There are 576 total players.

Out[14]:

Total Players	
0	576

In [3]:

```
# Purchasing Analysis
num_unique_items = purchase_data["Item ID"].value_counts().count()
genders = purchase_data.Gender.unique()
avg_price = round(purchase_data["Price"].mean(),2)
num_purchases = purchase_data["SN"].count()
total_rev = purchase_data["Price"].sum()
data_summary = pd.DataFrame([{" ": "Value", "Number of Unique Items":num_unique_items,
                              "Average Price ($)":avg_price,
                              "Number of Purchases":num_purchases,
                              "Total Revenue ($)":total_rev}])
data_summary.style.format({"Average Price ($)":'${0:,.2f}', "Total Revenue ($)":'${0:,.2f}'})
.hide_index()
```

Out[3]:

	Number of Unique Items	Average Price (\$)	Number of Purchases	Total Revenue (\$)
Value	179	\$3.05	780	\$2,379.77

In [4]:

```
# Gender Demographics
no_dups = purchase_data.drop_duplicates(subset=["SN"])
SN_Gender = no_dups[["SN", "Gender"]]
SN_Gender = SN_Gender.groupby("Gender").count()
percent_gender = [100*SN_Gender.iloc[:,0][0]/total_num_players,
                  100*SN_Gender.iloc[:,0][1]/total_num_players,
                  100*SN_Gender.iloc[:,0][2]/total_num_players]
SN_Gender.insert(1, "Percentage of Players", percent_gender)
SN_Gender.rename(columns={"SN": "Total Count"}, inplace=True)
```

In [5]:

```
SN_Gender.style.format({"Percentage of Players": "{0:,.2f}%"})
```

Out[5]:

	Total Count	Percentage of Players
Gender		
Female	81	14.06%
Male	484	84.03%
Other / Non-Disclosed	11	1.91%

In [6]:

```
# Purchasing Analysis (Gender)
all_genders = purchase_data.Gender.unique()
gender_df = purchase_data.groupby("Gender")
purchase_count = gender_df['Gender'].count()
average_price = gender_df.mean()["Price"]
total_purchase_val = purchase_count*average_price
avg_tot_purchase_per_person = total_purchase_val/SN_Gender["Total Count"]
```

In [7]:

```
gender_summary = pd.DataFrame({
    "Purchase Count":purchase_count,
    "Average Purchase Price":list(average_price),
    "Total Purchase Value":list(total_purchase_val),
    "Avg Total Purchase per Person":list(avg_tot_purchase_per_per
son)
    })
gender_summary = gender_summary.style.format({"Average Purchase Price":'${0:,.2f}',
    "Total Purchase Value":'${0:,.2f}',
    "Avg Total Purchase per Person":'${0:,.2f}'})
gender_summary
```

Out[7]:

	Purchase Count	Average Purchase Price	Total Purchase Value	Avg Total Purchase per Person
Gender				
Female	113	\$3.20	\$361.94	\$4.47
Male	652	\$3.02	\$1,967.64	\$4.07
Other / Non-Disclosed	15	\$3.35	\$50.19	\$4.56

In [8]:

```
# Age Demographics
age_bins = [0,9,14,19,24,29,34,39,200]
age_labels = [ '<10', '10-14', '15-19', '20-24', '25-29', '30-34', '35-39', '40+' ]
age_sorted = pd.cut(no_dups['Age'],age_bins, labels=age_labels).value_counts()
percent_age = 100*age_sorted[age_labels]/total_num_players
age_summary = pd.DataFrame({"":age_labels,
                            "Total Count":list(age_sorted[age_labels]),
                            "Percentage of Players":list(percent_age)
                            })
age_summary = age_summary.style.format({"Percentage of Players":'{0:,.2f}%'}).hide_index()
age_summary
```

Out[8]:

	Total Count	Percentage of Players
<10	17	2.95%
10-14	22	3.82%
15-19	107	18.58%
20-24	258	44.79%
25-29	77	13.37%
30-34	52	9.03%
35-39	31	5.38%
40+	12	2.08%

In [9]:

```
# Purchasing Analysis (Age)
age_analysis = purchase_data
age_analysis = age_analysis.drop(columns=["SN", "Gender", "Item ID", "Item Name", "Purchase ID"
])
age_analysis["Age Range"] = pd.cut(purchase_data['Age'],age_bins, labels=age_labels)
age_sorts = age_analysis.groupby("Age Range")
purchase_count_age = age_sorts["Age"].count()
avg_purch_price_age = age_sorts["Price"].mean()
total_purch_val_age = age_sorts["Price"].sum()
avg_tot_purch_per_person_age = total_purch_val_age / age_sorted[age_labels]
age_purchasing_df = pd.DataFrame({'Purchase Count':purchase_count_age,
                                  'Average Purchase Price':avg_purch_price_age,
                                  'Total Purchase Value':total_purch_val_age,
                                  'Avg Total Purchase per Person':avg_tot_purch_per_person_a
ge})
age_purchasing_df = age_purchasing_df.style.format({"Average Purchase Price":'${0:,.2f}',
                                                    "Total Purchase Value":'${0:,.2f}',
                                                    "Avg Total Purchase per Person":'${0:,.2f}'})

age_purchasing_df
```

Out[9]:

	Purchase Count	Average Purchase Price	Total Purchase Value	Avg Total Purchase per Person
Age Range				
<10	23	\$3.35	\$77.13	\$4.54
10-14	28	\$2.96	\$82.78	\$3.76
15-19	136	\$3.04	\$412.89	\$3.86
20-24	365	\$3.05	\$1,114.06	\$4.32
25-29	101	\$2.90	\$293.00	\$3.81
30-34	73	\$2.93	\$214.00	\$4.12
35-39	41	\$3.60	\$147.67	\$4.76
40+	13	\$2.94	\$38.24	\$3.19

In [10]:

```
# Top Spenders
SN_sorted = purchase_data.groupby("SN")
SN_purch_count = SN_sorted["SN"].count()
SN_avg_pp = SN_sorted['Price'].mean()
SN_tot_purch_vol = SN_sorted['Price'].sum()
SN_summary_df = pd.DataFrame({'Purchase Count':SN_purch_count,
                              'Average Purchase Price':SN_avg_pp,
                              'Total Purchase Value':SN_tot_purch_vol})

SN_summary_df = SN_summary_df.sort_values(by='Total Purchase Value',axis=0,ascending=False)
SN_head = SN_summary_df.head()
SN_head.style.format({'Average Purchase Price':'${0:,.2f}',
                     'Total Purchase Value':'${0:,.2f}'})
```

Out[10]:

	Purchase Count	Average Purchase Price	Total Purchase Value
SN			
Lisosia93	5	\$3.79	\$18.96
Idastidru52	4	\$3.86	\$15.45
Chamjask73	3	\$4.61	\$13.83
Iral74	4	\$3.40	\$13.62
Iskadarya95	3	\$4.37	\$13.10

In [11]:

```
# Most Popular Items
popular_items_df = purchase_data[['Item ID','Item Name','Price']]
pop_main = popular_items_df.groupby(['Item ID','Item Name'])
item_prices = popular_items_df.drop_duplicates(subset='Item Name')['Price']
pop_purch_count = pop_main.count()['Price']
pop_item_price = pop_main.mean()['Price']
pop_total_val = pop_main.sum()['Price']
pop_summary = pd.DataFrame({'Purchase Count':pop_purch_count,
                             'Item Price':pop_item_price,
                             'Total Purchase Value':pop_total_val})

pop_summary_count_sorted = pop_summary.sort_values(by='Purchase Count',axis=0,ascending=False)

pop_head_count = pop_summary_count_sorted.head()
pop_head_count.style.format({'Item Price':'${0:,.2f}',
                             'Total Purchase Value':'${0:,.2f}'})
```

Out[11]:

		Purchase Count	Item Price	Total Purchase Value
Item ID	Item Name			
92	Final Critic	13	\$4.61	\$59.99
178	Oathbreaker, Last Hope of the Breaking Storm	12	\$4.23	\$50.76
145	Fiery Glass Crusader	9	\$4.58	\$41.22
132	Persuasion	9	\$3.22	\$28.99
108	Extraction, Quickblade Of Trembling Hands	9	\$3.53	\$31.77

In [12]:

```
# Most Profitable Items
pop_summary_value_sorted = pop_summary.sort_values(by='Total Purchase Value',axis=0,ascending=False)
pop_head_value = pop_summary_value_sorted.head()
pop_head_value.style.format({'Item Price':'${0:,.2f}',
                             'Total Purchase Value':'${0:,.2f}'})
```

Out[12]:

		Purchase Count	Item Price	Total Purchase Value
Item ID	Item Name			
92	Final Critic	13	\$4.61	\$59.99
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
103	Singed Scalpel	8	\$4.35	\$34.80

In []: