# SUPER-MARKET DATA ANALYSIS BY PYTHON - GANESH DEVARE

# DATASET

### Jupyter SuperMarket\_Hackathon Last Checkpoint: Yesterday at 3:08 PM (unsaved changes)



Logout

File Edit View Insert Cell Kernel Widgets Help

Trusted Python 3 O

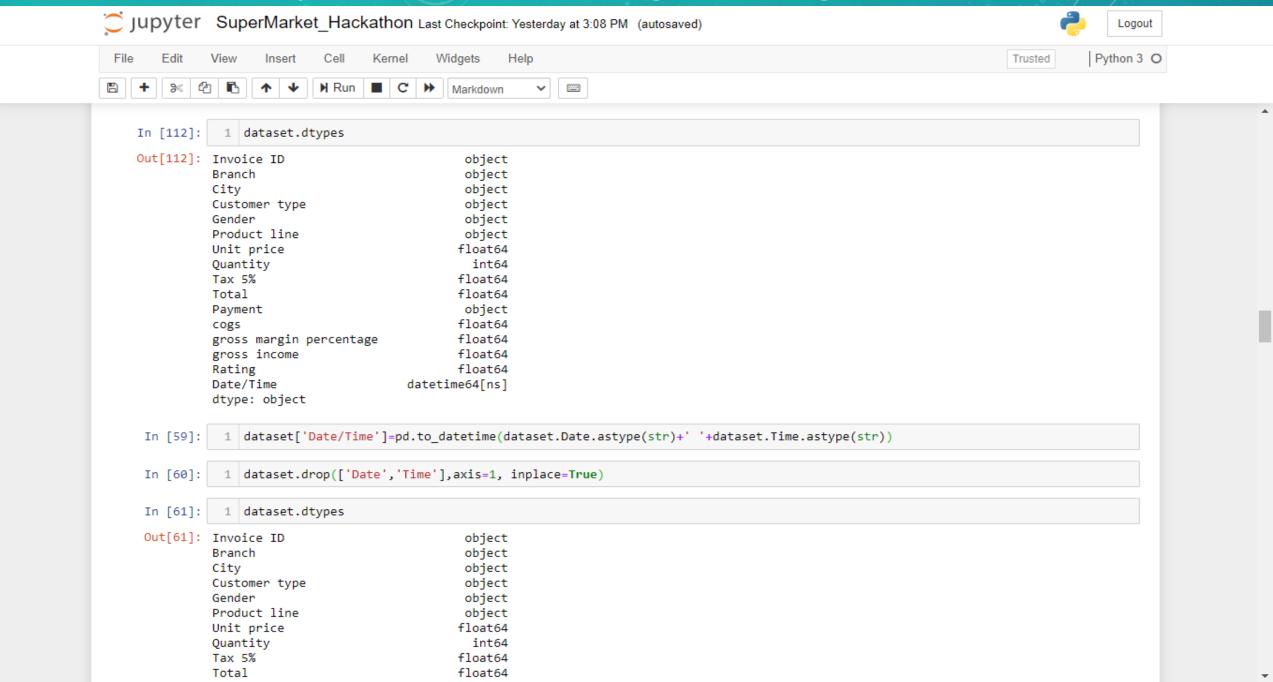
Trusted Python 3 O

In [116]: 1 dataset.head(20)

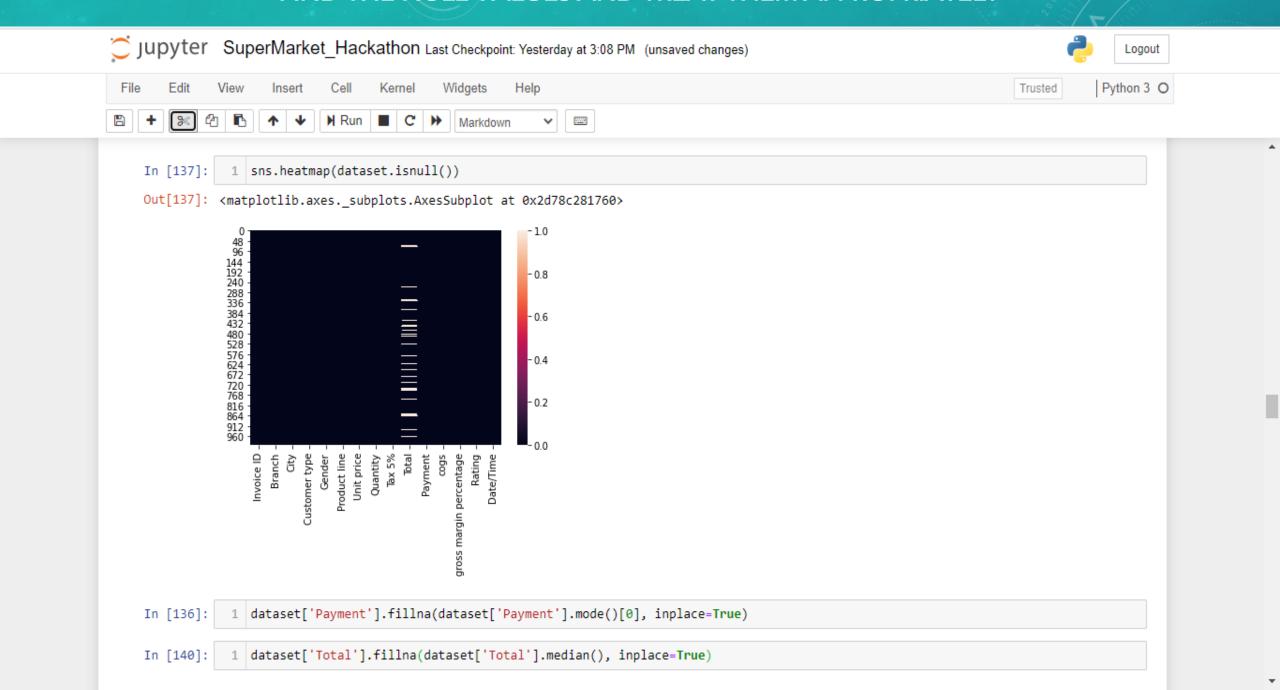
Out[116]:

_		Invoice ID	Branch	City	Customer type	Gender	Product line	Unit price	Quantity	Tax 5%	Total	Payment	cogs	gross margin percentage	gross income	Rating	Date/Time
	0	750-67- 8428	Α	Yangon	Member	Female	Health and beauty	74.69	7	26.1415	548.9715	Ewallet	522.83	4.761905	26.1415	9.1	2019-01- 05 13:08:00
	1	226-31- 3081	С	Naypyitaw	Normal	Female	Electronic accessories	15.28	5	3.8200	80.2200	Cash	76.40	4.761905	3.8200	9.6	2019-03- 08 10:29:00
	2	631-41- 3108	Α	Yangon	Normal	Male	Home and lifestyle	46.33	7	16.2155	340.5255	Credit card	324.31	4.761905	16.2155	7.4	2019-03- 03 13:23:00
	3	123-19- 1176	Α	Yangon	Member	Male	Health and beauty	58.22	8	23.2880	489.0480	Ewallet	465.76	4.761905	23.2880	8.4	2019-01- 27 20:33:00
	4	373-73- 7910	Α	Yangon	Normal	Male	Sports and travel	86.31	7	30.2085	634.3785	Ewallet	604.17	4.761905	30.2085	5.3	2019-02- 08 10:37:00
	5	699-14- 3026	С	Naypyitaw	Normal	Male	Electronic accessories	85.39	7	29.8865	627.6165	Ewallet	597.73	4.761905	29.8865	4.1	2019-03- 25 18:30:00
	6	355-53- 5943	Α	Yangon	Member	Female	Electronic accessories	68.84	6	20.6520	433.6920	Ewallet	413.04	4.761905	20.6520	5.8	2019-02- 25 14:36:00
	7	315-22- 5665	С	Naypyitaw	Normal	Female	Home and lifestyle	73.56	10	36.7800	772.3800	Ewallet	735.60	4.761905	36.7800	8.0	2019-02- 24 11:38:00
	8	665-32- 9167	Α	Yangon	Member	Female	Health and beauty	36.26	2	3.6260	76.1460	Credit card	72.52	4.761905	3.6260	7.2	2019-01- 10 17:15:00
	9	692-92- 5582	В	Mandalay	Member	Female	Food and beverages	54.84	3	8.2260	172.7460	Credit card	164.52	4.761905	8.2260	5.9	2019-02- 20 13:27:00
																	2040.02

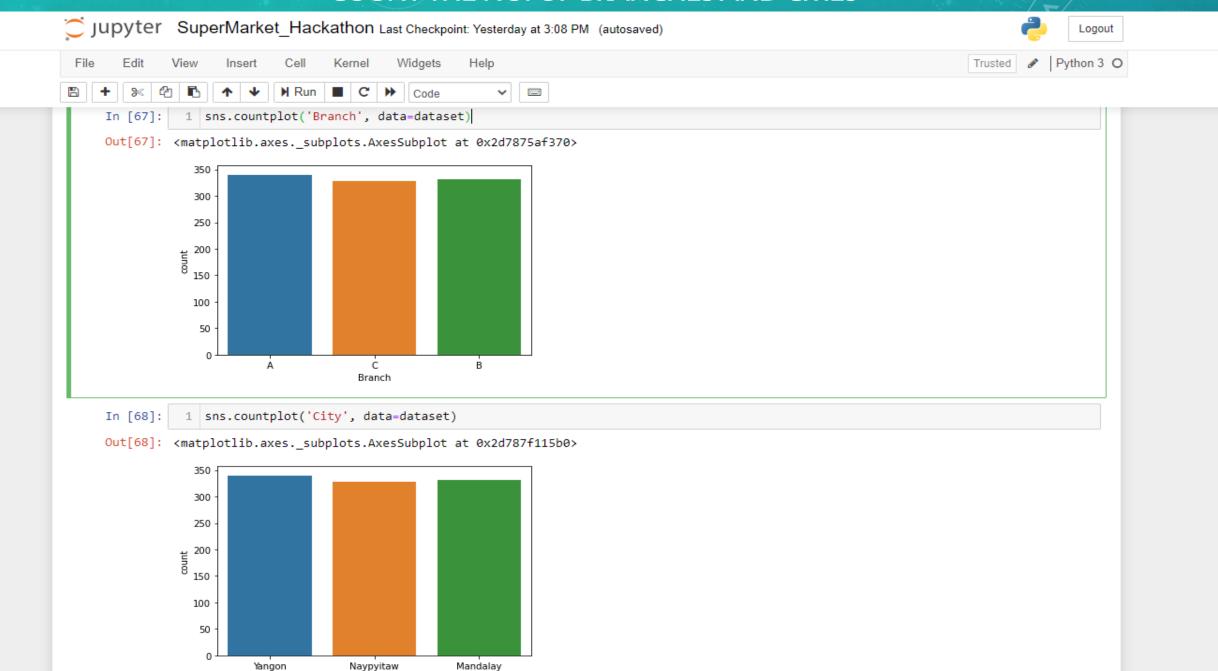
### Check Datatypes of each column and Bring Date time together in one column.



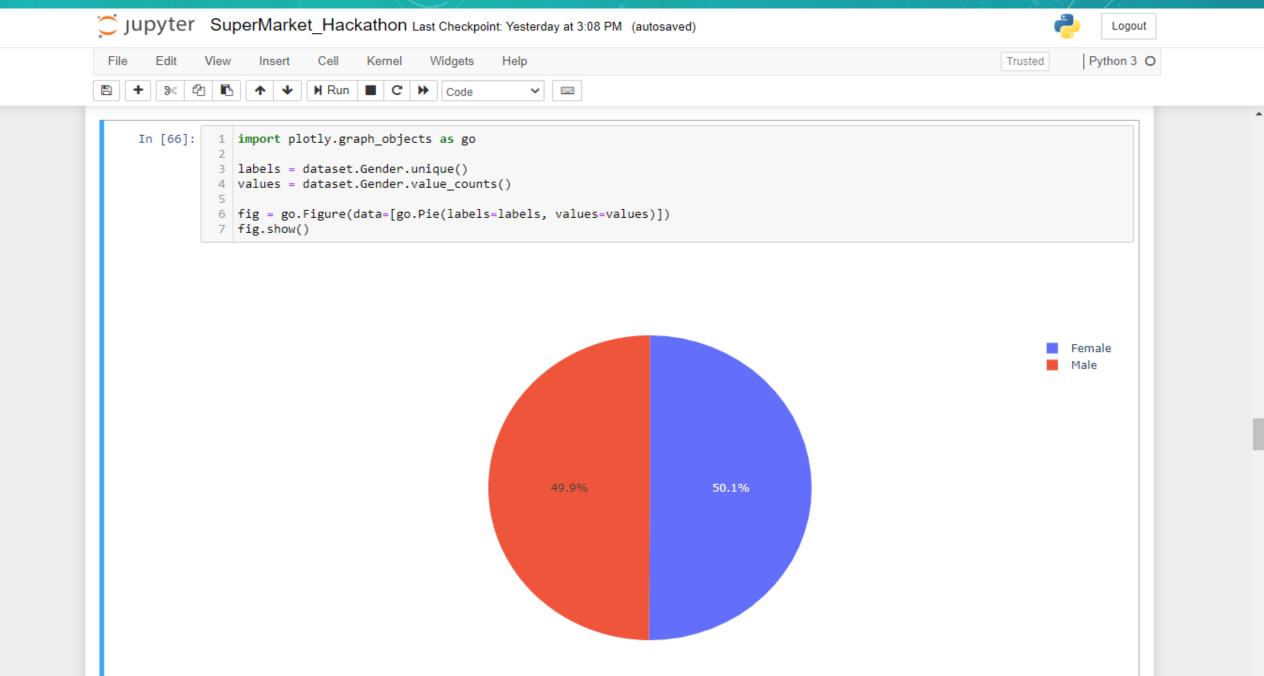
### FIND THE NULL VALUES AND TREAT THEM APPROPRIATELY



### COUNT THE NO. OF BRANCHES AND CITIES



# PIE CHART OF GENDER



# CALCULATE AVERAGE RATING OF EACH PRODUCT LINE

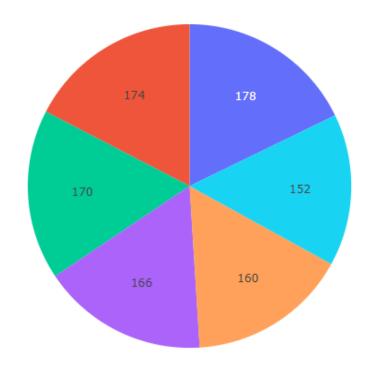
```
In [115]:
             1 plt.figure(figsize=(15,6))
             2 plt.bar(dataset['Product line'].unique(), height=
                         dataset.groupby('Product line')['Rating'].mean(),color=['black', 'red', 'green', 'blue', 'cyan','grey'])
Out[115]: <BarContainer object of 6 artists>
                     Health and beauty
                                                                              Sports and travel
                                                          Home and lifestyle
                                      Electronic accessories
                                                                                               Food and beverages
                                                                                                                  Fashion accessories
```

# NUMBER PRODUCTS IN EACH CATEGORY

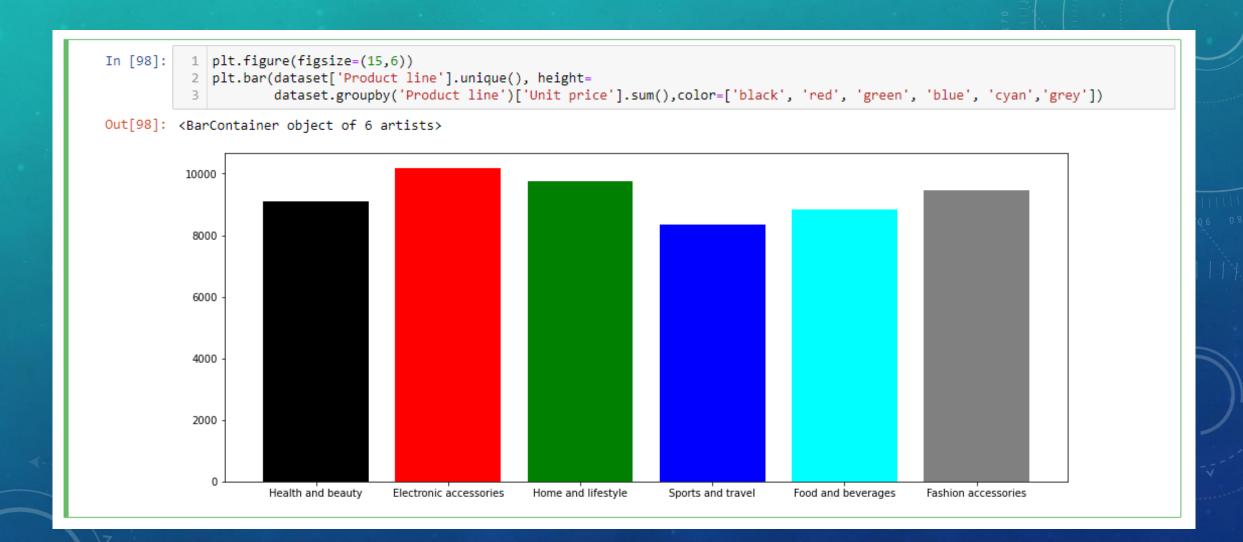
```
In [90]: 1 labels = dataset['Product line'].unique()
2 values = dataset['Product line'].value_counts()
3
4 fig = go.Figure(data=[go.Pie(labels=labels, values=values,textinfo='value')])
5 fig.show()
6 # dataset['Product line'].value_counts(),
```

Health and beauty Electronic accessories Home and lifestyle Sports and travel

Food and beverages Fashion accessories



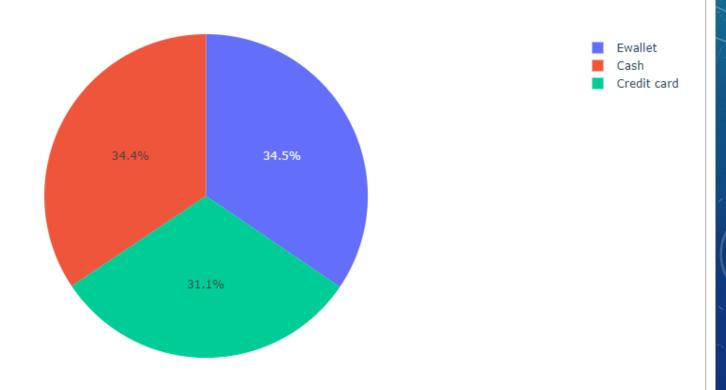
# TOTAL AMOUNT COLLECTED IN EACH PRODUCT LINE



# FIND OUT HIGHEST PERCENTAGE OF PAYMENT METHOD

```
In [102]: 1     labels = dataset['Payment'].unique()
     values = dataset['Payment'].value_counts()

4     fig = go.Figure(data=[go.Pie(labels=labels, values=values)])
     fig.show()
6     # dataset['Product line'].value_counts(),
```



### FIND OUT THE CATEGORY WITH HIGHEST RATING

Home and lifestyle

Sports and travel

Food and beverages

Fashion accessories

Health and beauty

Electronic accessories