Exploratory_Data_Analysis_Retail

```
In [45]: import numpy as np
import pandas as pd
import seaborn as sns
from plotnine import *
import warnings
warnings.filterwarnings('ignore')
import matplotlib.pyplot as plt
```

Reading the dataset

: S	<pre>sample = pd.read_csv("SampleSuperstore.csv")</pre>													
7]: s	ample	.head()												
47]:	S	hip Mode	Segment	Country	City	State	Postal Code	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
0	Sec	ond Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Bookcases	261.9600	2	0.00	41.9136
1	Sec	ond Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Chairs	731.9400	3	0.00	219.5820
2	Sec	ond Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	Labels	14.6200	2	0.00	6.8714
3	Stan	dard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	Tables	957.5775	5	0.45	-383.0310
4	Stan	dard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	Storage	22.3680	2	0.20	2.5164
: s	ample	.tail()												
	р													
]: _			de Segme	nt Countr	y City	State	Postal Code	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
_	989	Ship Mod		nt Counti			Postal Code 33180	Region South	Category Furniture	Sub-Category Furnishings		Quantity 3	Discount 0.2	Profit 4.1028
9	989	Ship Mod Second Cla	ss Consum	er United State		Florida					25.248		0.2	
9	989 990 S	Ship Moo Second Cla	ss Consum	er United State	es Miami es Costa Mesa	Florida California	33180	South	Furniture	Furnishings Furnishings	25.248	3	0.2	4.1028
9	989 990 S	Ship Mod Second Cla Standard Cla	iss Consum iss Consum	er United State er United State er United State	es Miami es Costa Mesa	Florida California California	33180 92627	South West West	Furniture Furniture	Furnishings Furnishings Phones	25.248 91.960	3	0.2 0.0 0.2	4.1028 15.6332
9 9	989 990 S	Ship Moo Second Cla Standard Cla Standard Cla	iss Consum iss Consum iss Consum iss Consum	er United State er United State er United State	es Miami es Costa Mesa es Costa Mesa	Florida California California California	33180 92627 92627	South West West West	Furniture Furniture Technology	Furnishings Furnishings Phones	25.248 91.960 258.576 29.600	3 2 2	0.2 0.0 0.2 0.0	4.1028 15.6332 19.3932
9 9 9 9	989 990 S 991 S 992 S	Ship Moo Second Cla Standard Cla Standard Cla	iss Consum iss Consum iss Consum iss Consum	er United State er United State er United State	es Miami es Costa Mesa es Costa Mesa es Costa Mesa	Florida California California California	33180 92627 92627 92627	South West West West	Furniture Furniture Technology Office Supplies	Furnishings Furnishings Phones Paper	25.248 91.960 258.576 29.600	3 2 2 4	0.2 0.0 0.2 0.0	4.1028 15.6332 19.3932 13.3200

Checking for the data's information, i.e type

```
In [50]: sample.info()
           <class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
          # Column Non-Null Coun
                                Non-Null Count Dtype
                              9994 non-null object
            0 Ship Mode
                                 9994 non-null
                                 9994 non-null
                Country
                                                    object
                                                    object
object
int64
                City
                                 9994 non-null
                State
Postal Code
                                 9994 non-null
9994 non-null
                 Region
                                 9994 non-null
                                                    object
                Category
Sub-Category
Sales
                                 9994 non-null
9994 non-null
9994 non-null
                                                    object
object
float64
            10 Quantity
11 Discount
12 Profit
                                                    int64
float64
float64
                                 9994 non-null
                                 9994 non-null
9994 non-null
          dtypes: float64(3), int64(2), object(8) memory usage: 1015.1+ KB
In [51]: sample.describe()
                   Postal Code
                                      Sales Quantity Discount
                                                                               Profit
           count 9994.000000 9994.000000 9994.000000 9994.000000

        mean
        55190.379428
        229.858001
        3.789574
        0.156203
        28.656896

             std 32063.693350 623.245101 2.225110 0.206452 234.260108
           min 1040.000000 0.444000 1.000000 0.000000 -6599.978000
            25% 23223.000000 17.280000 2.000000 0.000000
                                                                            1 728750
                                                                          8.666500
            50% 56430.500000 54.490000 3.000000 0.200000
            75% 90008.000000 209.940000 5.000000 0.200000
                                                                          29.364000
           max 99301.000000 22638.480000 14.000000 0.800000 8399.976000
In [52]: sample.isnull().sum()
           Ship Mode
          Segment
Country
          City
State
Postal Code
           Region
           Category
Sub-Category
```

Checking for the duplicate data. If yes, then dropping those data

Sales Quantity Discount Profit

In [53]: sample.duplicated().sum()

Out[53]: **17** In [54]: sample.drop_duplicates() Out[54]: Ship Mode Segment State Postal Code Region Category Sub-Category Sales Quantity Discount Country City Profit

0	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Bookcases	261.9600	2	0.00	41.9136
1	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	Chairs	731.9400	3	0.00	219.5820
2	Second Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	Labels	14.6200	2	0.00	6.8714
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	Tables	957.5775	5	0.45	-383.0310
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	Storage	22.3680	2	0.20	2.5164

9989	Second Class	Consumer	United States	Miami	Florida	33180	South	Furniture	Furnishings	25.2480	3	0.20	4.1028
9990	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Furniture	Furnishings	91.9600	2	0.00	15.6332
9991	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Technology	Phones	258.5760	2	0.20	19.3932
9992	Standard Class	Consumer	United States	Costa Mesa	California	92627	West	Office Supplies	Paper	29.6000	4	0.00	13.3200
9993	Second Class	Consumer	United States	Westminster	California	92683	West	Office Supplies	Appliances	243.1600	2	0.00	72.9480

9977 rows × 13 columns

In [55]: sample.nunique()

Out[55]: Ship Mode Segment Country City 531 State 49 Postal Code 631 Region Region Category Sub-Category Sales 5825 Quantity Discount Profit dtype: int64

Dropping irrelevant columns

In [56]: col = ['Postal Code']
sample1 = sample.drop(columns=col, axis=1) sample1

[56]:		Ship Mode	Segment	Country	City	State	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
	0	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Bookcases	261.9600	2	0.00	41.9136
	1	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Chairs	731.9400	3	0.00	219.5820
	2	Second Class	Corporate	United States	Los Angeles	California	West	Office Supplies	Labels	14.6200	2	0.00	6.8714
	3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Furniture	Tables	957.5775	5	0.45	-383.0310
	4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Office Supplies	Storage	22.3680	2	0.20	2.5164
		•••											
9	9989	Second Class	Consumer	United States	Miami	Florida	South	Furniture	Furnishings	25.2480	3	0.20	4.1028
9	9990	Standard Class	Consumer	United States	Costa Mesa	California	West	Furniture	Furnishings	91.9600	2	0.00	15.6332
9	9991	Standard Class	Consumer	United States	Costa Mesa	California	West	Technology	Phones	258.5760	2	0.20	19.3932
9	9992	Standard Class	Consumer	United States	Costa Mesa	California	West	Office Supplies	Paper	29.6000	4	0.00	13.3200
9	9993	Second Class	Consumer	United States	Westminster	California	West	Office Supplies	Appliances	243.1600	2	0.00	72.9480

9994 rows × 12 columns

Checking statistical relation between the various rows & columns

In [57]: # Correlation between variables

sample1.corr()

Out[57]: Sales Quantity Discount Sales 1.000000 0.200795 -0.028190 0.479064 **Quantity** 0.200795 1.000000 0.008623 0.066253 **Discount** -0.028190 0.008623 1.000000 -0.219487 Profit 0.479064 0.066253 -0.219487 1.000000

In [58]: # Covariance of columns sample1.cov()

Sales Quantity Discount Profit Out[58]: Sales 388434.455308 278.459923 -3.627228 69944.096586 **Quantity** 278.459923 4.951113 0.003961 34.534769

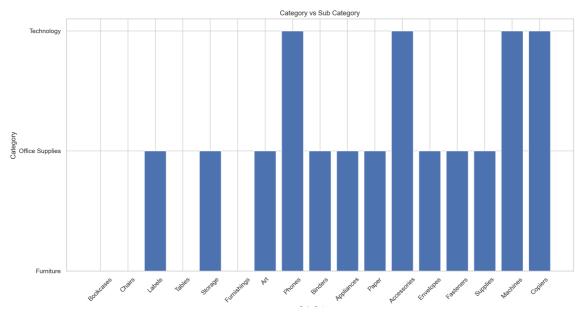
> **Discount** -3.627228 0.003961 0.042622 -10.615173 Profit 69944.096586 34.534769 -10.615173 54877.798055

In [59]: sample1.head() # loads the first five rows

[59]:	Ship Mode	Segment	Country	City	State	Region	Category	Sub-Category	Sales	Quantity	Discount	Profit
0	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Bookcases	261.9600	2	0.00	41.9136
1	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Chairs	731.9400	3	0.00	219.5820
2	Second Class	Corporate	United States	Los Angeles	California	West	Office Supplies	Labels	14.6200	2	0.00	6.8714
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Furniture	Tables	957.5775	5	0.45	-383.0310
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Office Supplies	Storage	22.3680	2	0.20	2.5164

Data Visualization

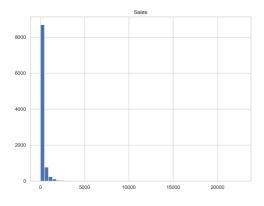
```
In [60]: plt.figure(figsize=(16,8))
   plt.ban('Sub-Category', 'Category', data=sample1)
   plt.title('Category vs Sub Category')
   plt.xlabel('Sub-Category')
   plt.ylabel('Category')
   plt.xticks(rotation=45)
   plt.show()
```

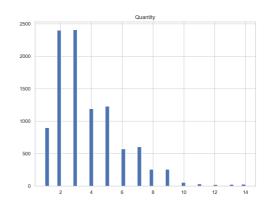


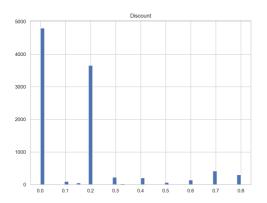
In [61]: sample1.corr() # Checking the correlation

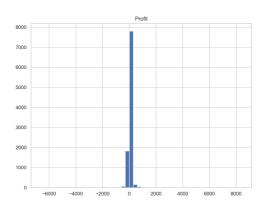
t[61]:		Sales	Quantity	Discount	Profit
	Sales	1.000000	0.200795	-0.028190	0.479064
	Quantity	0.200795	1.000000	0.008623	0.066253
	Discount	-0.028190	0.008623	1.000000	-0.219487
	Dundia	0.470064	0.066353	0.210497	1 000000

In [62]: sample1.hist(bins=50,figsize=(20,15))



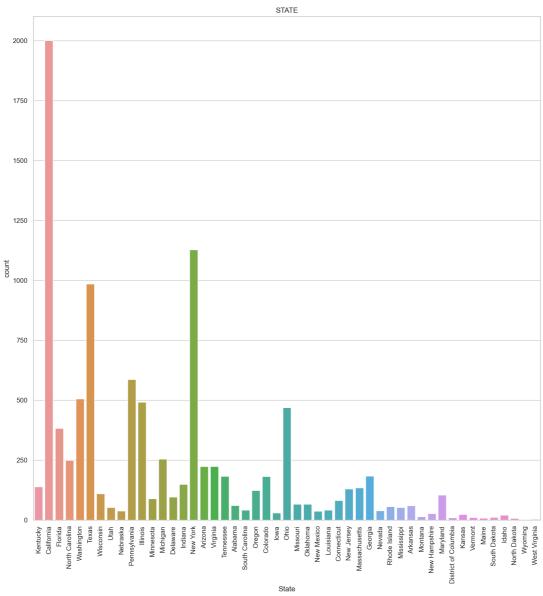




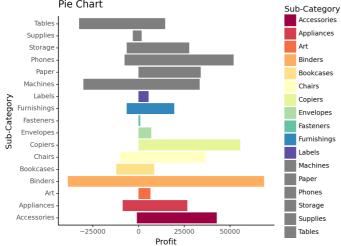


So, the data here is not normal as revealed by this histogram graph.

```
In [63]: # Count the total repeatable states
                    sample1['State'].value_counts()
                   California
New York
Out[63]:
                                                                      1128
                   Texas
Pennsylvania
Washington
Illinois
                                                                       985
587
596
492
469
383
255
249
224
184
183
182
149
135
130
124
110
                   Ohio
Florida
Michigan
North Carolina
Arizona
Virginia
                   Georgia
Tennessee
Colorado
                    Indiana
                   Kentucky
Massachusetts
New Jersey
                   Oregon
Wisconsin
Maryland
                   Delaware
Minnesota
Connecticut
                                                                         96
89
82
66
66
61
60
56
53
42
42
39
38
37
30
27
24
21
15
                   Oklahoma
Missouri
Alabama
                   Arkansas
Rhode Island
Utah
                    Mississippi
                    Louisiana
South Carolina
                    Nevada
                    Nebraska
New Mexico
                    Iowa
                   New Hampshire
Kansas
Idaho
                    Montana
                    South Dakota
                    Vermont
                    District of Columbia
                                                                         10
8
7
                   Maine
North Dakota
                   West Virginia
Wyoming
Name: State, dtype: int64
In [66]: plt.figure(figsize=(15,15))
    sns.countplot(x=sample1['State'])
    plt.xticks(rotation=90)
    plt.title("STATE")
    plt.show()
```







<Figure Size: (640 x 480)>

The above pie chart shows the profit and loss of each and every subcategories.

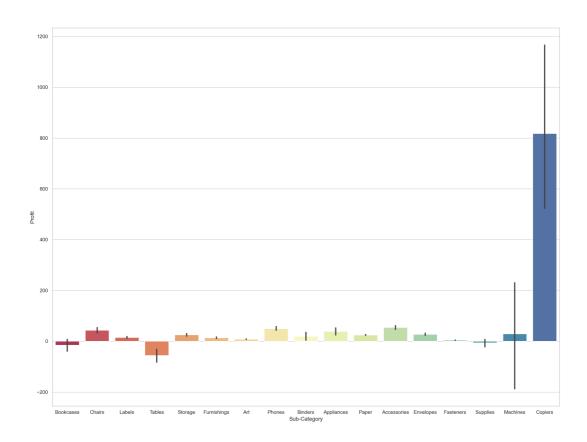
Here from the graph we can visualize that "binders" sub-category has suffered the highest amount of loss and also profit amongst all other sub-Categories (For now we can't say that what is the reason it may be because of discounts given on binders subcategory).

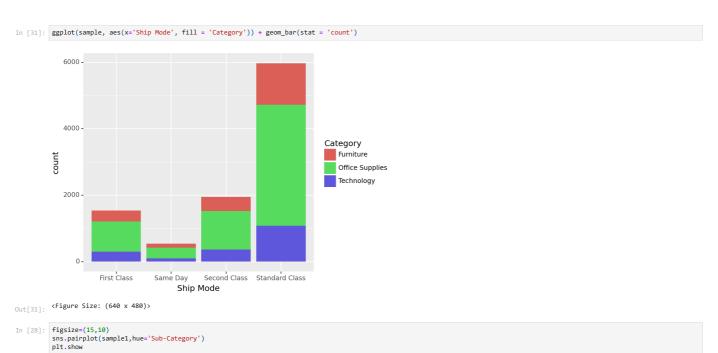
Next, "Copiers" sub-category has gained highest amount of profit with no loss. There are other sub-categories too haven't faced any kind of losses but their profit margins are also low.

Next, suffering from highest loss is machines.

```
In [26]: sns.set(style="whitegrid")
plt.figure(2, figsize=(20,15))
sns.barplot(x='Sub-Category',y='Profit', data=sample, palette='Spectral')
plt.suptitle('Pie Consumption Patterns in the United States', fontsize=16)
plt.show()
```

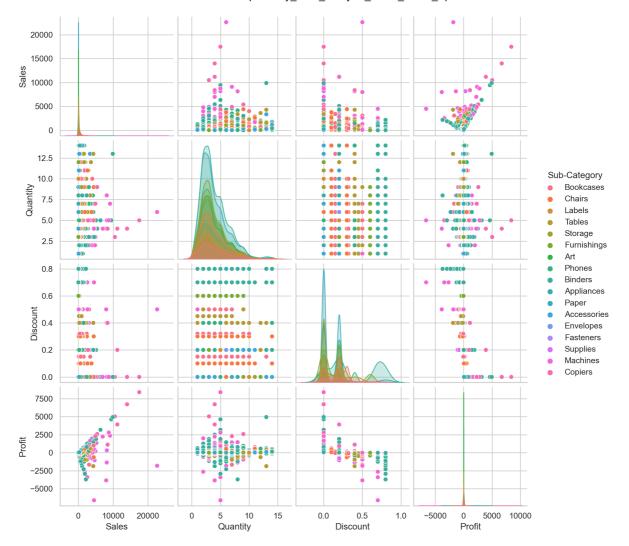
Pie Consumption Patterns in the United States





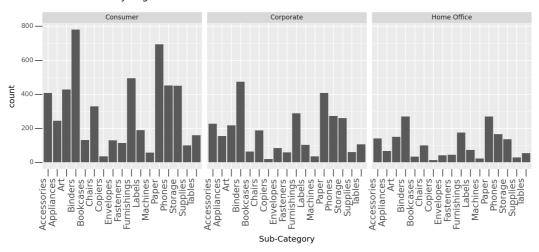
<function matplotlib.pyplot.show(close=None, block=None)>

Out[28]:



From the above plot we can say that our data is not normal and it has some amount of outliers too. Let's explore more about these outliers by using boxplots. 1st we'll check sales from every segments of whole data.

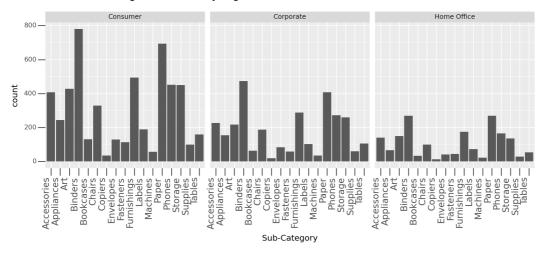
Sales From Every Segment Of United States of Whole Data



Out[32]: <Figure Size: (1000 x 500)>

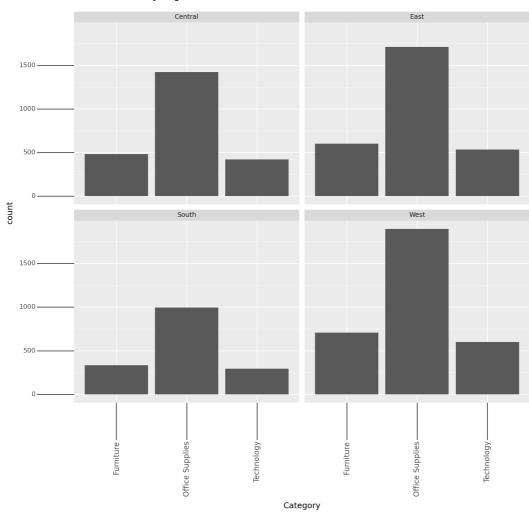
From the above Graph we can say that "Home Office" segment has less purchased sub-categories and in that "Tables", "Supplies", "Machines", "Copiers", "Bookcases" has the lowest sales. "Consumer" has purchased more sub-categories as compared to other segments.

Discount on Categories From Every Segment Of United States of Whole Data



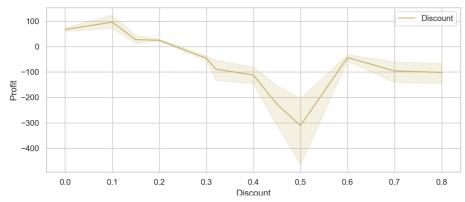
Out[33]: <Figure Size: (1000 x 500)>

Sales From Every Region Of United States of Whole Data

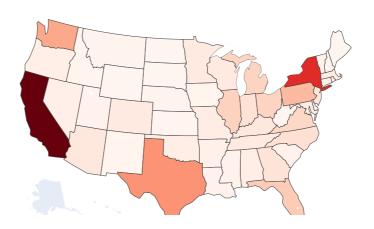


Out[34]: <Figure Size: (1000 x 1000)>

```
In [36]: plt.figure(figsize=(10,4))
sns.lineplot(x='Discount', y='Profit', data=sample1, color='y', label='Discount')
plt.legend()
plt.show()
```



Total State-Wise Sales



Now, let us analyze the sales of a few random states from each profit bracket (high profit, medium profit, low profit, low loss and high loss) and try to observe some crucial trends which might help us in increasing the sales.

We have a few **questions** to answer here.

- 1. What products do the most profit making states buy?
- 2. What products do the loss bearing states buy?
- 3. What product segment needs to be improved in order to drive the profits higher?

```
In [40]: def state_data_viewer(states):
    """Plots the turnover generated by different product categories and sub-categories for the list of given states.
Args:
```

```
states- List of all the states you want the plots for

Returns:

None

product_data = sample1.groupby(['State'])

for state in states:

data = product_data.get_group(state).groupby(['Gategory'])

fig, suptitic(state, fontsize-14)

ax_index = 0

for cat in ['Turniture', 'Office Supplies', 'Technology']:

cat_data = data.get_group(cat).groupby(['Sub-Category']).sum()

sx.index = 1

fig.suphic(x) = cat_data.profit, y = cat_data.index, ax = ax[ax_index])

ax_index = 1

fig.show()

In [41]:

states = ['California', 'Washington', 'Mississippi', 'Arizona', 'Texas']

state_data_viewer(states)

Datama

Datama

Datama

Datama

Datama

Datama

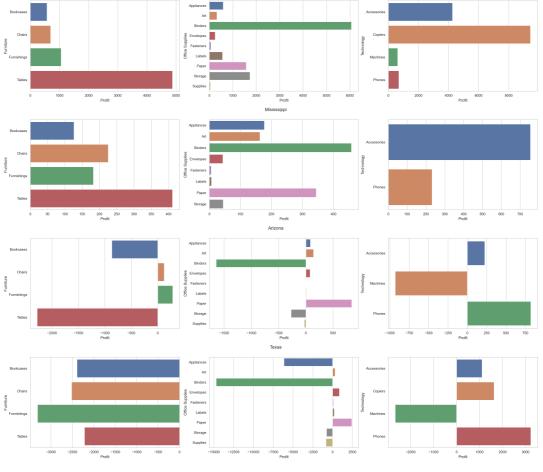
Datama

Datama

Datama

Answers

Answ
```



Thank You!

GitHub: https://github.com/anujtiwari21?tab=repositories