Data Science and Business Analytics Task 3:-

INTERN NAME: TWINKLE PANDEY

To conduct Exploratory Data Analysis (EDA) on the 'SampleSuperstore' dataset, we will start by loading the dataset and examining its structure. We will clean the data to handle any missing values, duplicates, and inconsistencies. Using various visualizations, we will identify trends and patterns to uncover weak areas and potential business problems. Our goal is to provide insights and suggest strategies to improve profitability

Exploratory Data Analysis

Problem Statement:

- 1. Perform 'Exploratory Data Analysis' on dataset 'SampleSuperstore'.
- 2. As manager, try to find weak areas where you can work to make more profit.
- 3. What all business problems you can derive by exploring the data?

Importing Required Libraries

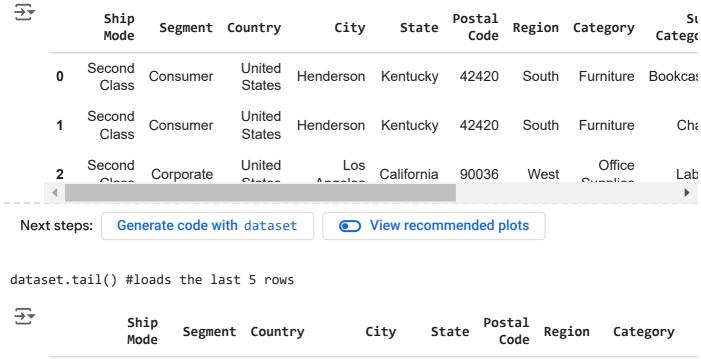
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

Importing Dataset

```
path = "/content/SampleSuperstore.csv"
dataset = pd.read_csv(path)
```

Reading the dataset

```
dataset.head() #loads the first 5 rows
```



Second United 9989 Consumer Miami Florida 33180 South Furniture Class States United Standard 9990 Consumer Costa Mesa California 92627 West Furniture Class States United Standard 9991 Consumer Costa Mesa California 92627 Technology West Class Ctataa

Checking the number of elements in each dimension in an array

dataset.shape

→ (9994, 13)

Checking the information of Data

dataset.info() #Returns the concise summary of the dataset

<class 'pandas.core.frame.DataFrame'> RangeIndex: 9994 entries, 0 to 9993 Data columns (total 13 columns): # Column Non-Null Count Dtype Ship Mode 0 9994 non-null object 1 Segment 9994 non-null object 2 Country 9994 non-null object 3 City 9994 non-null object 4 State 9994 non-null object 5 9994 non-null int64 Postal Code 6 9994 non-null object Region 7 Category 9994 non-null object

```
8 Sub-Category 9994 non-null object
9 Sales 9994 non-null float64
10 Quantity 9994 non-null int64
11 Discount 9994 non-null float64
12 Profit 9994 non-null float64
dtypes: float64(3), int64(2), object(8)
```

memory usage: 1015.1+ KB

dataset.describe() #returns the statistical data

→		Postal Code	Sales	Quantity	Discount	Profit
	count	9994.000000	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
	50%	56430.500000	54.490000	3.000000	0.200000	8.666500
	75%	90008.000000	209.940000	5.000000	0.200000	29.364000
	max	99301.000000	22638.480000	14.000000	0.800000	8399.976000

Checking the missing values

dataset.isnull().sum()

\rightarrow	Ship Mode	0
	Segment	0
	Country	0
	City	0
	State	0
	Postal Code	0
	Region	0
	Category	0
	Sub-Category	0
	Sales	0
	Quantity	0
	Discount	0
	Profit	0
	dtype: int64	

Checking for the duplicate data

dataset.duplicated().sum()

→ 17

Dropping the duplicated data

dataset.drop_duplicates()

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	Ship Mode	Segment	Country	City	State	Postal Code	Region	Category	
0	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	E
1	Second Class	Consumer	United States	Henderson	Kentucky	42420	South	Furniture	
2	Second Class	Corporate	United States	Los Angeles	California	90036	West	Office Supplies	
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Furniture	
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	33311	South	Office Supplies	
9989	Second Class	Consumer	United States	Miami	Florida	33180	South	Furniture	F
1	Standard	^	United	O1- NA	O = 1:f = ! =	00007	\\/ L	F 14	>

dataset.nunique() # Displays the unique data now

$\overline{\Rightarrow}$	Ship Mode	4
	Segment	3
	Country	1
	City	531
	State	49
	Postal Code	631
	Region	4
	Category	3
	Sub-Category	17
	Sales	5825
	Quantity	14
	Discount	12
	Profit	7287
	dtype: int64	

Dropping irrelevant columns

```
col = ["Postal Code"]
dataset1 = dataset.drop(columns = col,axis = 1)
```

dataset1

→		Ship Mode	Segment	Country	City	State	Region	Category	Sub- Category
	0	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Bookcases
	1	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Chairs
	2	Second Class	Corporate	United States	Los Angeles	California	West	Office Supplies	Labels
	3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Furniture	Tables
	4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	South	Office Supplies	Storage
	9989	Second Class	Consumer	United States	Miami	Florida	South	Furniture	Furnishings
	2000	Standard	^	United	O1- NA	O - I'f ' -	1A11	······································	
									<u>'</u> -
Next steps:		Generat	e code with	dataset1	Viev	v recommer	nded plots		

Checking statistical relation between the various rows & columns

```
# Select only numeric columns
numeric_data = dataset1[['Sales', 'Quantity', 'Discount', 'Profit']]
# Calculate the correlation matrix
correlation_matrix = numeric_data.corr()
# Print the correlation matrix
correlation_matrix
```

→		Sales	Quantity	Discount	Profit	
	Sales	1.000000	0.200795	-0.028190	0.479064	ıl.
	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	

Next steps:

Generate code with correlation_matrix



View recommended plots

```
# Calculate the covariance matrix
covariance_matrix = numeric_data.cov()
```

Print the covariance matrix
covariance_matrix

→		Sales	Quantity	Discount	Profit	\blacksquare
	Sales	388434.455308	278.459923	-3.627228	69944.096586	11.
	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	

Next steps: Generate code with covariance_matrix View recommended plots

dataset1.head() #loads first five rows

→		Ship Mode	Segment	Country	City	State	Region	Category	Sub- Category	S
	0	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Bookcases	261.
	1	Second Class	Consumer	United States	Henderson	Kentucky	South	Furniture	Chairs	731.
	2	Second	Corporate	United	Los	California	West	Office	Labels	14. •

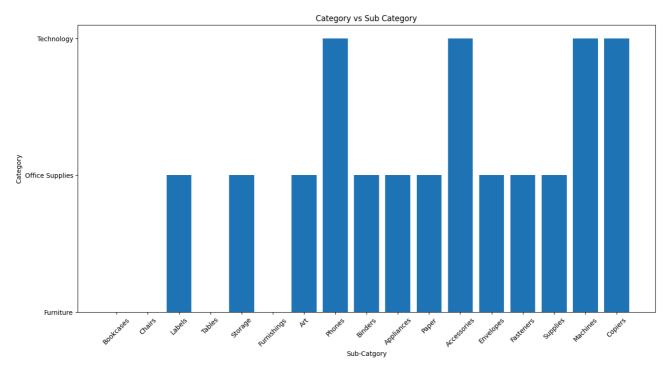
Next steps: Generate code with dataset1

View recommended plots

Data Visualisation

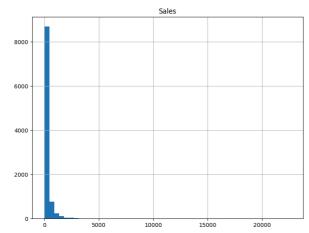
```
plt.figure(figsize=(16,8))
plt.bar('Sub-Category','Category', data=dataset1)
plt.title('Category vs Sub Category')
plt.xlabel('Sub-Catgory')
plt.ylabel('Category')
plt.xticks(rotation=45)
plt.show()
```

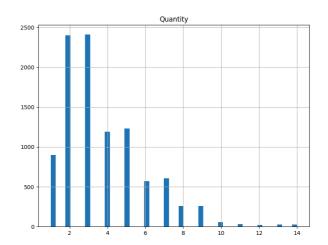


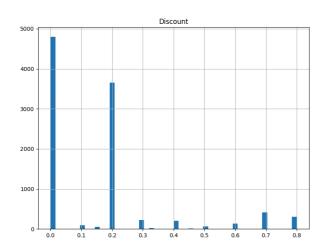


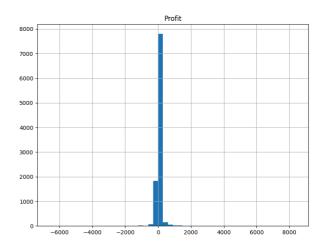
dataset1.hist(bins=50,figsize=(20,15))
plt.show();









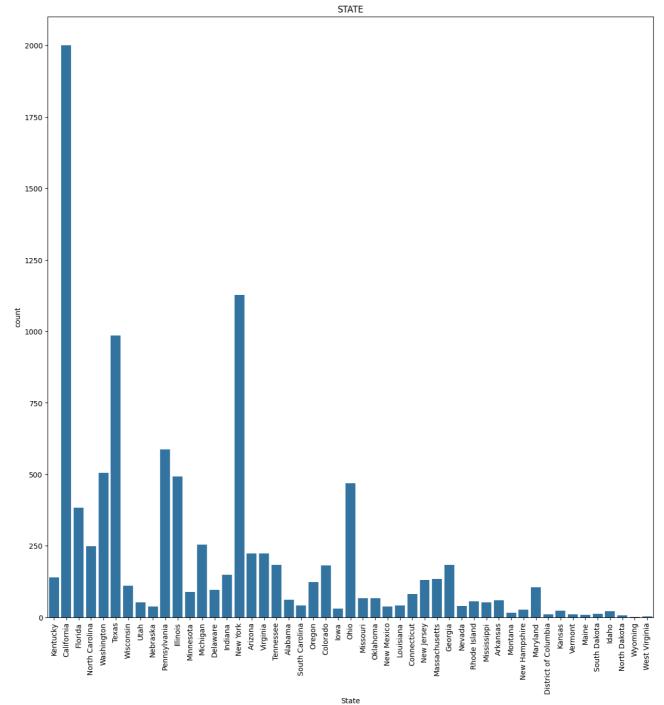


Count the total repeatable states
dataset1['State'].value_counts()

→	State	
<u> </u>	California	2001
	New York	1128
	Texas	985
	Pennsylvania	587
	Washington	506
	Illinois	492
	Ohio	
	Florida	469 383
	Michigan	255
	North Carolina	
		249
	Arizona	224
	Virginia	224
	Georgia	184
	Tennessee	183
	Colorado	182
	Indiana	149
	Kentucky	139
	Massachusetts	135
	New Jersey	130
	Oregon	124
	Wisconsin	110
	Maryland	105
	Delaware	96
	Minnesota	89
	Connecticut	82
	Oklahoma	66
	Missouri	66
	Alabama	61
	Arkansas	60
	Rhode Island	56
	Utah	53
	Mississippi	53
	Louisiana	42
	South Carolina	42
	Nevada	39
	Nebraska	38
	New Mexico	37
	Iowa	30
	New Hampshire	27
	Kansas	24
	Idaho	21
	Montana	15
	South Dakota	12
	Vermont	11
	District of Columbia	10
	Maine	8
	North Dakota	7
	West Virginia	4
	Wyoming	1
	Name: count, dtype: inte	54

```
plt.figure(figsize=(15,15))
sns.countplot(x=dataset1['State'])
plt.xticks(rotation=90)
plt.title("STATE")
plt.show()
```





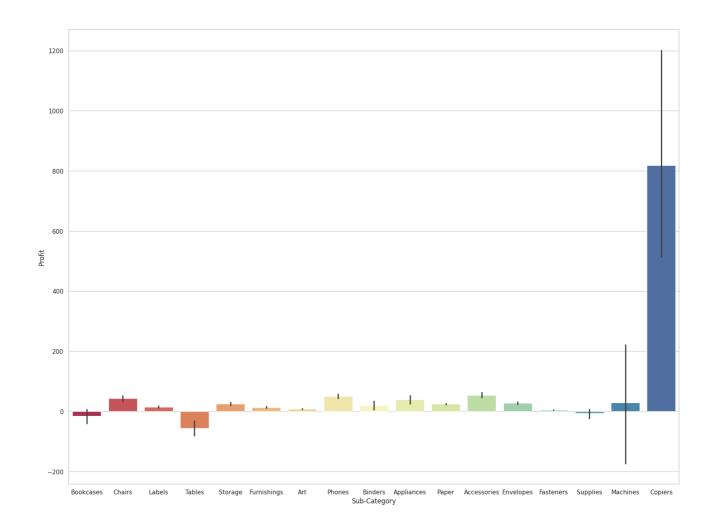
```
sns.set(style="whitegrid")
plt.figure(2, figsize=(20,15))
sns.barplot(x='Sub-Category',y='Profit', data=dataset, palette='Spectral')
plt.suptitle('Pie Consumption Patterns in the United States', fontsize=16)
plt.show()
```



<ipython-input-47-8a12df664a8c>:3: FutureWarning:

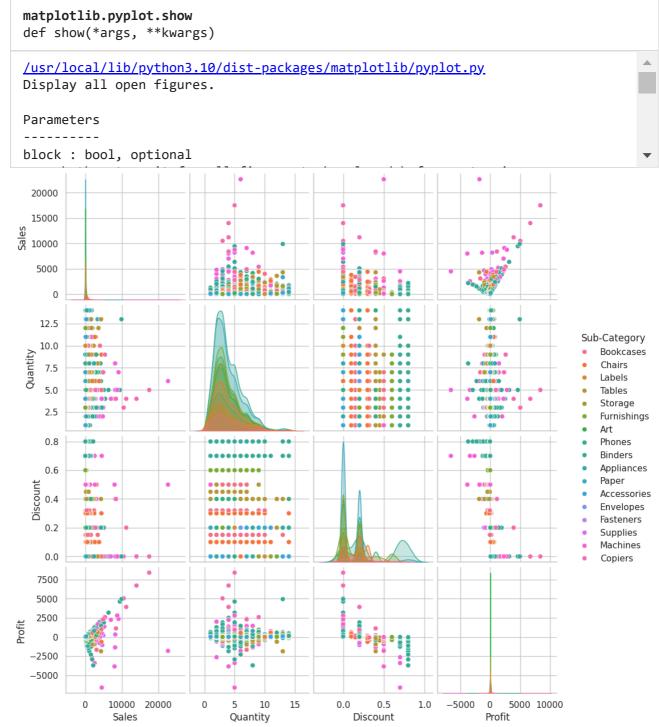
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14 sns.barplot(x='Sub-Category',y='Profit', data=dataset, palette='Spectral')

Pie Consumption Patterns in the United States



figsize=(15,10)
sns.pairplot(dataset1,hue='Sub-Category')
plt.show

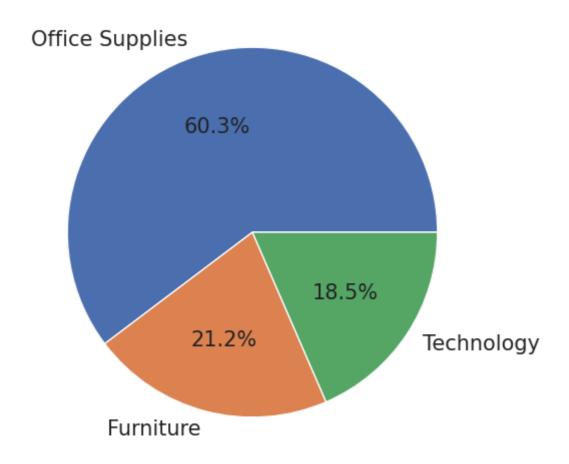




```
plt.figure(figsize = (6,6))
textprops = {"fontsize":15}
plt.title('Category')
plt.pie(dataset['Category'].value_counts(), labels=dataset['Category'].value_counts().inc
plt.show()
```

 $\overline{\Rightarrow}$

Category



```
# computing top categories in terms of sales from first 100 observations

top_category_s = dataset.groupby("Category").Sales.sum().nlargest(n=100)

# computing top categories in terms of profit from first 100 observations

top_category_p = dataset.groupby("Category").Profit.sum().nlargest(n=100)

# plotting to see it visually

plt.style.use('seaborn')

top_category_s.plot(kind = 'bar',figsize = (10,5),fontsize = 14)

top_category_p.plot(kind = 'bar',figsize = (10,5),fontsize = 14,color='yellow')

plt.xlabel('Category',fontsize = 15)

plt.ylabel('Total Sales/Profits',fontsize = 15)

plt.title("Top Category Sales vs Profit",fontsize = 15)

plt.show()
```

<ipython-input-57-1747aad01761>:11: MatplotlibDeprecationWarning: The seaborn styles
 plt.style.use('seaborn')



Visualising the Sub Categories

```
# computing top sub-categories in terms of sales from first 100 observations

top_subcategory_s = dataset.groupby("Sub-Category").Sales.sum().nlargest(n = 100)

# computing top sub-categories in terms of profit from first 100 observations

top_subcategory_p = dataset.groupby("Sub-Category").Profit.sum().nlargest(n = 100)

# plotting to see it visually

plt.style.use('seaborn')

top_subcategory_s.plot(kind = 'bar',figsize = (10,5),fontsize = 14)

top_subcategory_p.plot(kind = 'bar',figsize = (10,5),fontsize = 14, color = 'red')

plt.xlabel('Sub-Category',fontsize = 15)

plt.ylabel('Total Sales/Profits',fontsize = 15)

plt.title("Top Sub-Category Sales vs Profit",fontsize = 15)

plt.show()
```



<ipython-input-58-f6bf79d7e4f8>:11: MatplotlibDeprecationWarning: The seaborn styles
 plt.style.use('seaborn')



```
# A more detailed view
plt.figure(figsize=(14,12))
statewise = dataset.groupby(['Sub-Category'])['Profit'].sum().nlargest(50)
statewise.plot.barh() # h for horizontal
```

→*

<Axes: ylabel='Sub-Category'>

