analysis-retail-samplesuperstore

October 20, 2023

GRIP - THE SPARK FOUNDATION

AUTHOR - SHIVAM DUBEY

DATA SCIENCE & BUSINESS ANALYTICS TASK -

Exploratory Data Analysis-Retail(SampleSuperStore) TASK - 3

Exploratory Data Analysis-Retail(SampleSuperStore)

Import libraries

[ ]:
```
import numpy as np # linear algebra
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import plotly.graph_objects as go
from plotly.subplots import make_subplots
```

Loading the DataSet

[ ]:
```
df = pd.read_csv('SampleSuperstore.csv')
df
```

[ ]:

	Ship Mode	Segment	Country	City	State	\
0	Second Class	Consumer	United States	Henderson	Kentucky	
1	Second Class	Consumer	United States	Henderson	Kentucky	
2	Second Class	Corporate	United States	Los Angeles	California	
3	Standard Class	Consumer	United States	Fort Lauderdale	Florida	
4	Standard Class	Consumer	United States	Fort Lauderdale	Florida	
…	…	…	…	…	…	
9989	Second Class	Consumer	United States	Miami	Florida	
9990	Standard Class	Consumer	United States	Costa Mesa	California	
9991	Standard Class	Consumer	United States	Costa Mesa	California	
9992	Standard Class	Consumer	United States	Costa Mesa	California	

```
9993    Second Class   Consumer  United States      Westminster  California

      Postal Code Region         Category Sub-Category     Sales  Quantity  \
0           42420  South        Furniture    Bookcases  261.9600         2
1           42420  South        Furniture       Chairs  731.9400         3
2           90036   West  Office Supplies       Labels   14.6200         2
3           33311  South        Furniture       Tables  957.5775         5
4           33311  South  Office Supplies      Storage   22.3680         2
…             …    …              …           …       …       …
9989        33180  South        Furniture  Furnishings   25.2480         3
9990        92627   West        Furniture  Furnishings   91.9600         2
9991        92627   West       Technology       Phones  258.5760         2
9992        92627   West  Office Supplies        Paper   29.6000         4
9993        92683   West  Office Supplies   Appliances  243.1600         2

      Discount    Profit
0         0.00   41.9136
1         0.00  219.5820
2         0.00    6.8714
3         0.45 -383.0310
4         0.20    2.5164
…          …       …
9989      0.20    4.1028
9990      0.00   15.6332
9991      0.20   19.3932
9992      0.00   13.3200
9993      0.00   72.9480

[9994 rows x 13 columns]
```

Reading Datasets The following code would imply these instructions

df_orders = is the name of the variable, that will be using throughout the example of this tutorial. pd = stands for Panda, it's the convention the community is using. .read_csv = is a method within to read the CSV file.

[ ]:
```
df.sample(9)
```

[ ]:
```
          Ship Mode      Segment        Country           City         State  \
499   Standard Class     Consumer  United States     Costa Mesa    California
1361  Standard Class    Corporate  United States   Philadelphia  Pennsylvania
8530  Standard Class     Consumer  United States        Detroit      Michigan
6469  Standard Class  Home Office  United States     Providence  Rhode Island
5387    Second Class    Corporate  United States    Los Angeles    California
2930        Same Day     Consumer  United States  San Francisco    California
```

```
9877     First Class  Home Office  United States      Cleveland          Ohio
3933  Standard Class     Consumer  United States    Bakersfield    California
119      First Class     Consumer  United States     Wilmington      Delaware

      Postal Code   Region         Category Sub-Category    Sales  Quantity  \
499         92627     West        Furniture  Furnishings   69.300         9
1361        19120     East  Office Supplies        Paper   23.680         4
8530        48227  Central  Office Supplies        Paper   33.360         4
6469         2908     East        Furniture  Furnishings   72.420         6
5387        90045     West       Technology       Phones  167.976         3
2930        94109     West  Office Supplies      Binders    6.608         2
9877        44105     East  Office Supplies      Binders    8.700         5
3933        93309     West  Office Supplies          Art    9.400         5
119         19805     East        Furniture  Furnishings   47.040         3

      Discount   Profit
499        0.0  22.8690
1361       0.2   7.4000
8530       0.0  16.6800
6469       0.0  23.8986
5387       0.2  10.4985
2930       0.2   2.2302
9877       0.7  -6.3800
3933       0.0   2.7260
119        0.0  18.3456
```

[ ]:
```
df.tail()
```

[ ]:
```
           Ship Mode   Segment        Country         City       State  \
9989    Second Class  Consumer  United States        Miami     Florida
9990  Standard Class  Consumer  United States   Costa Mesa  California
9991  Standard Class  Consumer  United States   Costa Mesa  California
9992  Standard Class  Consumer  United States   Costa Mesa  California
9993    Second Class  Consumer  United States  Westminster  California

      Postal Code Region         Category Sub-Category    Sales  Quantity  \
9989        33180  South        Furniture  Furnishings   25.248         3
9990        92627   West        Furniture  Furnishings   91.960         2
9991        92627   West       Technology       Phones  258.576         2
9992        92627   West  Office Supplies        Paper   29.600         4
9993        92683   West  Office Supplies   Appliances  243.160         2

      Discount   Profit
9989       0.2   4.1028
9990       0.0  15.6332
9991       0.2  19.3932
9992       0.0  13.3200
```

```
9993       0.0  72.9480
```

check the missing value

[ ]:
```
df.isnull().sum()
```

[ ]:
```
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
```

Finding Total number of null values in a dataset

[ ]:
```
print("total number of null values = ",df.isnull().sum().sum())
```

```
total number of null values =  0
```

[ ]:
```
print(df.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 13 columns):
 #   Column        Non-Null Count  Dtype
---  ------        --------------  -----
 0   Ship Mode     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   Postal Code   9994 non-null   int64
 6   Region        9994 non-null   object
 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
None
```

Statistical details of the dataset

[ ]:
```
df.describe()
```

[ ]:

	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

shape of the DataSet

[ ]:
```
df.shape
```

[ ]:
```
(9994, 13)
```

Find the dtypes in the Dataset

[ ]:
```
df.dtypes
```

[ ]:
```
Ship Mode        object
Segment          object
Country          object
City             object
State            object
Postal Code       int64
Region           object
Category         object
Sub-Category     object
Sales           float64
Quantity          int64
Discount        float64
Profit          float64
dtype: object
```

[ ]:
```
df.columns
```

[ ]:
```
Index(['Ship Mode', 'Segment', 'Country', 'City', 'State', 'Postal Code',
       'Region', 'Category', 'Sub-Category', 'Sales', 'Quantity', 'Discount',
       'Profit'],
      dtype='object')
```

Statistic Figures The following code would imply these instructions

- df_orders = is the name of the variable, that will be using throughout the example of this tutorial.
- .count = is the count value to a specific column.
- .mean = is the mean value to a specific column.
- .std = is the std value to a specific column.
- .min = is the min value to a specific column.

[ ]:
```
df["Sales"].count()
```

[ ]:
```
9994
```

[ ]:
```
df["Sales"].mean()
```

[ ]:
```
229.85800083049833
```

[ ]:
```
df["Sales"].std()
```

[ ]:
```
623.2451005086807
```

[ ]:
```
df["Sales"].min()
```

[ ]:
```
0.444
```

Exporting Dataset Once that we've satisfied with out results, let's export them a new CSV dataset, so we could work with them on the next notebook.

- df_orders = is the name of the variable, that will be using throughout the example of this tutorial.
- .to_csv = is the export method to a CSV dataset.
- index = False = we need to definet this index value set to False, since we don't want the index column.

[ ]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 13 columns):
 #   Column        Non-Null Count  Dtype
---  ------        --------------  -----
 0   Ship Mode     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   Postal Code   9994 non-null   int64
 6   Region        9994 non-null   object
 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
```

Bonus Stage Now that we've come a long way of exploring our superstore.csv dataset, it's time to dive a little bit deeper of what, both Python and Pandas capable of delivering. Let's try to create a custom class in Python by leveraging our builtin Pandas method available in the library.

[ ]:
```
# Let's create a class named `display_all`, by which later we call on the next␣
↪command.

def display_all(df_orders):
    with pd.option_context("display.max_rows", 20, "display.max_columns", 20):
        display(df_orders)
```

[ ]:
```
display_all(df.head(10).T)
```

	0	1	2	3	\
Ship Mode	Second Class	Second Class	Second Class	Standard Class	
Segment	Consumer	Consumer	Corporate	Consumer	
Country	United States	United States	United States	United States	
City	Henderson	Henderson	Los Angeles	Fort Lauderdale	
State	Kentucky	Kentucky	California	Florida	
Postal Code	42420	42420	90036	33311	
Region	South	South	West	South	
Category	Furniture	Furniture	Office Supplies	Furniture	
Sub-Category	Bookcases	Chairs	Labels	Tables	
Sales	261.96	731.94	14.62	957.5775	
Quantity	2	3	2	5	
Discount	0.0	0.0	0.0	0.45	
Profit	41.9136	219.582	6.8714	-383.031	

	4	5	6	\
Ship Mode	Standard Class	Standard Class	Standard Class	
Segment	Consumer	Consumer	Consumer	
Country	United States	United States	United States	
City	Fort Lauderdale	Los Angeles	Los Angeles	
State	Florida	California	California	
Postal Code	33311	90032	90032	
Region	South	West	West	
Category	Office Supplies	Furniture	Office Supplies	
Sub-Category	Storage	Furnishings	Art	
Sales	22.368	48.86	7.28	
Quantity	2	7	4	

Discount	0.2	0.0	0.0
Profit	2.5164	14.1694	1.9656

	7	8	9
Ship Mode	Standard Class	Standard Class	Standard Class
Segment	Consumer	Consumer	Consumer
Country	United States	United States	United States
City	Los Angeles	Los Angeles	Los Angeles
State	California	California	California
Postal Code	90032	90032	90032
Region	West	West	West
Category	Technology	Office Supplies	Office Supplies
Sub-Category	Phones	Binders	Appliances
Sales	907.152	18.504	114.9
Quantity	6	3	5
Discount	0.2	0.2	0.0
Profit	90.7152	5.7825	34.47

[ ]:
```
display_all(df.describe(include='all').T)
```

	count	unique	top	freq	mean	std
Ship Mode	9994	4	Standard Class	5968	NaN	NaN
Segment	9994	3	Consumer	5191	NaN	NaN
Country	9994	1	United States	9994	NaN	NaN
City	9994	531	New York City	915	NaN	NaN
State	9994	49	California	2001	NaN	NaN
Postal Code	9994.0	NaN	NaN	NaN	55190.379428	32063.69335
Region	9994	4	West	3203	NaN	NaN
Category	9994	3	Office Supplies	6026	NaN	NaN
Sub-Category	9994	17	Binders	1523	NaN	NaN
Sales	9994.0	NaN	NaN	NaN	229.858001	623.245101
Quantity	9994.0	NaN	NaN	NaN	3.789574	2.22511
Discount	9994.0	NaN	NaN	NaN	0.156203	0.206452
Profit	9994.0	NaN	NaN	NaN	28.656896	234.260108

	min	25%	50%	75%	max
Ship Mode	NaN	NaN	NaN	NaN	NaN
Segment	NaN	NaN	NaN	NaN	NaN
Country	NaN	NaN	NaN	NaN	NaN
City	NaN	NaN	NaN	NaN	NaN
State	NaN	NaN	NaN	NaN	NaN
Postal Code	1040.0	23223.0	56430.5	90008.0	99301.0
Region	NaN	NaN	NaN	NaN	NaN
Category	NaN	NaN	NaN	NaN	NaN
Sub-Category	NaN	NaN	NaN	NaN	NaN
Sales	0.444	17.28	54.49	209.94	22638.48
Quantity	1.0	2.0	3.0	5.0	14.0
Discount	0.0	0.0	0.2	0.2	0.8

```
Profit       -6599.978  1.72875   8.6665   29.364  8399.976
```

Data Cleaning

[ ]:
```
# Set the total rows and columns default numbers.
pd.set_option('display.max_columns', 21)
pd.set_option('display.max_rows', 5)
```

[ ]:
```
Cat = [i for i in df.columns if df.dtypes[i] == 'object']
for j in Cat:
    print('\033[95m' + j + '\033[0m')
    print(sorted(df[j].unique()))
```

```
Ship Mode
['First Class', 'Same Day', 'Second Class', 'Standard Class']
Segment
['Consumer', 'Corporate', 'Home Office']
Country
['United States']
City
['Aberdeen', 'Abilene', 'Akron', 'Albuquerque', 'Alexandria', 'Allen',
'Allentown', 'Altoona', 'Amarillo', 'Anaheim', 'Andover', 'Ann Arbor',
'Antioch', 'Apopka', 'Apple Valley', 'Appleton', 'Arlington', 'Arlington
Heights', 'Arvada', 'Asheville', 'Athens', 'Atlanta', 'Atlantic City', 'Auburn',
'Aurora', 'Austin', 'Avondale', 'Bakersfield', 'Baltimore', 'Bangor',
'Bartlett', 'Bayonne', 'Baytown', 'Beaumont', 'Bedford', 'Belleville',
'Bellevue', 'Bellingham', 'Bethlehem', 'Beverly', 'Billings', 'Bloomington',
'Boca Raton', 'Boise', 'Bolingbrook', 'Bossier City', 'Bowling Green', 'Boynton
Beach', 'Bozeman', 'Brentwood', 'Bridgeton', 'Bristol', 'Broken Arrow',
'Broomfield', 'Brownsville', 'Bryan', 'Buffalo', 'Buffalo Grove', 'Bullhead
City', 'Burbank', 'Burlington', 'Caldwell', 'Camarillo', 'Cambridge', 'Canton',
'Carlsbad', 'Carol Stream', 'Carrollton', 'Cary', 'Cedar Hill', 'Cedar Rapids',
'Champaign', 'Chandler', 'Chapel Hill', 'Charlotte', 'Charlottesville',
'Chattanooga', 'Chesapeake', 'Chester', 'Cheyenne', 'Chicago', 'Chico', 'Chula
Vista', 'Cincinnati', 'Citrus Heights', 'Clarksville', 'Cleveland', 'Clifton',
'Clinton', 'Clovis', 'Coachella', 'College Station', 'Colorado Springs',
'Columbia', 'Columbus', 'Commerce City', 'Concord', 'Conroe', 'Conway', 'Coon
Rapids', 'Coppell', 'Coral Gables', 'Coral Springs', 'Corpus Christi', 'Costa
Mesa', 'Cottage Grove', 'Covington', 'Cranston', 'Cuyahoga Falls', 'Dallas',
'Danbury', 'Danville', 'Davis', 'Daytona Beach', 'Dearborn', 'Dearborn Heights',
'Decatur', 'Deer Park', 'Delray Beach', 'Deltona', 'Denver', 'Des Moines', 'Des
Plaines', 'Detroit', 'Dover', 'Draper', 'Dublin', 'Dubuque', 'Durham', 'Eagan',
'East Orange', 'East Point', 'Eau Claire', 'Edinburg', 'Edmond', 'Edmonds', 'El
Cajon', 'El Paso', 'Elkhart', 'Elmhurst', 'Elyria', 'Encinitas', 'Englewood',
'Escondido', 'Eugene', 'Evanston', 'Everett', 'Fairfield', 'Fargo',
'Farmington', 'Fayetteville', 'Florence', 'Fort Collins', 'Fort Lauderdale',
'Fort Worth', 'Frankfort', 'Franklin', 'Freeport', 'Fremont', 'Fresno',
'Frisco', 'Gaithersburg', 'Garden City', 'Garland', 'Gastonia', 'Georgetown',
```

'Gilbert', 'Gladstone', 'Glendale', 'Glenview', 'Goldsboro', 'Grand Island', 'Grand Prairie', 'Grand Rapids', 'Grapevine', 'Great Falls', 'Greeley', 'Green Bay', 'Greensboro', 'Greenville', 'Greenwood', 'Gresham', 'Grove City', 'Gulfport', 'Hackensack', 'Hagerstown', 'Haltom City', 'Hamilton', 'Hampton', 'Harlingen', 'Harrisonburg', 'Hattiesburg', 'Helena', 'Hempstead', 'Henderson', 'Hendersonville', 'Hesperia', 'Hialeah', 'Hickory', 'Highland Park', 'Hillsboro', 'Holland', 'Hollywood', 'Holyoke', 'Homestead', 'Hoover', 'Hot Springs', 'Houston', 'Huntington Beach', 'Huntsville', 'Independence', 'Indianapolis', 'Inglewood', 'Iowa City', 'Irving', 'Jackson', 'Jacksonville', 'Jamestown', 'Jefferson City', 'Johnson City', 'Jonesboro', 'Jupiter', 'Keller', 'Kenner', 'Kenosha', 'Kent', 'Kirkwood', 'Kissimmee', 'Knoxville', 'La Crosse', 'La Mesa', 'La Porte', 'La Quinta', 'Lafayette', 'Laguna Niguel', 'Lake Charles', 'Lake Elsinore', 'Lake Forest', 'Lakeland', 'Lakeville', 'Lakewood', 'Lancaster', 'Lansing', 'Laredo', 'Las Cruces', 'Las Vegas', 'Laurel', 'Lawrence', 'Lawton', 'Layton', 'League City', 'Lebanon', 'Lehi', 'Leominster', 'Lewiston', 'Lincoln Park', 'Linden', 'Lindenhurst', 'Little Rock', 'Littleton', 'Lodi', 'Logan', 'Long Beach', 'Longmont', 'Longview', 'Lorain', 'Los Angeles', 'Louisville', 'Loveland', 'Lowell', 'Lubbock', 'Macon', 'Madison', 'Malden', 'Manchester', 'Manhattan', 'Mansfield', 'Manteca', 'Maple Grove', 'Margate', 'Marietta', 'Marion', 'Marlborough', 'Marysville', 'Mason', 'Mcallen', 'Medford', 'Medina', 'Melbourne', 'Memphis', 'Mentor', 'Meriden', 'Meridian', 'Mesa', 'Mesquite', 'Miami', 'Middletown', 'Midland', 'Milford', 'Milwaukee', 'Minneapolis', 'Miramar', 'Mishawaka', 'Mission Viejo', 'Missoula', 'Missouri City', 'Mobile', 'Modesto', 'Monroe', 'Montebello', 'Montgomery', 'Moorhead', 'Moreno Valley', 'Morgan Hill', 'Morristown', 'Mount Pleasant', 'Mount Vernon', 'Murfreesboro', 'Murray', 'Murrieta', 'Muskogee', 'Naperville', 'Nashua', 'Nashville', 'New Albany', 'New Bedford', 'New Brunswick', 'New Castle', 'New Rochelle', 'New York City', 'Newark', 'Newport News', 'Niagara Falls', 'Noblesville', 'Norfolk', 'Normal', 'Norman', 'North Charleston', 'North Las Vegas', 'North Miami', 'Norwich', 'Oak Park', 'Oakland', 'Oceanside', 'Odessa', 'Oklahoma City', 'Olathe', 'Olympia', 'Omaha', 'Ontario', 'Orange', 'Orem', 'Orland Park', 'Orlando', 'Ormond Beach', 'Oswego', 'Overland Park', 'Owensboro', 'Oxnard', 'Palatine', 'Palm Coast', 'Park Ridge', 'Parker', 'Parma', 'Pasadena', 'Pasco', 'Passaic', 'Paterson', 'Pearland', 'Pembroke Pines', 'Pensacola', 'Peoria', 'Perth Amboy', 'Pharr', 'Philadelphia', 'Phoenix', 'Pico Rivera', 'Pine Bluff', 'Plainfield', 'Plano', 'Plantation', 'Pleasant Grove', 'Pocatello', 'Pomona', 'Pompano Beach', 'Port Arthur', 'Port Orange', 'Port Saint Lucie', 'Portage', 'Portland', 'Providence', 'Provo', 'Pueblo', 'Quincy', 'Raleigh', 'Rancho Cucamonga', 'Rapid City', 'Reading', 'Redding', 'Redlands', 'Redmond', 'Redondo Beach', 'Redwood City', 'Reno', 'Renton', 'Revere', 'Richardson', 'Richmond', 'Rio Rancho', 'Riverside', 'Rochester', 'Rochester Hills', 'Rock Hill', 'Rockford', 'Rockville', 'Rogers', 'Rome', 'Romeoville', 'Roseville', 'Roswell', 'Round Rock', 'Royal Oak', 'Sacramento', 'Saginaw', 'Saint Charles', 'Saint Cloud', 'Saint Louis', 'Saint Paul', 'Saint Peters', 'Saint Petersburg', 'Salem', 'Salinas', 'Salt Lake City', 'San Angelo', 'San Antonio', 'San Bernardino', 'San Clemente', 'San Diego', 'San Francisco', 'San Gabriel', 'San Jose', 'San Luis Obispo', 'San Marcos', 'San Mateo', 'Sandy Springs', 'Sanford', 'Santa Ana', 'Santa Barbara', 'Santa Clara',

```
'Santa Fe', 'Santa Maria', 'Scottsdale', 'Seattle', 'Sheboygan', 'Shelton',
'Sierra Vista', 'Sioux Falls', 'Skokie', 'Smyrna', 'South Bend', 'Southaven',
'Sparks', 'Spokane', 'Springdale', 'Springfield', 'Sterling Heights',
'Stockton', 'Suffolk', 'Summerville', 'Sunnyvale', 'Superior', 'Tallahassee',
'Tamarac', 'Tampa', 'Taylor', 'Temecula', 'Tempe', 'Texarkana', 'Texas City',
'The Colony', 'Thomasville', 'Thornton', 'Thousand Oaks', 'Tigard', 'Tinley
Park', 'Toledo', 'Torrance', 'Trenton', 'Troy', 'Tucson', 'Tulsa', 'Tuscaloosa',
'Twin Falls', 'Tyler', 'Urbandale', 'Utica', 'Vacaville', 'Vallejo',
'Vancouver', 'Vineland', 'Virginia Beach', 'Visalia', 'Waco', 'Warner Robins',
'Warwick', 'Washington', 'Waterbury', 'Waterloo', 'Watertown', 'Waukesha',
'Wausau', 'Waynesboro', 'West Allis', 'West Jordan', 'West Palm Beach',
'Westfield', 'Westland', 'Westminster', 'Wheeling', 'Whittier', 'Wichita',
'Wilmington', 'Wilson', 'Woodbury', 'Woodland', 'Woodstock', 'Woonsocket',
'Yonkers', 'York', 'Yucaipa', 'Yuma']
State
['Alabama', 'Arizona', 'Arkansas', 'California', 'Colorado', 'Connecticut',
'Delaware', 'District of Columbia', 'Florida', 'Georgia', 'Idaho', 'Illinois',
'Indiana', 'Iowa', 'Kansas', 'Kentucky', 'Louisiana', 'Maine', 'Maryland',
'Massachusetts', 'Michigan', 'Minnesota', 'Mississippi', 'Missouri', 'Montana',
'Nebraska', 'Nevada', 'New Hampshire', 'New Jersey', 'New Mexico', 'New York',
'North Carolina', 'North Dakota', 'Ohio', 'Oklahoma', 'Oregon', 'Pennsylvania',
'Rhode Island', 'South Carolina', 'South Dakota', 'Tennessee', 'Texas', 'Utah',
'Vermont', 'Virginia', 'Washington', 'West Virginia', 'Wisconsin', 'Wyoming']
Region
['Central', 'East', 'South', 'West']
Category
['Furniture', 'Office Supplies', 'Technology']
Sub-Category
['Accessories', 'Appliances', 'Art', 'Binders', 'Bookcases', 'Chairs',
'Copiers', 'Envelopes', 'Fasteners', 'Furnishings', 'Labels', 'Machines',
'Paper', 'Phones', 'Storage', 'Supplies', 'Tables']
```

[ ]:
```
df.dtypes.to_frame()
```

[ ]:
```
                0
Ship Mode   object
Segment     object
…            …
Discount   float64
Profit     float64

[13 rows x 1 columns]
```

[ ]:
```
palette_color = sns.color_palette("flare")

sns.palplot(palette_color, size = 3)
```

```
plt.text(-0.75,-0.75, "Superstore's Retails: Visualizations &␣
 ↪Classifications",{'font':'serif', 'size':24, 'weight':'bold'})
plt.text(-0.75,-0.65, 'Lets try to stick to these colors throughout␣
 ↪presentation.',{'font':'serif', 'size':16},alpha = 0.9)
plt.gcf().set_facecolor('#f5f6f6')
plt.gcf().set_dpi(100)
plt.box(None)
plt.axis('off')
plt.show()
```

[ ]:
```
df.describe()
```

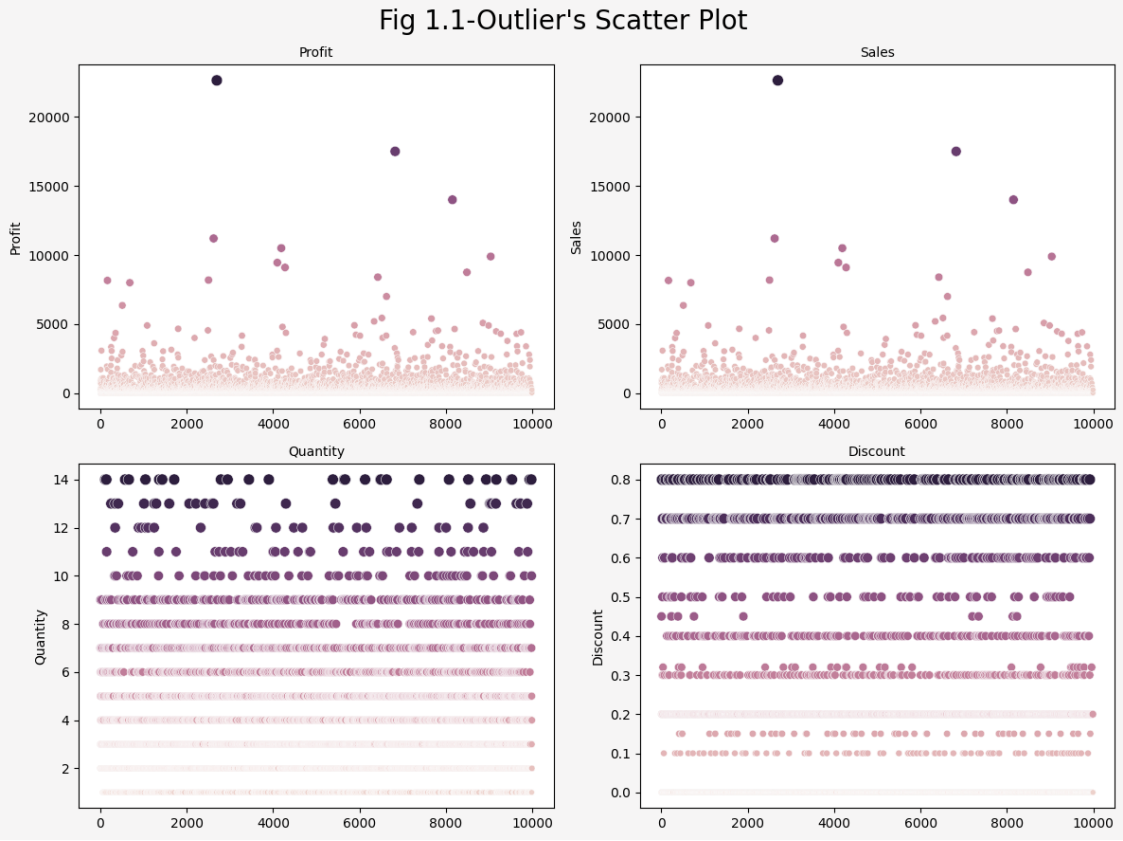
[ ]:

	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
…	…	…	…	…	…
75%	90008.000000	209.940000	5.000000	0.200000	29.364000
max	99301.000000	22638.480000	14.000000	0.800000	8399.976000

[8 rows x 5 columns]

[ ]:
```
fig, axs = plt.subplots(nrows = 2, ncols = 2, figsize=(12, 9));
fig.patch.set_facecolor('#f6f5f5')

# Create a new column named `Profit`
df['Profit'] = df['Sales'] - df['Discount']

sns.scatterplot(data=df, y="Profit", x = df.index, ax = axs[0][0],hue =␣
 ↪"Profit", size = "Profit", legend=False)
axs[0][0].set_title('Profit', fontsize = 10)

sns.scatterplot(data=df, y ="Sales", x = df.index, ax = axs[0][1], hue␣
 ↪="Sales", size = "Sales" , legend=False)
axs[0][1].set_title('Sales', fontsize = 10)

sns.scatterplot(data=df, y = "Quantity", x = df.index, ax = axs[1][0], hue =␣
 ↪"Quantity", size = "Quantity", legend=False)
```

```
axs[1][0].set_title('Quantity', fontsize = 10)

sns.scatterplot(data= df, y = "Discount", x = df.index, ax = axs[1][1], hue =␣
 ↪"Discount", size = "Discount", legend=False)
axs[1][1].set_title('Discount', fontsize = 10)
plt.suptitle("Fig 1.1-Outlier's Scatter Plot",fontsize = 20)

plt.tight_layout()
plt.show()
```



Find the covariance of dataset

[ ]:
```
df.cov()
```

```
<ipython-input-38-6f98a29763d5>:1: FutureWarning: The default value of
numeric_only in DataFrame.cov is deprecated. In a future version, it will
default to False. Select only valid columns or specify the value of numeric_only
to silence this warning.
  df.cov()
```

[ ]:
```
               Postal Code          Sales    Quantity    Discount  \
Postal Code  1.028080e+09 -476682.766590  910.415885  386.870404
Sales       -4.766828e+05  388434.455308  278.459923   -3.627228
Quantity     9.104159e+02     278.459923    4.951113    0.003961
Discount     3.868704e+02      -3.627228    0.003961    0.042622
Profit      -4.770696e+05  388438.082536  278.455961   -3.669851

                     Profit
Postal Code -477069.636994
Sales        388438.082536
Quantity        278.455961
Discount         -3.669851
Profit       388441.752387
```

Find the Series containing counts of unique values

[ ]:
```
df.value_counts()
```

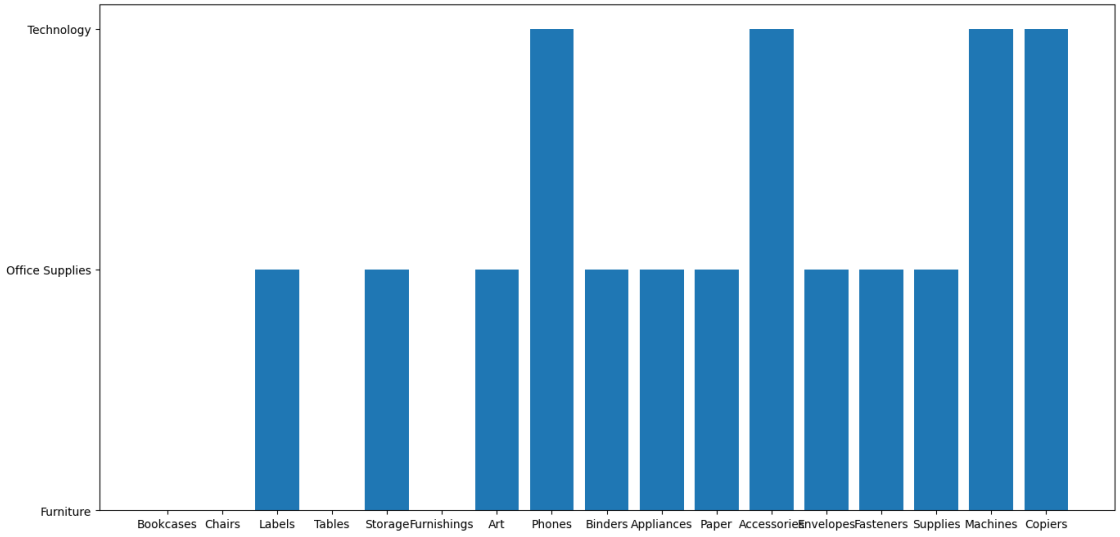
[ ]:
```
Ship Mode       Segment      Country        City          State       Postal Code
Region   Category         Sub-Category  Sales    Quantity  Discount  Profit
Second Class    Consumer     United States  Seattle       Washington  98115
West     Office Supplies  Paper         12.960   2         0.0       12.960
2
                Corporate    United States  Chicago       Illinois    60653
Central  Office Supplies  Binders       3.564    3         0.8       2.764
2
 ..
                                            Little Rock   Arkansas    72209
South    Office Supplies  Storage       62.040   4         0.0       62.040
1
Standard Class  Home Office  United States  Yuma          Arizona     85364
West     Technology       Machines      599.985  5         0.7       599.285
1
Length: 9971, dtype: int64
```

Deleting the Variable

[ ]:
```
col=['Postal Code']
df1=df.drop(columns=col,axis=1)
```

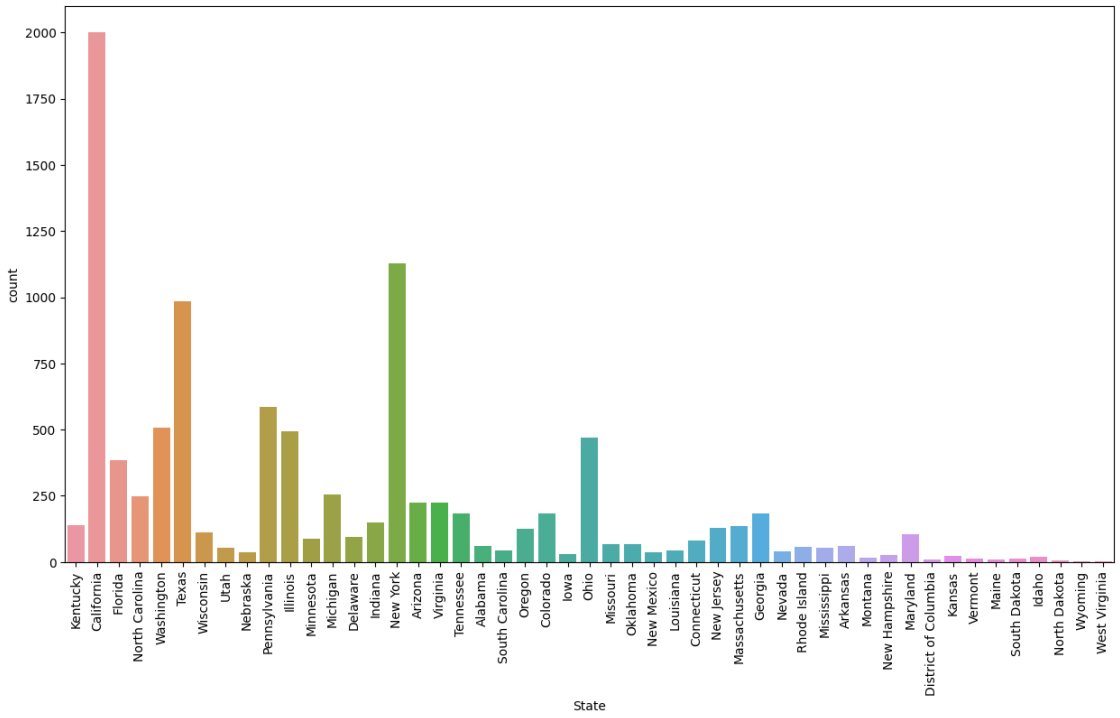
Proper Visualization of the data set

[ ]:
```
plt.figure(figsize=(16,8))
plt.bar('Sub-Category','Category', data=df)
plt.show()
```



[ ]:
```
print(df1['State'].value_counts())
plt.figure(figsize=(15,8))
sns.countplot(x=df1['State'])
plt.xticks(rotation=90)
plt.show()
```

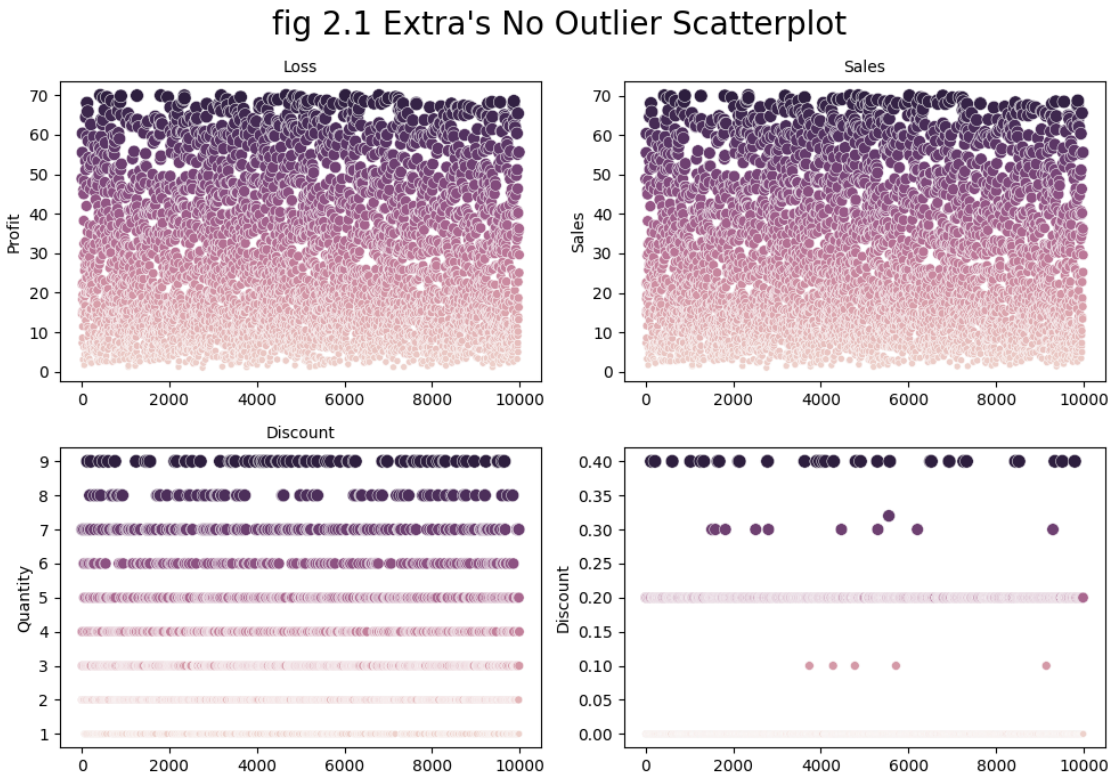
```
California       2001
New York         1128
                 …
West Virginia       4
Wyoming             1
Name: State, Length: 49, dtype: int64
```



[ ]:
```
data_outliers = df[df["Profit"]<70]; data_outliers =␣
 ↪data_outliers[data_outliers["Profit"]>-40]
data_outliers = data_outliers[df["Sales"]<498.93] ; data_outliers =␣
 ↪data_outliers[data_outliers["Quantity"]<9.5]
data_outliers = data_outliers[data_outliers["Discount"]<0.5]


print(f"The total number of records containing outliers = {df.
 ↪shape[0]-data_outliers.shape[0]}\nThe outliers forms {round((df.
 ↪shape[0]-data_outliers.shape[0])/df.shape[0]*100,2)}% of superstore's␣
 ↪dataset")

fig, axs = plt.subplots(nrows = 2, ncols = 2, figsize=(10, 7));

sns.scatterplot(data=data_outliers, y = "Profit", x = data_outliers.index, ax =␣
 ↪axs[0][0], hue="Profit", size="Profit", legend=False)
sns.scatterplot(data=data_outliers, y = "Sales", x = data_outliers.index, ax =␣
 ↪axs[0][1], hue ="Sales", size ="Sales", legend=False)
sns.scatterplot(data=data_outliers,y = "Quantity",x = data_outliers.index, ax =␣
 ↪axs[1][0], hue = "Quantity", size ="Quantity", legend=False)
sns.scatterplot(data=data_outliers,y = "Discount",x = data_outliers.index, ax =␣
 ↪axs[1][1], hue = "Discount", size = "Discount", legend=False)
axs[0][0].set_title('Loss', fontsize = 10)
```

```
axs[0][1].set_title('Sales', fontsize = 10)
axs[1][0].set_title('Quantity', fontsize = 10)
axs[1][0].set_title('Discount', fontsize = 10)
plt.suptitle("fig 2.1 Extra's No Outlier Scatterplot",fontsize=20)
plt.tight_layout()
```

```
<ipython-input-49-ed00e5a4e069>:2: UserWarning: Boolean Series key will be
reindexed to match DataFrame index.
  data_outliers = data_outliers[df["Sales"]<498.93] ; data_outliers =
data_outliers[data_outliers["Quantity"]<9.5]

The total number of records containing outliers = 5245
The outliers forms 52.48% of superstore's dataset
```



Data Reduction

Data Reduction Dropping the variable 'Country' via attribute dimensionality reduction, because it contains 100% identical values of "United States" for all records. 'Postal Code' as we wont be using it in our data analysis as we also have the variabl e of "citys"

[ ]:
```
clean_data=df.drop(['Country','Postal Code'], axis=1)
```

Treating duplicate values

Duplicated rows or records can now by dropped from the dataset, as this redundancy may cause inaccurate results and outcomes (an assumption on the dataset).

[ ]:
```
print(f"Number of duplicate recocrds present in the dataset = {clean_data.
↪duplicated().sum()}")
```

```
Number of duplicate recocrds present in the dataset = 63
```

How does Sales, Quantity and Discount affects SuperStores Profits?

We can check this out by ploting a heatmap showcasing the corealation between each numrical columns across we won't be including the feature we added in the dataframe for this correlation.

[ ]:
```
fig = px.scatter(clean_data,x="Profit",y="Sales",color="Discount",
                 size="Quantity",symbol="Segment",title="How diffrent factors␣
↪affects Superstore's sales ")
fig.update_layout(height=600, width=800,
                  legend=dict(yanchor="top", y=0.99,
                              xanchor="left", x=0.01))
fig.show()
```

Profit/ Loss, Sales and Discounts diversity by States! Now We're going to dive into the state wise analysis:

We're going to see the sales in every state We're going to see the Loss ccured in every state We're also going to if the state which causes the most is also causing more loss in average? We're going to give the discounts given off in every state

[ ]:
```
state_s=clean_data.groupby(["State","City","Category","Sub-Category"])["Sales"].
↪sum().reset_index()
fig = px.treemap(state_s, path=["State","City","Category","Sub-Category"],␣
↪values='Sales',
                 color='Sales',
                 color_continuous_scale='RdBu',
                 color_continuous_midpoint=np.average(state_s['Sales'],␣
↪weights=state_s['Sales']),
                 title="Fig 4.1-Total State/City X Sales With Category &␣
↪Sub-Category Distribution")
fig.data[0].textinfo = 'label+text+value'

fig.update_layout(margin = dict(t=50, l=25, r=25, b=25))
fig.show()
```

[ ]:
```
state_profit=clean_data[clean_data.Profit>0]
state_profit=state_profit.
↪groupby(["State","City","Category","Sub-Category"])["Profit"].sum().
↪reset_index()
fig = px.treemap(state_profit, path=["State","City","Category","Sub-Category"],␣
↪values='Profit',
```

```
                  color='Profit',
                  color_continuous_scale='RdBu',
                  color_continuous_midpoint=np.average(clean_data['Profit'],␣
 ↪weights=clean_data['Profit']),
                title="Fig 4.2-Total State/City X Profit With Category &␣
 ↪Sub-Category Distribution")
fig.data[0].textinfo = 'label+text+value'
fig.layout.hovermode = False
fig.update_layout(margin = dict(t=50, l=25, r=25, b=25))
fig.show()
```

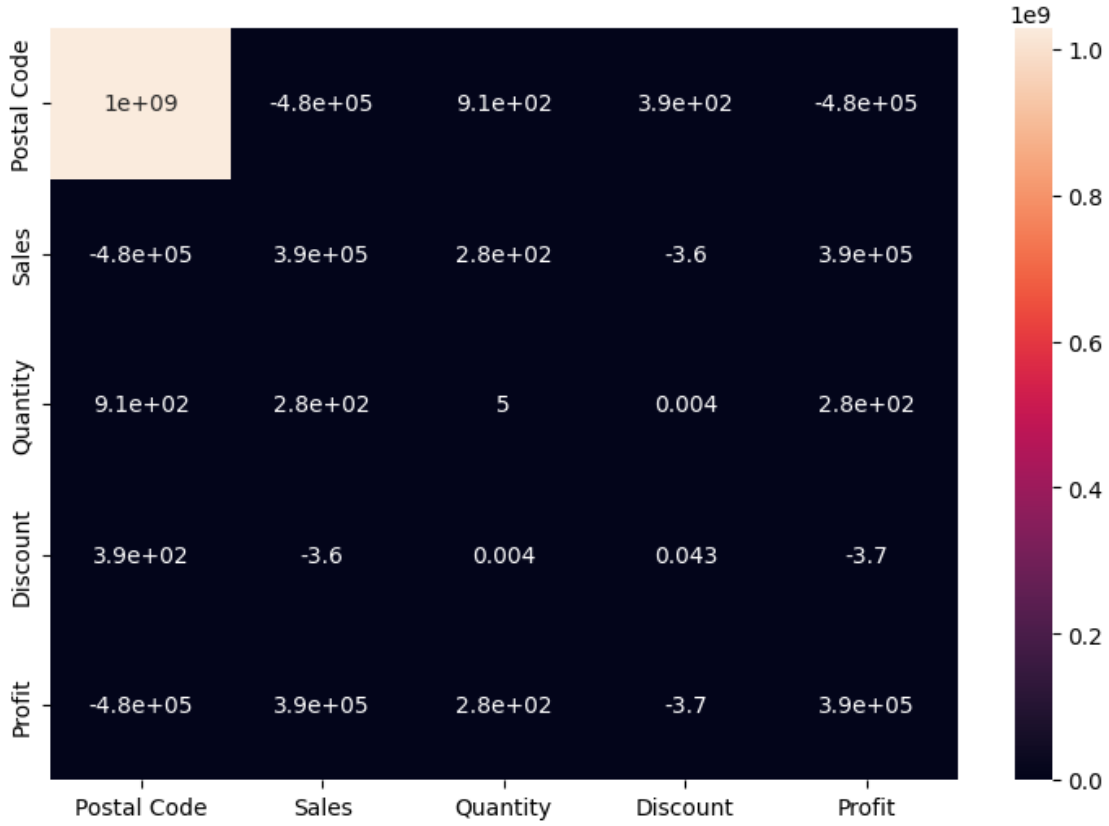
HeatMap of DataSet

[ ]:
```
fig,axes = plt.subplots(1,1,figsize=(9,6))
sns.heatmap(df.corr(), annot= True)
plt.show()
```

```
<ipython-input-62-7d19acb0cf78>:2: FutureWarning:

The default value of numeric_only in DataFrame.corr is deprecated. In a future
version, it will default to False. Select only valid columns or specify the
value of numeric_only to silence this warning.
```

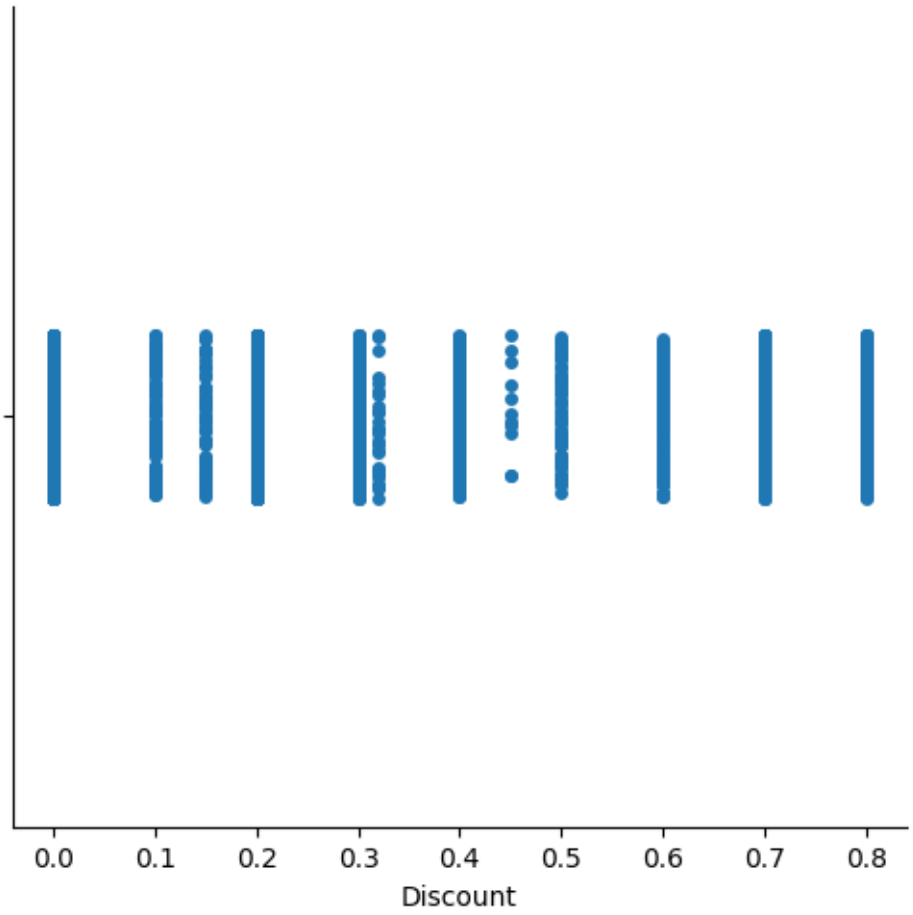
[ ]:
```
fig,axes = plt.subplots(1,1,figsize=(9,6))
sns.heatmap(df.cov(), annot= True)
plt.show()
```

```
<ipython-input-63-d3c100f5e8f3>:2: FutureWarning:

The default value of numeric_only in DataFrame.cov is deprecated. In a future
version, it will default to False. Select only valid columns or specify the
value of numeric_only to silence this warning.
```



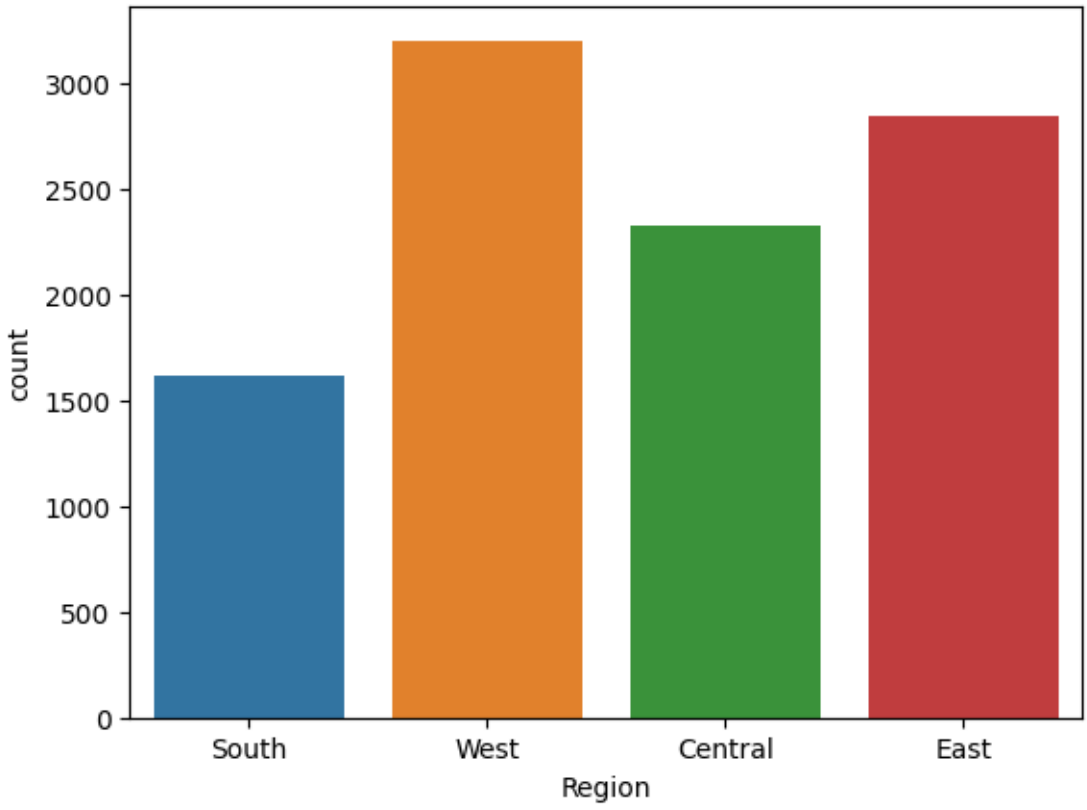
[ ]:
```
sns.catplot(x=df['Discount'])
```

[ ]:
```
<seaborn.axisgrid.FacetGrid at 0x7d0fd6b57640>
```



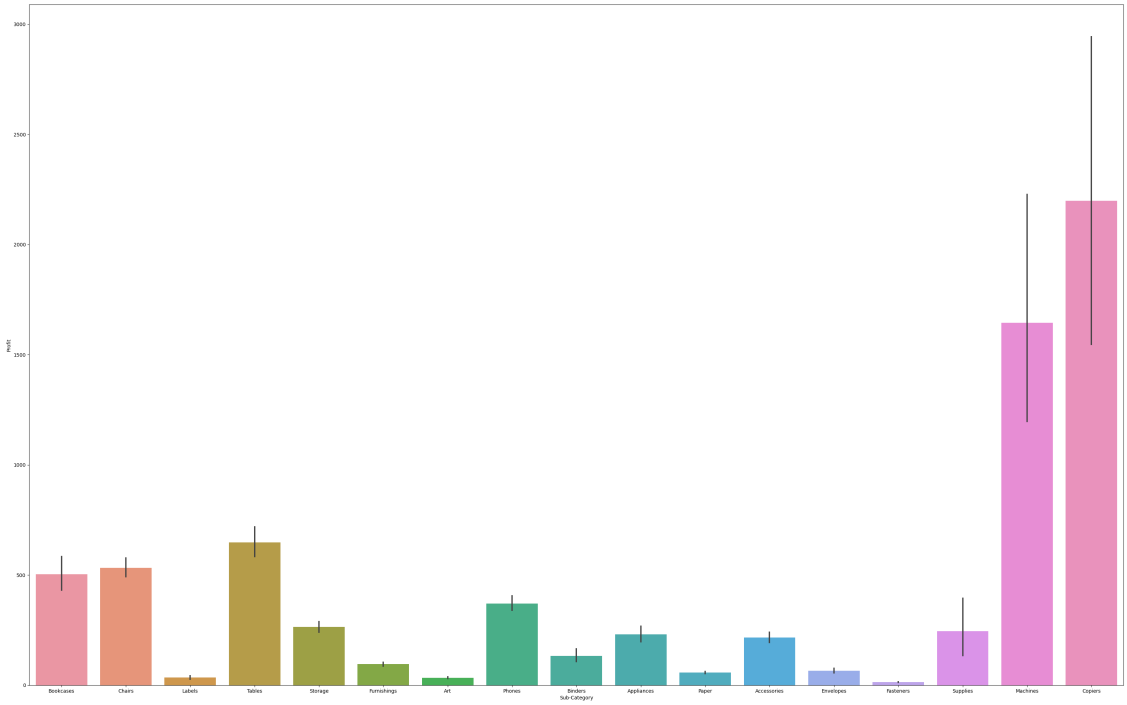
[ ]:
```
sns.countplot(x=df['Region'])
```

[ ]:
```
<Axes: xlabel='Region', ylabel='count'>
```

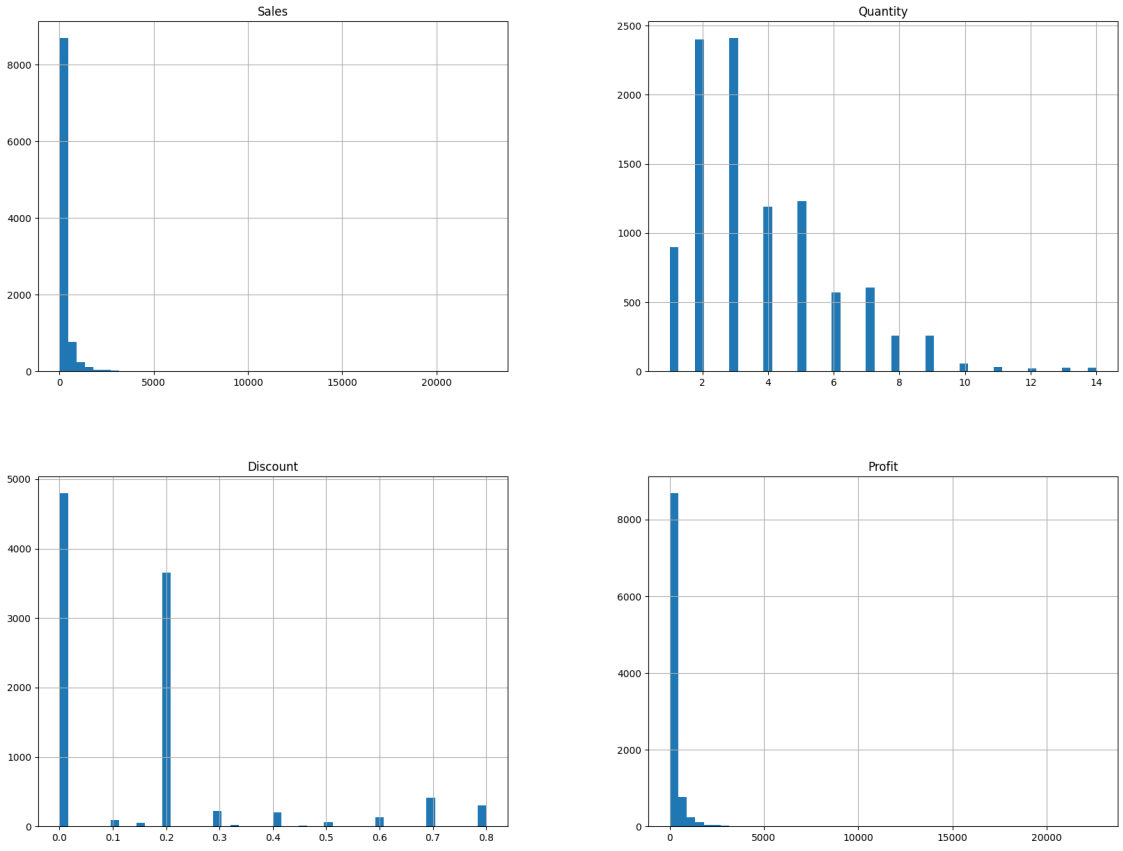


[ ]:
```
plt.figure(figsize=(40,25))
sns.barplot(x=df['Sub-Category'], y=df['Profit'])
```

[ ]: <Axes: xlabel='Sub-Category', ylabel='Profit'>

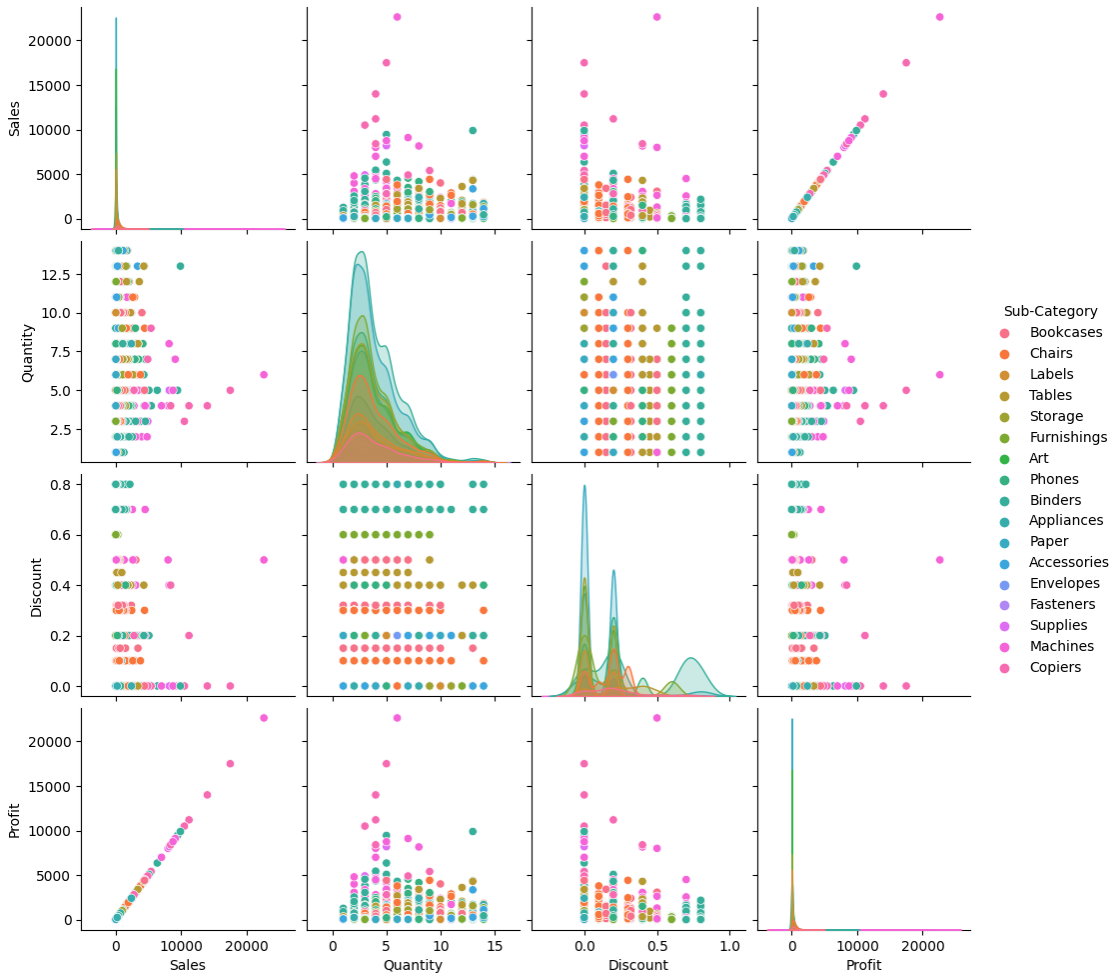


[ ]:
```
df1.hist(bins=50 ,figsize=(20,15))
plt.show()
```



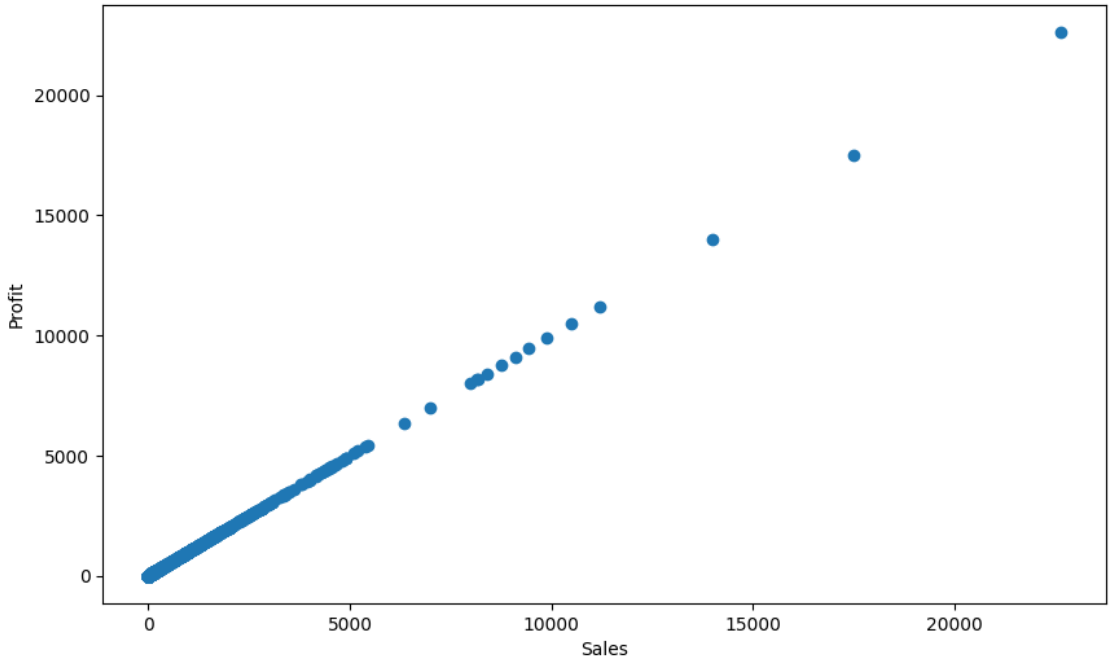
[77]:
```
figsize=(10,10)
sns.pairplot(df1,hue='Sub-Category')
```

[77]: <seaborn.axisgrid.PairGrid at 0x7d0fd1b194b0>



SCATTER PLOT

[78]:
```
fig, ax = plt.subplots(figsize = (10 , 6))
ax.scatter(df["Sales"] , df["Profit"])
ax.set_xlabel('Sales')
ax.set_ylabel('Profit')
plt.show()
```

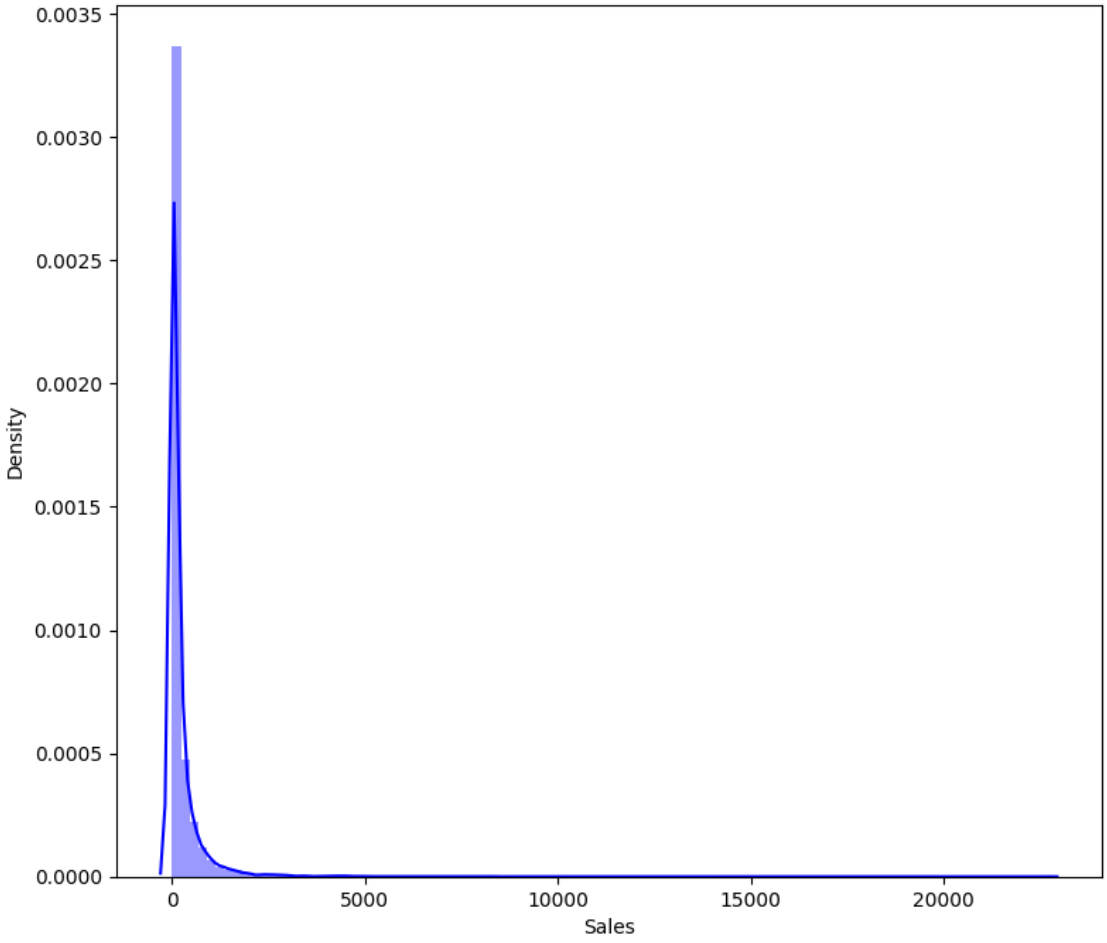


Distribution Plot

[79]:
```
print(df['Sales'].describe())
plt.figure(figsize = (9 , 8))
sns.distplot(df['Sales'], color = 'b', bins = 100, hist_kws = {'alpha': 0.4});
```

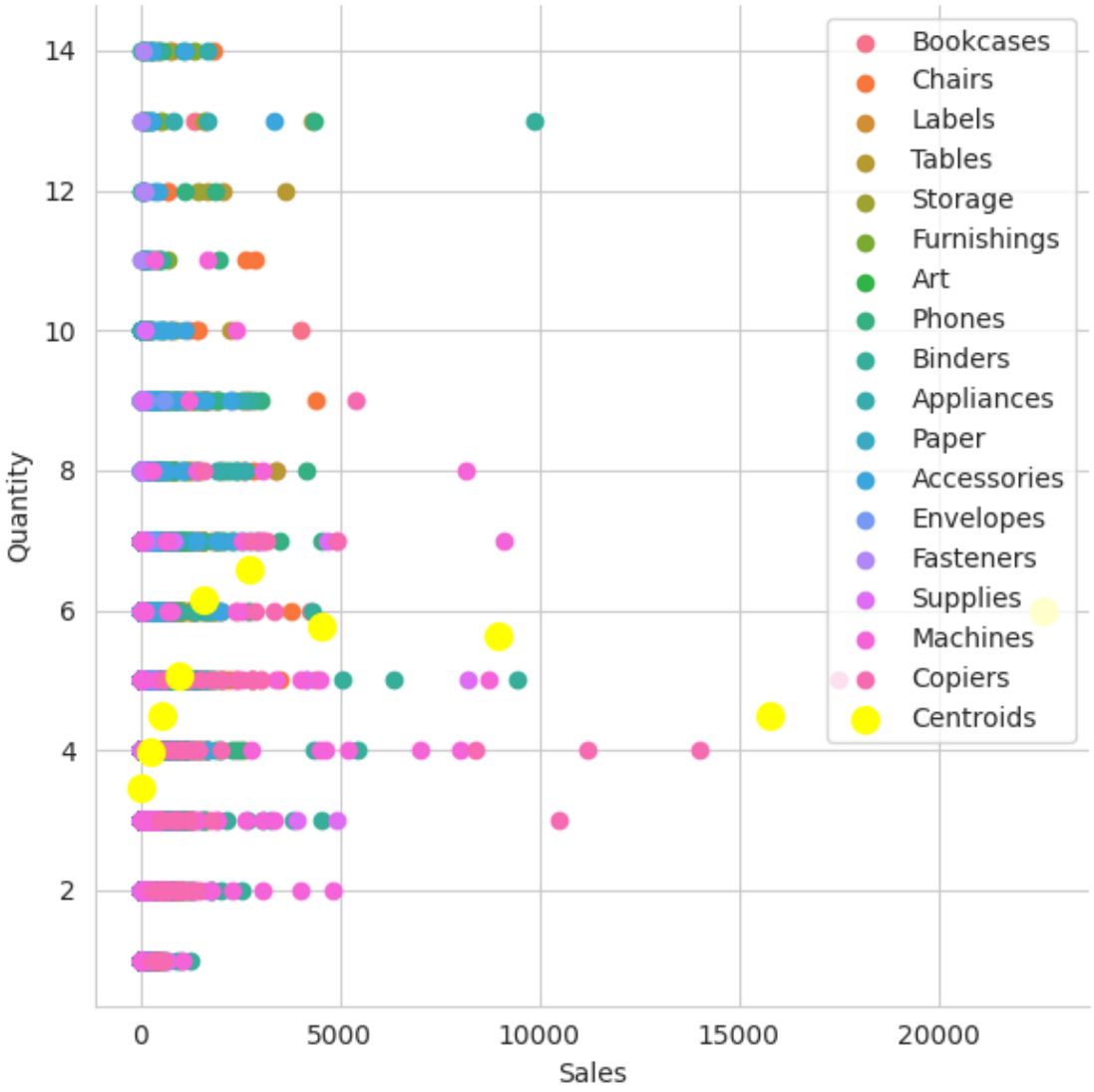
```
count     9994.000000
mean       229.858001
             …
75%        209.940000
max      22638.480000
Name: Sales, Length: 8, dtype: float64
```

```
<ipython-input-79-036949231e31>:3: UserWarning:


`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with
similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751
```



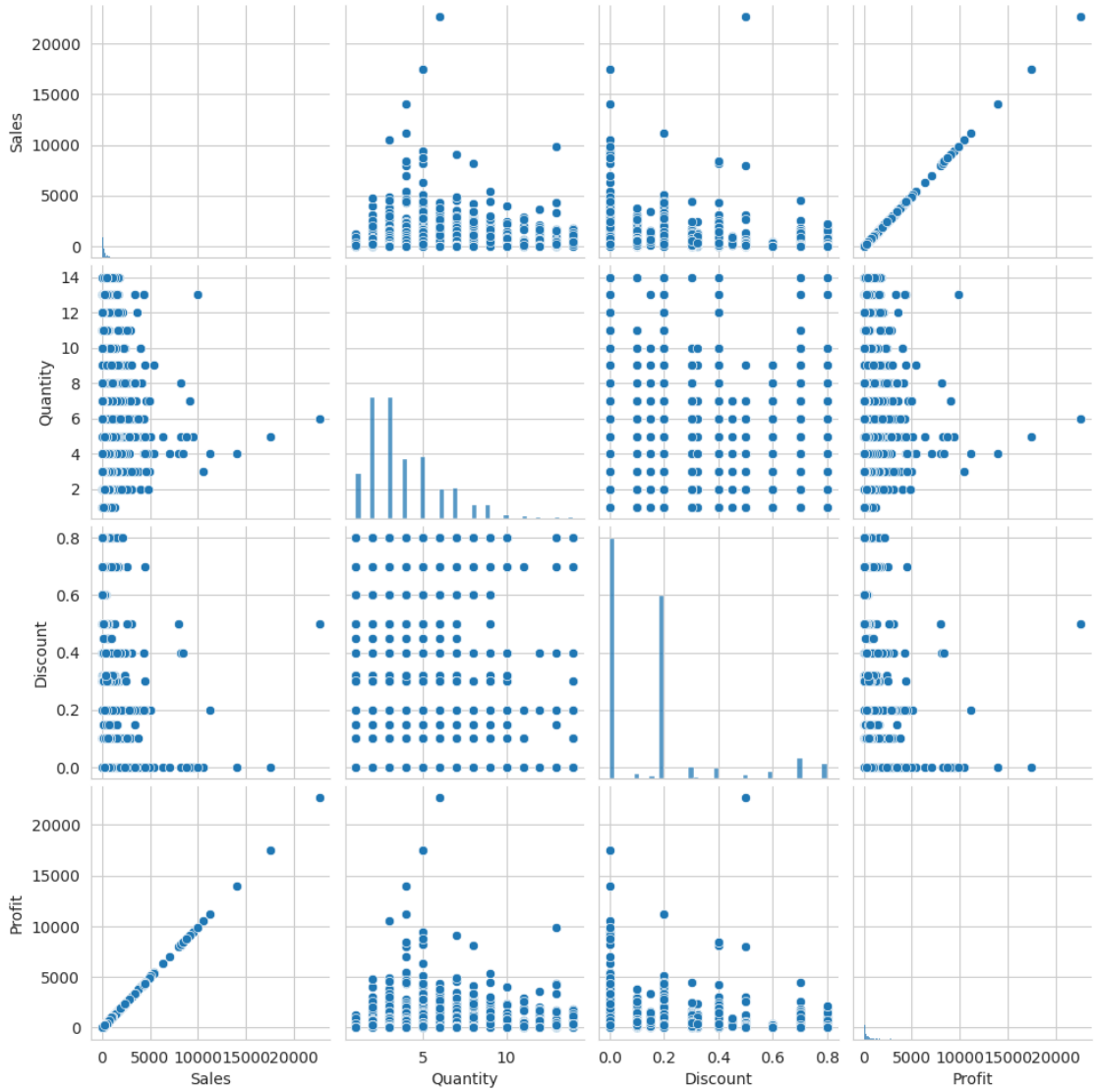
[80]:
```
x = df.iloc[:, [9, 10, 11, 12]].values
from sklearn.cluster import KMeans
wcss = []

for i in range(1, 11):
    kmeans = KMeans(n_clusters = i, init = 'k-means++',
                    max_iter = 300, n_init = 10, random_state = 0).fit(x)
    wcss.append(kmeans.inertia_)

sns.set_style("whitegrid")
sns.FacetGrid(df, hue ="Sub-Category",height = 6).map(plt.
 ↪scatter,'Sales','Quantity')
plt.scatter(kmeans.cluster_centers_[:, 0], kmeans.cluster_centers_[:,1],
            s = 100, c = 'yellow', label = 'Centroids')

plt.legend()
```

[80]: <matplotlib.legend.Legend at 0x7d0fd0ecb070>



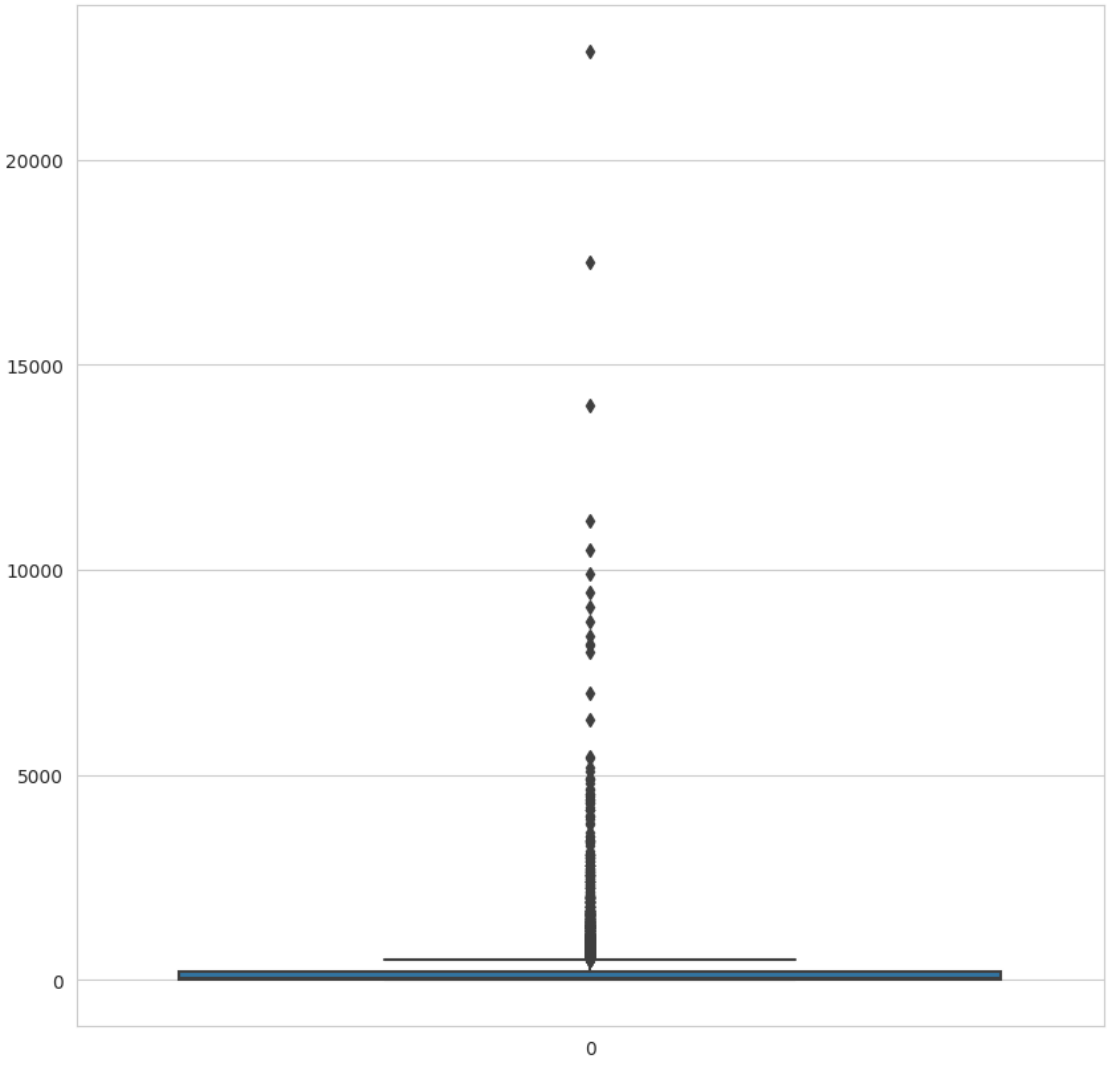
[81]:
```
sns.pairplot(df1)
```

[81]: <seaborn.axisgrid.PairGrid at 0x7d0fd20625f0>



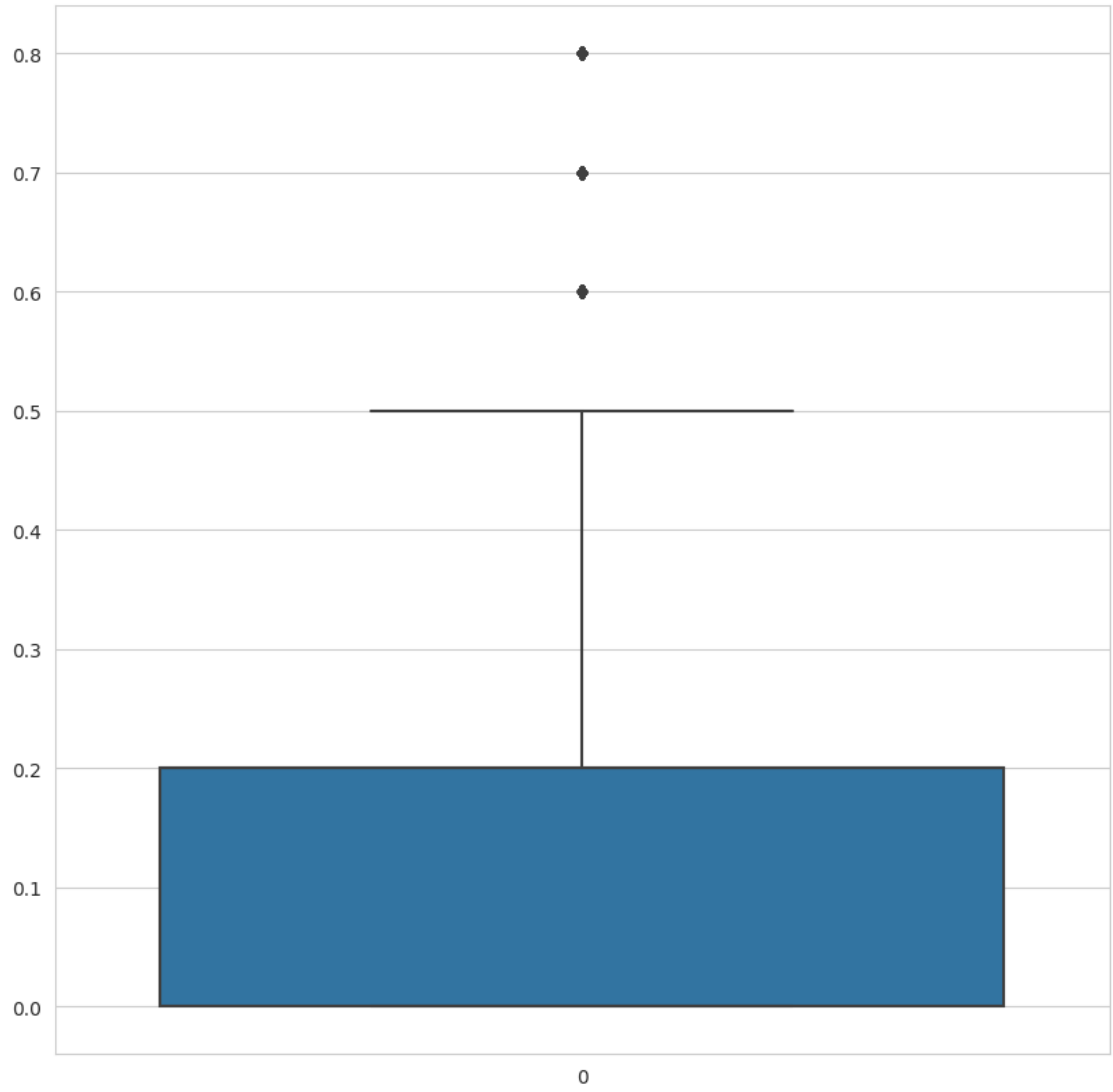
[82]:
```
fig, axes = plt.subplots(figsize = (10 , 10))

sns.boxplot(df['Sales'])
```

[82]: <Axes: >



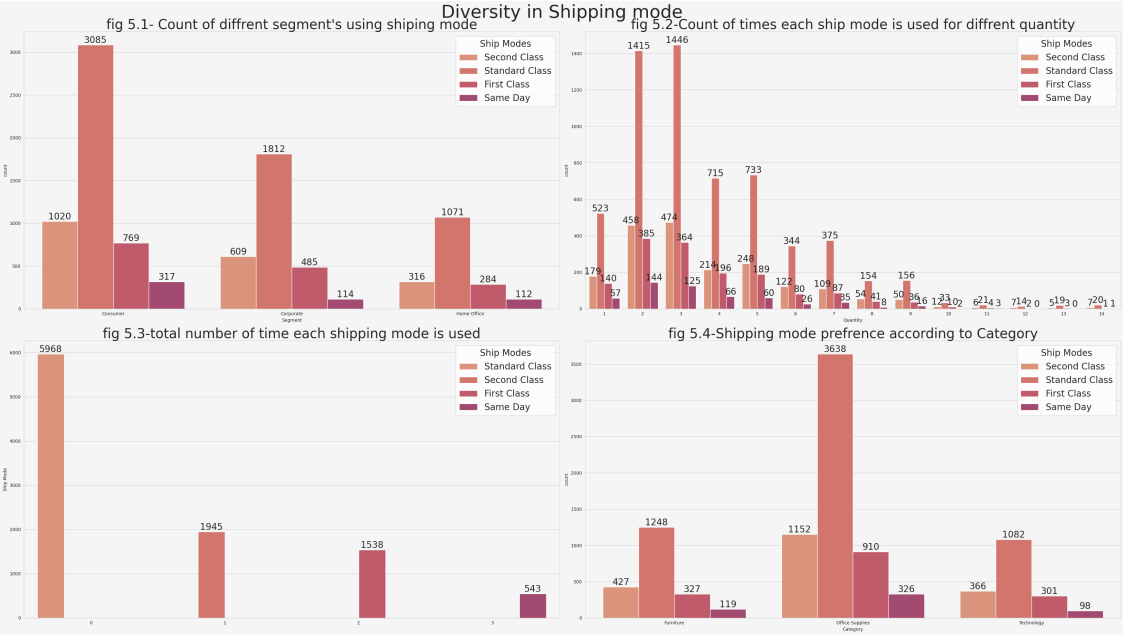
[83]:
```
fig, axes = plt.subplots(figsize = (10 , 10))

sns.boxplot(df['Discount'])
```

[83]: <Axes: >



[85]:
```
ship=ship=clean_data["Ship Mode"].value_counts().reset_index()
fig, axs = plt.subplots(2,2,figsize=(35,20))
fig.patch.set_facecolor('#f6f5f5')
sns.countplot(data=clean_data, x = "Segment",hue="Ship Mode",␣
 ↪palette=palette_color,ax=axs[0][0])
sns.countplot(data=clean_data, x = "Quantity",hue="Ship Mode",␣
 ↪palette=palette_color,ax=axs[0][1])
sns.barplot(data=ship, x = ship.index,y="Ship Mode",hue="index",␣
 ↪palette=palette_color,ax=axs[1][0])
sns.countplot(data=clean_data,x="Category",hue="Ship Mode",␣
 ↪palette=palette_color,ax=axs[1][1])
```

```
axs[0][0].set_title("fig 5.1- Count of diffrent segment's using shiping␣
 ↪mode",fontsize=30)
axs[0][0].set_facecolor('#f6f5f5')

axs[0][1].set_title("fig 5.2-Count of times each ship mode is used for diffrent␣
 ↪quantity",fontsize=30)
axs[0][1].set_facecolor('#f6f5f5')

axs[1][0].set_title("fig 5.3-total number of time each shipping mode is␣
 ↪used",fontsize=30)
axs[1][0].set_facecolor('#f6f5f5')

axs[1][1].set_title("fig 5.4-Shipping mode prefrence according to␣
 ↪Category",fontsize=30)
axs[1][1].set_facecolor('#f6f5f5')

for m in range(2):
    for n in range(2):
        axs[m][n].legend(fontsize = '11',
           title = 'Ship Modes', title_fontsize = '20',
           prop={'size': 20},
           loc="upper right")
        for i in axs[m][n].containers:
            axs[m][n].bar_label(i, fontsize=20)


plt.suptitle("Diversity in Shipping mode",fontsize=40)



plt.tight_layout()
```



Conclusion (Act Phase) Exploratory Data Analysis (EDA) on the dataset "SampleSuperstore" has been given to us, and as company managers, we are to identify the areas that need improvement in order to increase profit. identifying the weak points in the sales department to increase sales Based on the information discovered during the data analysis, the following techniques have been suggested.

[ ]: