

DVA Assignment 1

June 17, 2024

Q1. Write a program to create a DataFrame have E-commerce data and perform selection of row/column using loc() and iloc()

```
[1]: import pandas as pd
import numpy as np
data = {
    'Product_Id': np.arange(1, 11),
    'Price': np.random.randint(500, 2000, 10),
    'Quantity' : np.random.randint(1, 10, 10),
    'Customer_Id' : np.random.randint(1101, 1200, 10)
}
df = pd.DataFrame(data)
print(df)
data_iloc = df.iloc[1:4, 2:4]
print("\n [iloc] Selecting rows from 1 to 3 and columns from 2 to 3\n",
      data_iloc)
data_loc = df.loc[1:4]
print("\n [loc] Selecting rows from 1 to 4\n", data_loc)
```

	Product_Id	Price	Quantity	Customer_Id
0	1	1708	8	1156
1	2	1320	5	1115
2	3	1940	8	1171
3	4	1080	8	1186
4	5	1829	9	1159
5	6	1215	7	1156
6	7	901	6	1124
7	8	598	7	1101
8	9	707	8	1180
9	10	791	9	1171

```
[iloc] Selecting rows from 1 to 3 and columns from 2 to 3
  Quantity  Customer_Id
1         5         1115
2         8         1171
3         8         1186
```

```
[loc] Selecting rows from 1 to 4
  Product_Id  Price  Quantity  Customer_Id
1          2    1320         5         1115
2          3    1940         8         1171
3          4    1080         8         1186
```

1	2	1320	5	1115
2	3	1940	8	1171
3	4	1080	8	1186
4	5	1829	9	1159

Q2. Create a Series object S5 containing numbers. Write a program to store the square of the series values in object S6. Display S6's values which are >15.

```
[3]: import pandas as pd
s5= pd.Series(np.random.randint(1,10, 5))
print(s5)
s6 = s5**2
print(s6)
new_series = s6[s6 > 15]
print(new_series)
```

```
0    1
1    7
2    5
3    1
4    6
dtype: int32
0    1
1   49
2   25
3    1
4   36
dtype: int32
0    1
1   49
2   25
4   36
dtype: int32
```

Q3. Write a program to fill all missing values in a DataFrame with zero

```
[4]: data = {
    'A': [10, 24, np.nan, 41],
    'B': [65, np.nan, 27, 38],
    'C': [9, 10, 31, np.nan] }
df = pd.DataFrame(data)
print(df, "\n")
df_filled = df.fillna(0)
print(df_filled)
```

	A	B	C
0	10.0	65.0	9.0
1	24.0	NaN	10.0
2	NaN	27.0	31.0
3	41.0	38.0	NaN

	A	B	C
0	10.0	65.0	9.0
1	24.0	0.0	10.0
2	0.0	27.0	31.0
3	41.0	38.0	0.0

Q4. Program for combining DataFrames using concat(), join(),merge()

```
[6]: import pandas as pd
data1= pd.DataFrame({
    'id' : [10, 11, 12, 13],
    'name' : ['Sourav', 'Karthik', 'Suryadev', 'Shakti'],
})
data2= pd.DataFrame({
    'id': [12, 13, 14],
    'age' : [20, 24, 31]
})
concat_df = pd.concat([data1, data2])
print("CONCATENATED DATAFRAME (vertically) \n", concat_df)
joined_df = data1.set_index('id').join(data2.set_index('id'), how='inner')
print("\n JOIN DATAFRAME \n", joined_df)
merge_df = pd.merge(data1, data2, on='id')
print("\n MERGE DATAFRAME \n", merge_df)
```

CONCATENATED DATAFRAME (vertically)

	id	name	age
0	10	Sourav	NaN
1	11	Karthik	NaN
2	12	Suryadev	NaN
3	13	Shakti	NaN
0	12	NaN	20.0
1	13	NaN	24.0
2	14	NaN	31.0

JOIN DATAFRAME

	name	age
id		
12	Suryadev	20
13	Shakti	24

MERGE DATAFRAME

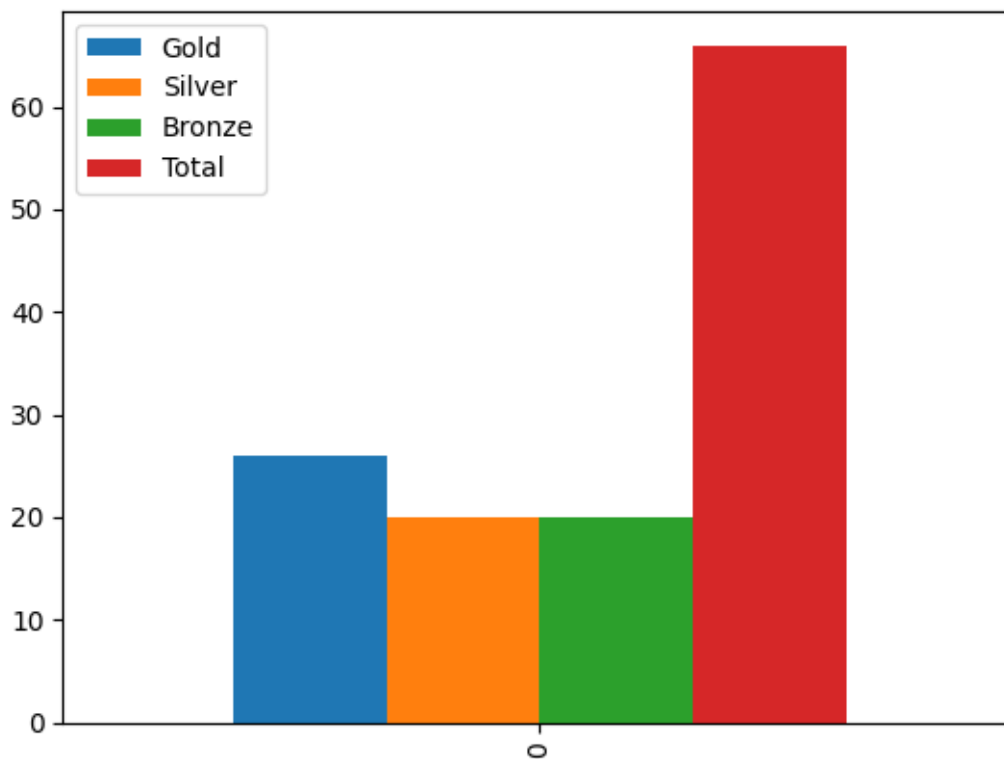
	id	name	age
0	12	Suryadev	20
1	13	Shakti	24

Q5. Write a program to draw bar graph for the following data for the Medal tally of Olympic games

Gold: 26 Silver: 20 Bronze: 20 Total: 66

```
[9]: import matplotlib.pyplot as plt
data = {
    'Gold': [26],
    'Silver': [20],
    'Bronze': [20],
    'Total': [66]
}
df = pd.DataFrame(data)
print(df)
df.plot(kind='bar', width = 0.9)
plt.show()
```

	Gold	Silver	Bronze	Total
0	26	20	20	66



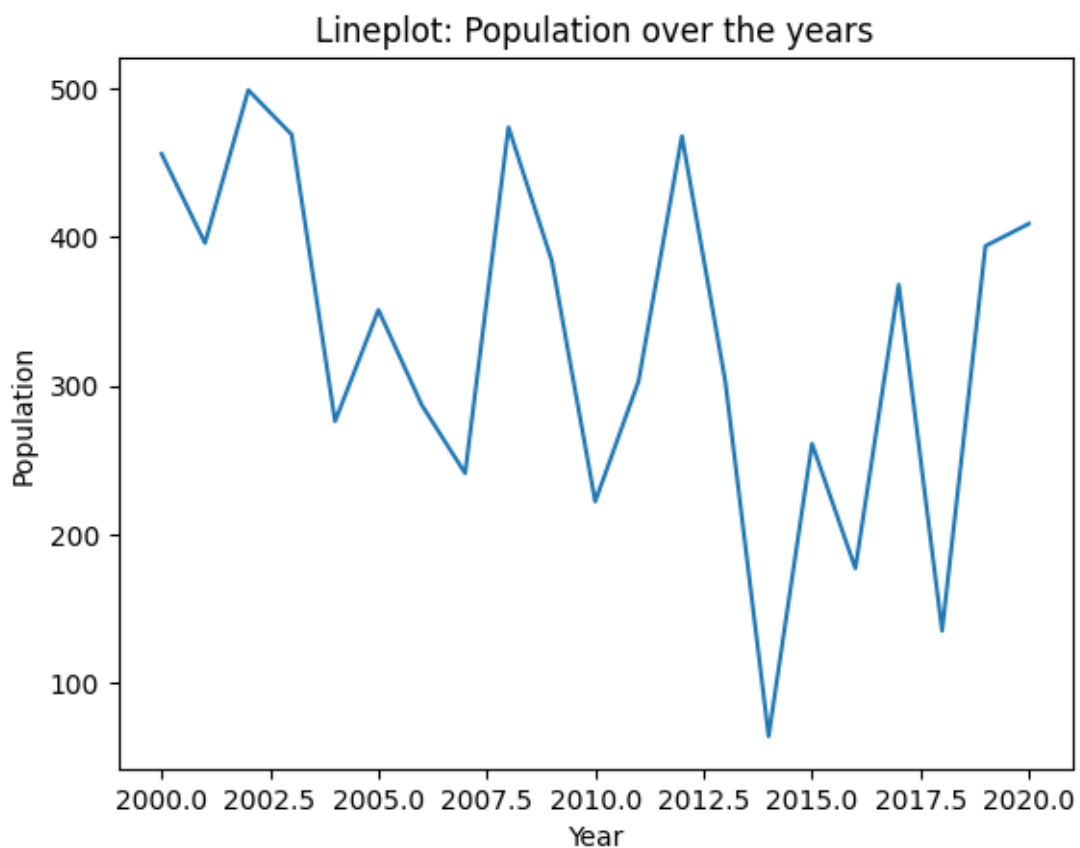
Q6. Implementing Line plot, Dist plot, Lmplot, Count plot using Seaborn library

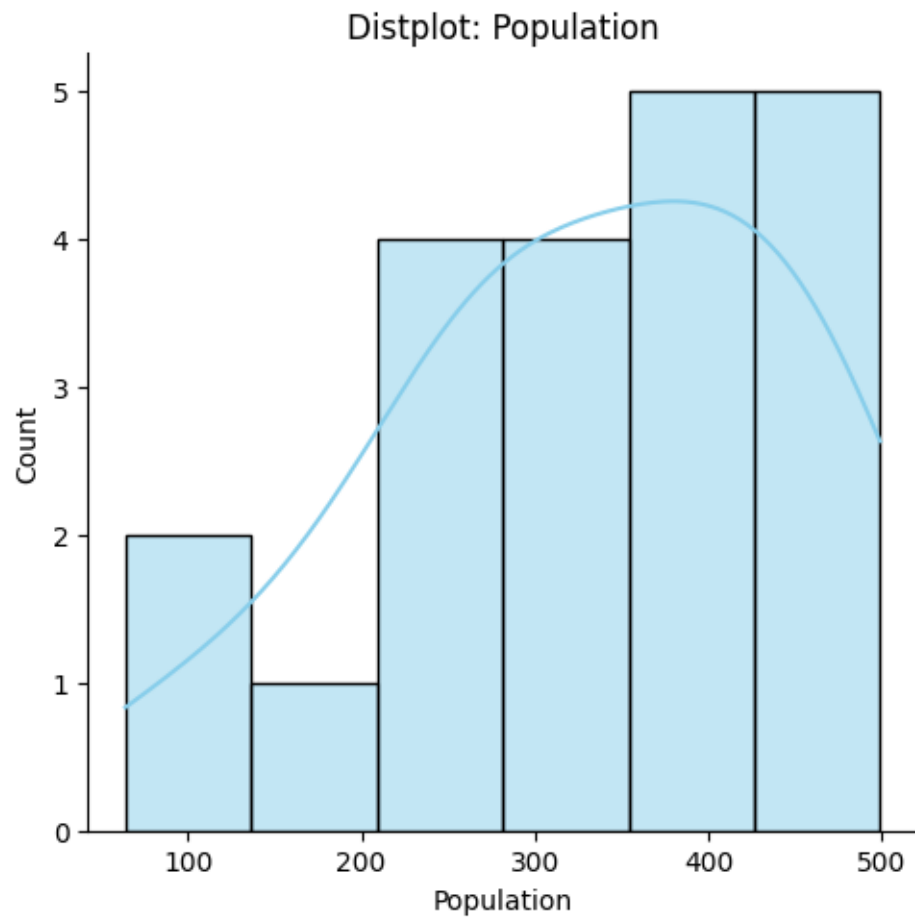
```
[13]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
years = np.arange(2000, 2021, 1, dtype = int)
population = np.random.randint(50, 500, 21)
```

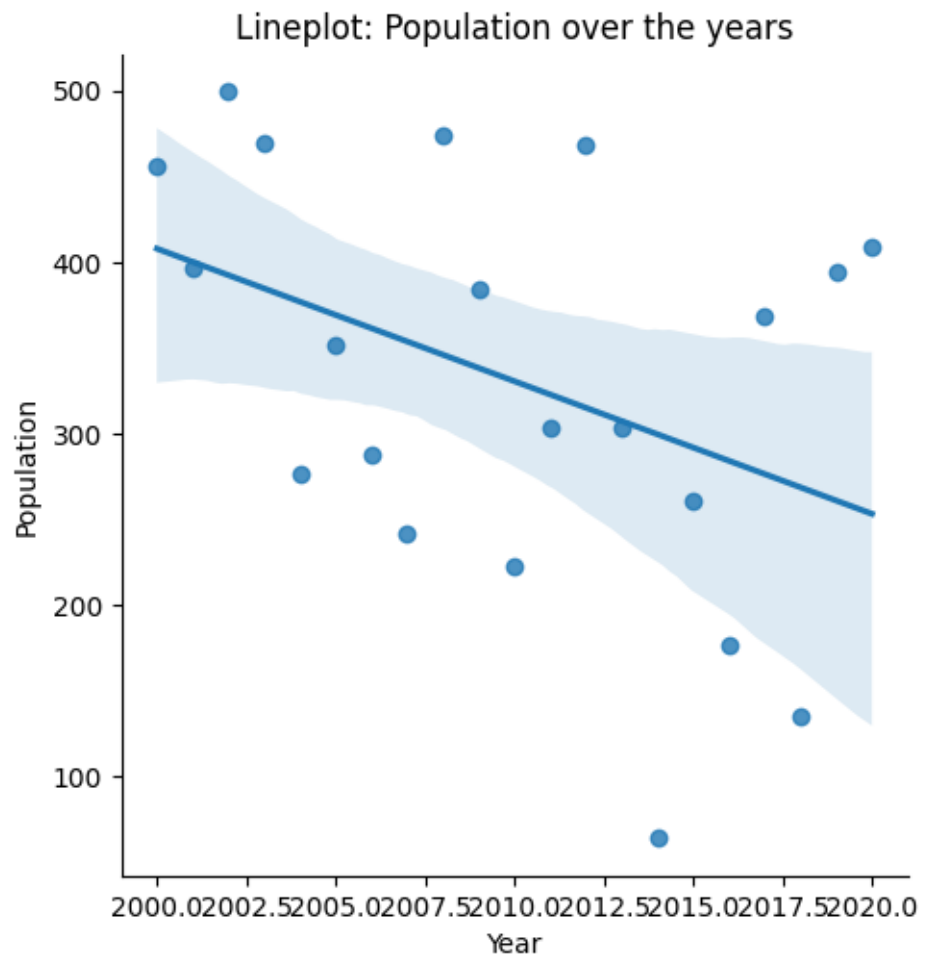
```

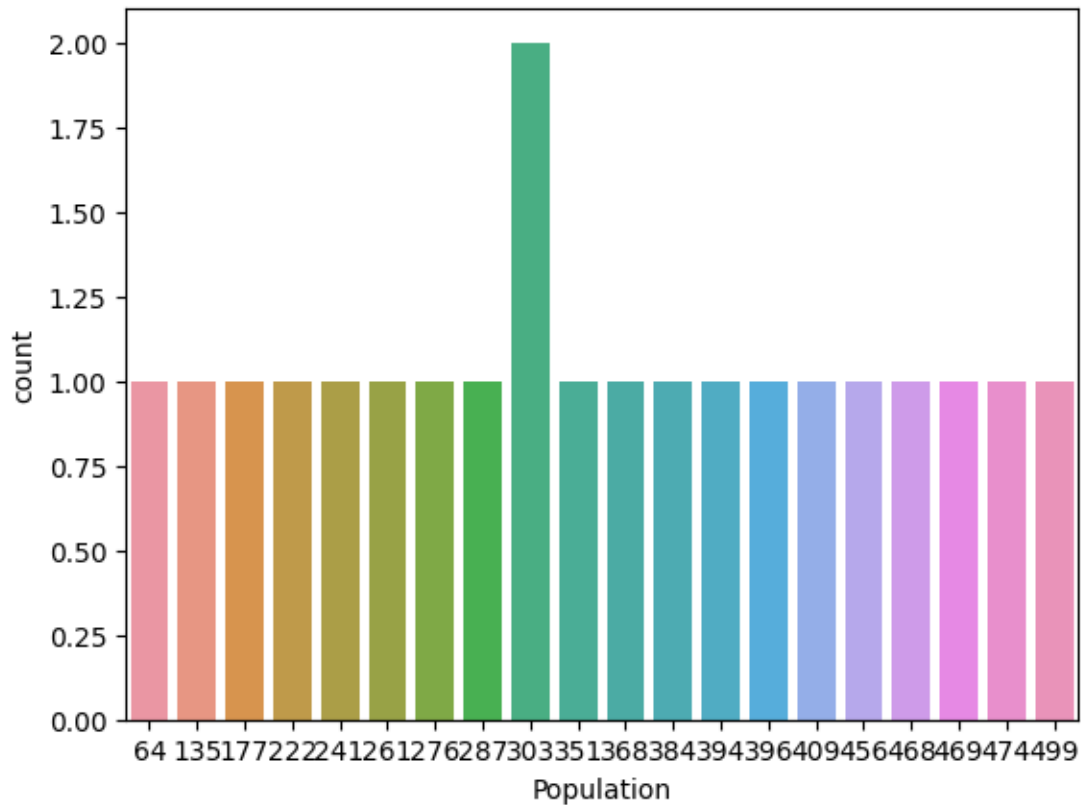
df = pd.DataFrame({'Year': years, 'Population': population })
sns.lineplot(x='Year', y='Population', data = df)
plt.title("Lineplot: Population over the years")
sns.displot(df['Population'], kde=True, color = 'skyblue')
plt.title("Distplot: Population")
plt.tight_layout()
plt.show()
sns.lmplot(x='Year', y='Population', data = df)
plt.title("Lineplot: Population over the years")
plt.show()
sns.countplot(x = 'Population', data = df)
plt.show()

```









Q7. Create a DataFrame namely aid that stores aid (Toys, books, uniform, shoes) by NGO's for different states. Write a program to display the aid for: - (a) Books and Uniforms only (b) Shoes only

```
[16]: import pandas as pd
import numpy as np
data = {

    'State': ['Delhi','Haryana', 'Uttar Pradesh', 'Rajasthan', 'Gujarat', 'Kerala', 'Maharashtra', ],
    'Toys' : np.random.randint(5000, 8000, 7, dtype= int),
    'Books' : np.random.randint(10000, 15000, 7, dtype = int),
    'uniform': np.random.randint(5000, 8000, 7, dtype= int),
    'Shoes' : np.random.randint(5000, 8000, 7, dtype= int)
}
aid = pd.DataFrame(data)
print(aid)
print("\n(a) Aid for Books and Uniforms only:")
print(aid[['State', 'Books', 'uniform']])
print("\n(b) Aid for Shoes only:")
print(aid[['State', 'Shoes']])
```


	State	Toys	Books	uniform	Shoes
0	Delhi	5772	10153	5403	6235
1	Haryana	5604	14083	7450	5120
2	Uttar Pradesh	7126	12675	6381	7104
3	Rajasthan	6634	13703	6416	7039
4	Gujarat	6544	11517	7735	6002
5	Kerala	6892	14313	6564	7866
6	Maharashtra	6896	14854	7841	6165

(a) Aid for Books and Uniforms only:

	State	Books	uniform
0	Delhi	10153	5403
1	Haryana	14083	7450
2	Uttar Pradesh	12675	6381
3	Rajasthan	13703	6416
4	Gujarat	11517	7735
5	Kerala	14313	6564
6	Maharashtra	14854	7841

(b) Aid for Shoes only:

	State	Shoes
0	Delhi	6235
1	Haryana	5120
2	Uttar Pradesh	7104
3	Rajasthan	7039
4	Gujarat	6002
5	Kerala	7866
6	Maharashtra	6165

Q8. Create a DataFrame df having Name, Gender, Position, City, Age, Projects. Write a program to summarize how many projects are being handled by each position for each city? Use pivot()

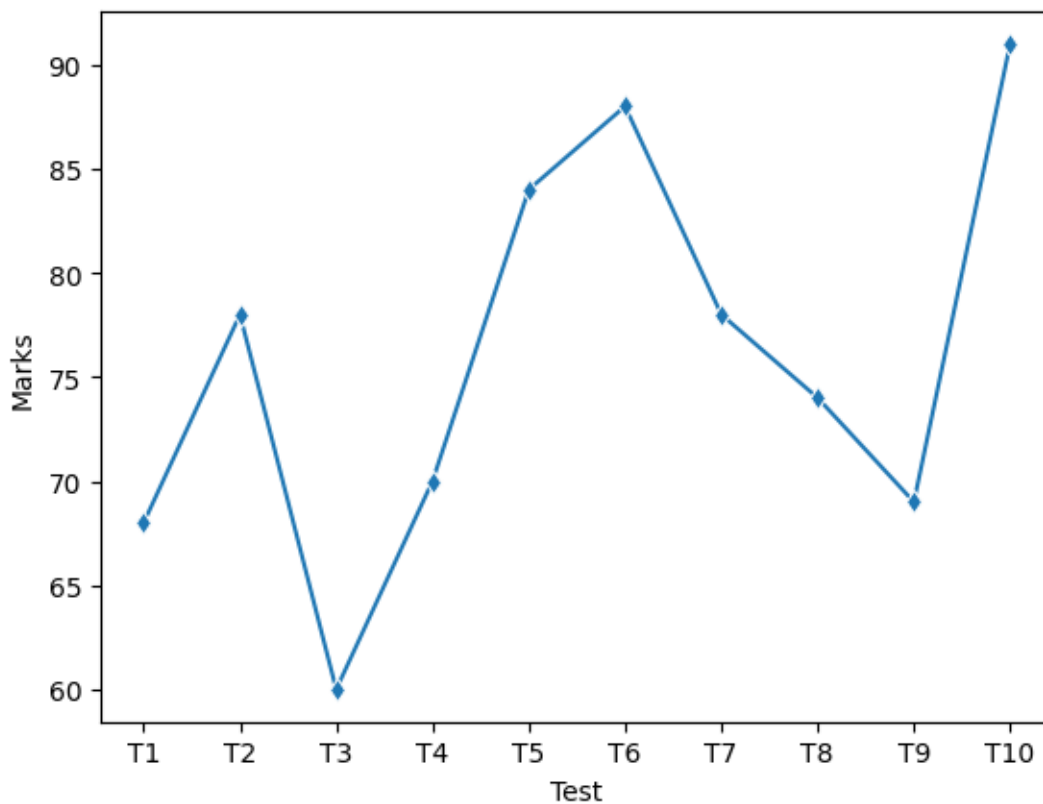
```
[19]: import pandas as pd
data = {
    'Name': ['Sourav', 'Aarti', 'Suryadev', 'Viswam', 'Wusat'],
    'Gender': ['Male', 'Female', 'Male', 'Male', 'Female'],
    'Position': ['Analyst', 'Developer', 'Analyst', 'Developer', 'Manager'],
    'City': ['Boston', 'Los Angeles', 'Boston', 'Los Angeles', 'Boston'],
    'Age': [20, 25, 35, 27, 32],
    'Projects': [5, 3, 6, 4, 7]
}
df = pd.DataFrame(data)
grouped_df = df.groupby(['Position', 'City'], as_index=False)['Projects'].sum()
pt = grouped_df.pivot(index='Position', columns='City', values = 'Projects').
    ↪ fillna(0)
print(pt)
```

City	Boston	Los Angeles
Position		
Analyst	11.0	0.0
Developer	0.0	7.0
Manager	7.0	0.0

Q9. Marks is a list that stores marks of a student in 10-unit test. Write a program to plot Line chart for the student's performance in these 10 tests

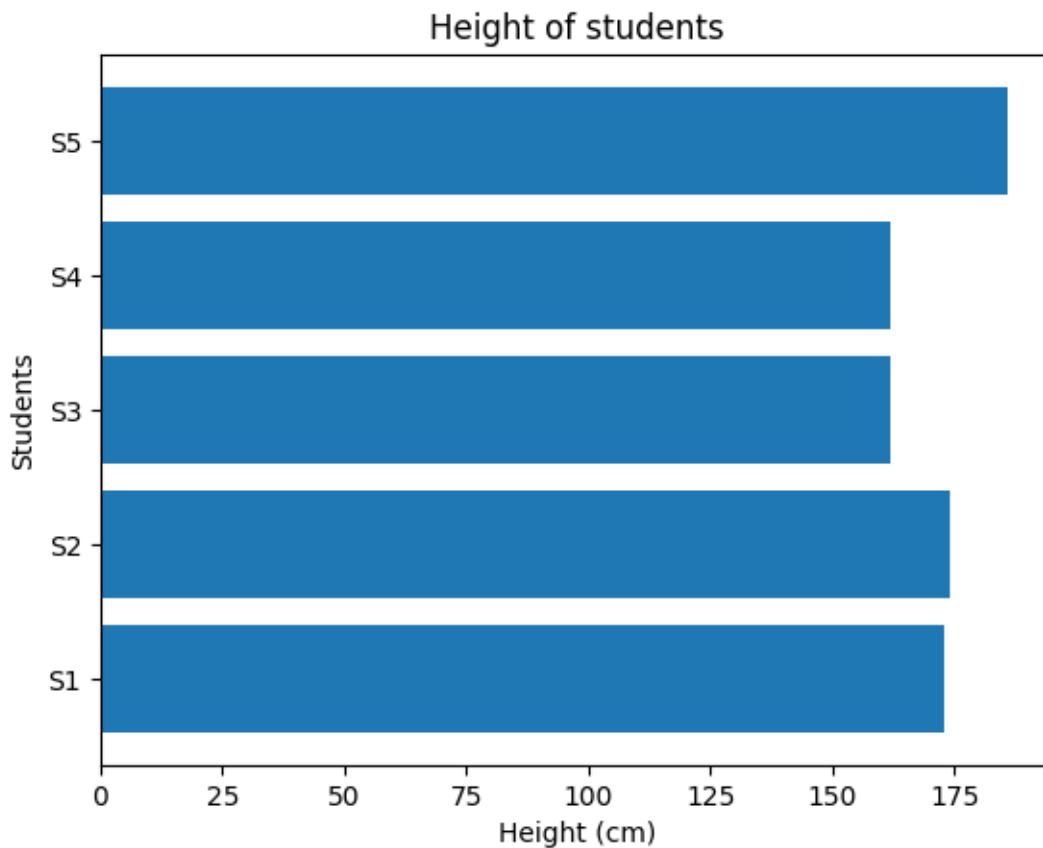
```
[21]: import seaborn as sns
import numpy as np
import matplotlib.pyplot as plt
Test = ['T1', 'T2', 'T3', 'T4', 'T5', 'T6', 'T7', 'T8', 'T9', 'T10']
marks = np.random.randint(30, 100, 10)
df = pd.DataFrame({'Test': Test, 'Marks': marks})
sns.lineplot(data=df, x='Test', y='Marks', marker = 'd')
```

```
[21]: <Axes: xlabel='Test', ylabel='Marks'>
```



Q10. Write a program to plot a horizontal bar chart from the height of some students

```
[22]: import matplotlib.pyplot as plt
import numpy as np
students = ['S1', 'S2', 'S3', 'S4', 'S5']
heights = np.random.randint(150, 190, len(students))
# Create a horizontal bar chart
plt.barh(students, heights)
plt.xlabel('Height (cm)')
plt.ylabel('Students')
plt.title('Height of students')
plt.show()
```



Q11. Write a program to implement Covariance.

```
[24]: import numpy as np
d1 = np.array([11,12,13,14,15])
d2 = np.array([30,40,50,10,20])
def covariance(X, Y):
    n = len(X)
    mean_X = np.mean(X)
    mean_Y = np.mean(Y)
```

```

cov = np.sum((X - mean_X) * (Y - mean_Y)) / (n - 1)
return cov
covariance_value = covariance(d1, d2)
print(f"Covariance between X and Y: {covariance_value}")
print("\n Covariance through np.cov()\n", np.cov(d1, d2))

```

Covariance between X and Y: -12.5

```

Covariance through np.cov()
[[ 2.5 -12.5]
 [-12.5 250. ]]

```

Q12. Segmentation: Clustering (K-Means) Customer Personality Analysis is a detailed analysis of a company's ideal customers. It helps a business to better understand its customers and makes it easier for them to modify products according to the specific needs, behaviors and concerns of different types of customers. Customer personality analysis helps a business to modify its product based on its target customers from different types of customer segments. For example, instead of spending money to market a new product to every customer in the company's database, a company can analyze which customer segment is most likely to buy the product and then market the product only on that particular segment. link to dataset: <https://www.kaggle.com/datasets/imakash3011/customerpersonality-analysis> Refer <https://github.com/ibrahim-ogunbiyi/CustomerSegmentation/blob/main/Customer%20Segmentation>

```

[1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt

df = pd.read_csv("C:\SOURAV\BCA\Semester-6\Data Visualisation and_
↳Analytics\Assignments\Supermarket store Dataset.csv")
df.head()

```

```

[1]:
Store ID  Store_Area  Items_Available  Daily_Customer_Count  Store_Sales
0         1         1659             1961                   530         66490
1         2         1461             1752                   210         39820
2         3         1340             1609                   720         54010
3         4         1451             1748                   620         53730
4         5         1770             2111                   450         46620

```

```

[41]: df.describe()

```

```

[41]:
Store ID  Store_Area  Items_Available  Daily_Customer_Count
count  896.000000    896.000000         896.000000         896.000000 \
mean    448.500000   1485.409598        1782.035714        786.350446
std     258.797218    250.237011         299.872053        265.389281
min       1.000000    775.000000         932.000000         10.000000
25%     224.750000   1316.750000        1575.500000        600.000000

```

50%	448.500000	1477.000000	1773.500000	780.000000
75%	672.250000	1653.500000	1982.750000	970.000000
max	896.000000	2229.000000	2667.000000	1560.000000

	Store_Sales
count	896.000000
mean	59351.305804
std	17190.741895
min	14920.000000
25%	46530.000000
50%	58605.000000
75%	71872.500000
max	116320.000000

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

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 896 entries, 0 to 895
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Store ID               896 non-null   int64
1   Store_Area             896 non-null   int64
2   Items_Available        896 non-null   int64
3   Daily_Customer_Count   896 non-null   int64
4   Store_Sales            896 non-null   int64
dtypes: int64(5)
memory usage: 35.1 KB
```

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

```
[43]: Store ID                0
Store_Area                 0
Items_Available            0
Daily_Customer_Count       0
Store_Sales                0
dtype: int64
```

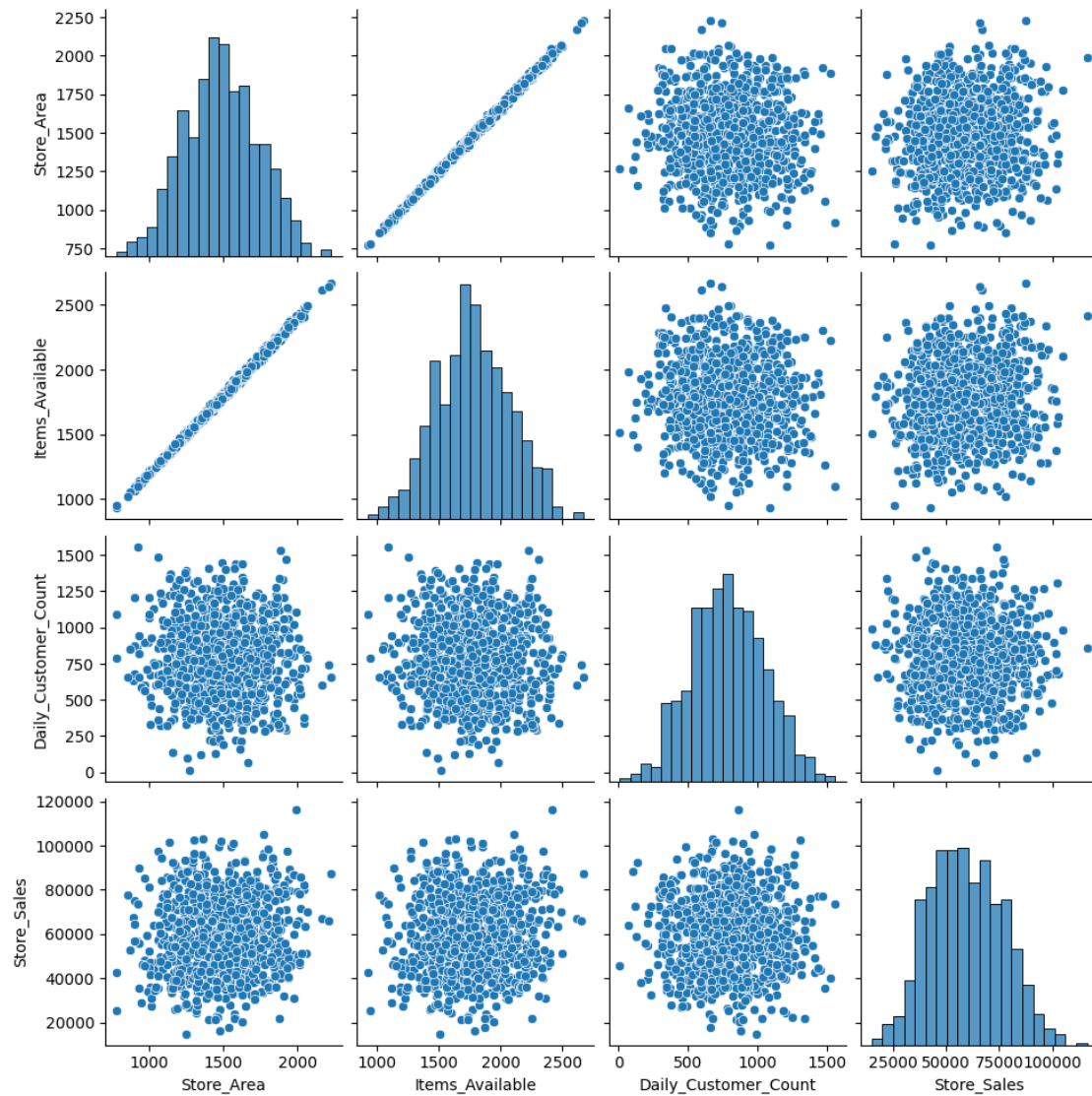
```
[47]: df.duplicated().sum()
```

```
[47]: 0
```

```
[ ]: df.drop(['Store ID'], axis=1)
```

```
[51]: plt.figure(figsize=(3,3))
sns.pairplot(data = df)
plt.show()
```

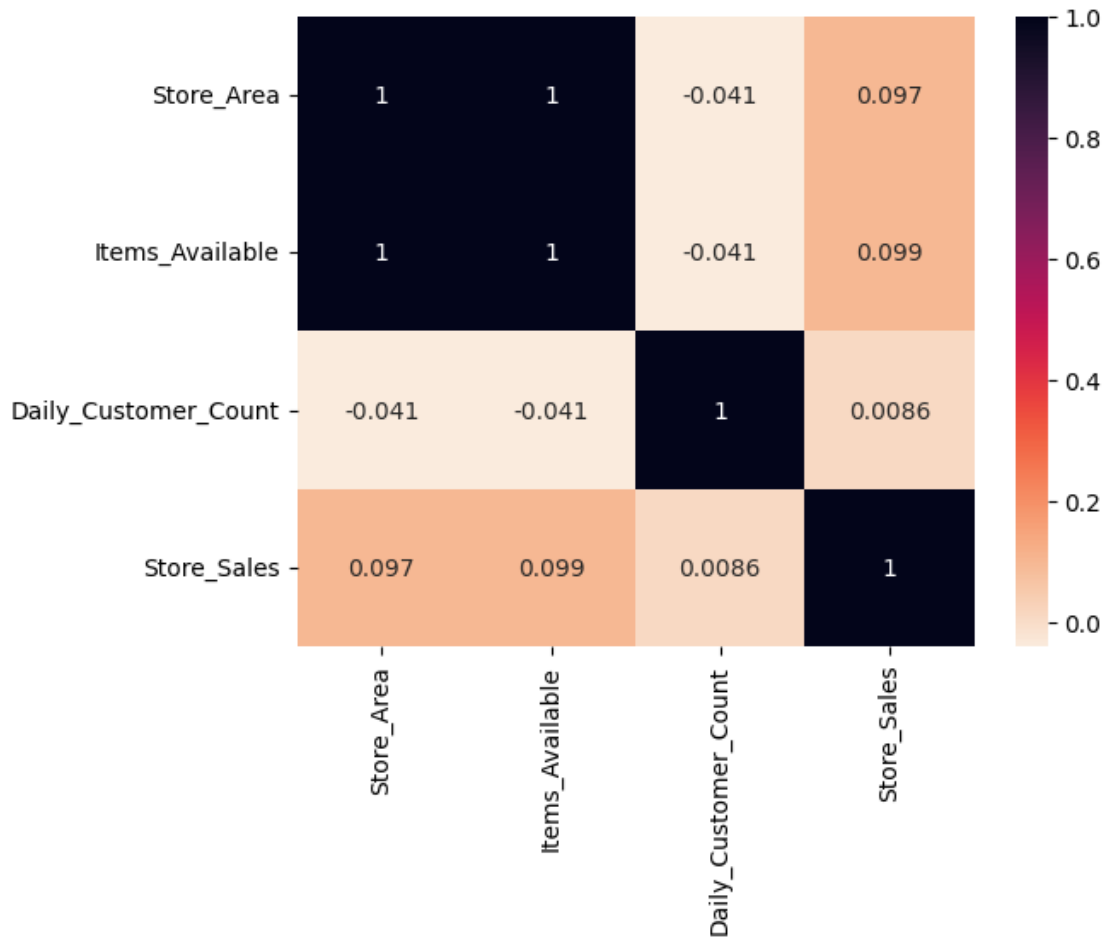
```
<Figure size 300x300 with 0 Axes>
```



The distribution of Store_Area, Items_Available, Daily_Customer_Count, Store_Sales is normal. Item_Available and Store_Area has high positive correlation

```
[52]: sns.heatmap(df.corr(),annot=True, cmap= 'rocket_r')
```

```
[52]: <Axes: >
```



From above heatmap, we can notice that Store_Area and Items_Available has a linear relationship. Therefore, it will be a good idea to drop one of them. And has more impact on Store_Sales.

```
[53]: df = df.drop(['Items_Available'], axis=1)
```

We'll only use 'Store_Area' and 'Store_Sales' columns to get insights into store characteristics and sales performance.

```
[56]: df = df[['Store_Area', 'Store_Sales']]
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
scaled_features = scaler.fit_transform(df)
```

```
[60]: from sklearn.cluster import KMeans
```

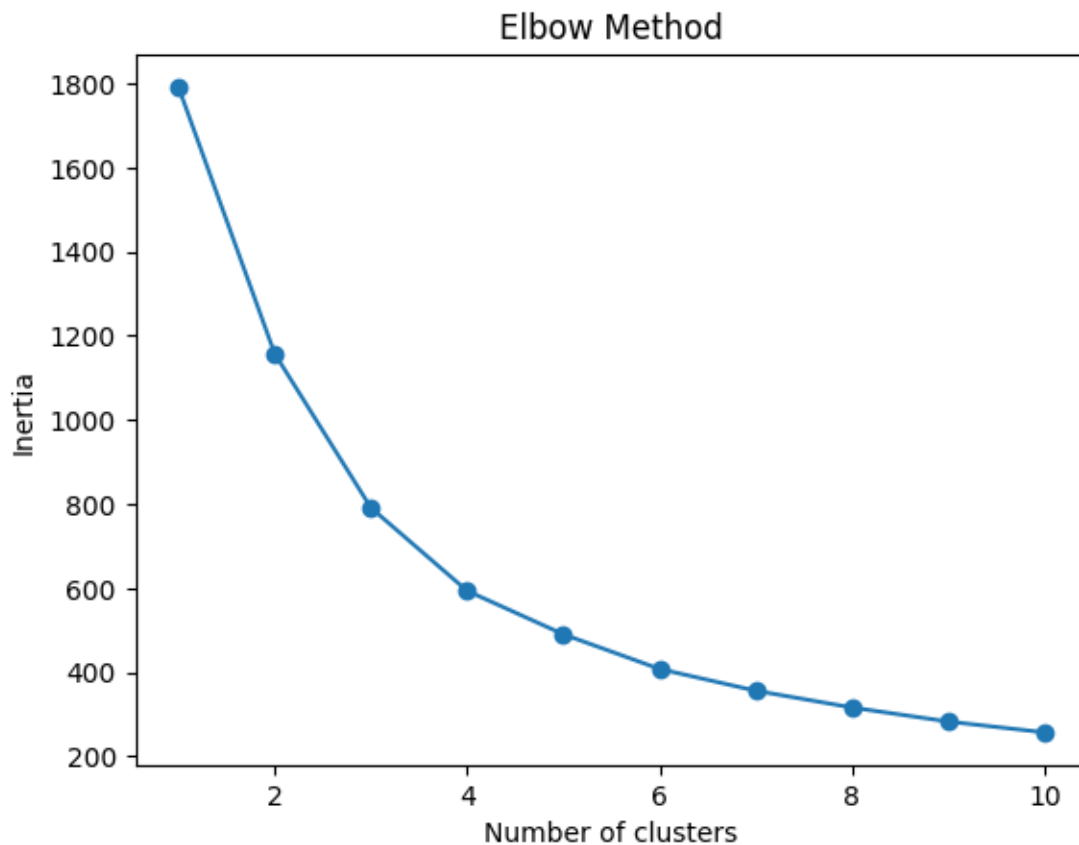
```
inertia = []
for i in range(1, 11):
```

```

kmeans = KMeans (n_clusters=i, n_init=10, random_state=42)
kmeans.fit(scaled_features)
inertia.append(kmeans.inertia_)

plt.plot(range(1, 11), inertia, marker='o')
plt.xlabel('Number of clusters')
plt.ylabel('Inertia')
plt.title('Elbow Method')
plt.show()

```



```

[59]: num_clusters = 3
      #Perform K-means clustering
      kmeans = KMeans(n_clusters=num_clusters, n_init=10, random_state=42)
      kmeans.fit(scaled_features)

      # Add cluster Labels to the DataFrame
      df['Cluster'] = kmeans.labels_

      # Visualize the clusters

```



```
sns.scatterplot(x= 'Store_Area', y='Store_Sales', hue='Cluster', data=df)
plt.xlabel('Store Area')
plt.ylabel('Store Sales')
plt.title('Clusters of Store Area vs. Store Sales')
plt.show()
```



[]: