# **Problem statement**

# Data collection ¶

## In [1]:

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

## In [73]:

```
df=pd.read_csv(r"E:\Dataset\6_Salesworkload1.csv")[0:500]
df
```

## Out[73]:

|       | MonthYear             | Time<br>index | Country           | StoreID | City          | Dept_ID | Dept.<br>Name          | HoursOwn  | HoursLeas |
|-------|-----------------------|---------------|-------------------|---------|---------------|---------|------------------------|-----------|-----------|
| 0     | 10.2016               | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 1.0     | Dry                    | 3184.764  | 0         |
| 1     | 10.2016               | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 2.0     | Frozen                 | 1582.941  | 0         |
| 2     | 10.2016               | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 3.0     | other                  | 47.205    | 0         |
| 3     | 10.2016               | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 4.0     | Fish                   | 1623.852  | 0         |
| 4     | 10.2016               | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 5.0     | Fruits &<br>Vegetables | 1759.173  | 0         |
|       |                       |               |                   |         |               |         |                        |           |           |
| 495   | 10.2016               | 1.0           | Italy             | 64983.0 | Milano        | 3.0     | other                  | 47.205    | 0         |
| 496   | 10.2016               | 1.0           | Italy             | 64983.0 | Milano        | 4.0     | Fish                   | 2451.513  | 0         |
| 497   | 10.2016               | 1.0           | Italy             | 64983.0 | Milano        | 5.0     | Fruits &<br>Vegetables | 1944.846  | 0         |
| 498   | 10.2016               | 1.0           | Italy             | 64983.0 | Milano        | 6.0     | Meat                   | 11980.629 | 122       |
| 499   | 10.2016               | 1.0           | Italy             | 64983.0 | Milano        | 13.0    | Food                   | 23665.44  | 122       |
| 500 ı | 500 rows × 14 columns |               |                   |         |               |         |                        |           |           |

## In [74]:

df.head()

## Out[74]:

|   | MonthYear | Time<br>index | Country           | StoreID | City          | Dept_ID | Dept.<br>Name          | HoursOwn | HoursLease |
|---|-----------|---------------|-------------------|---------|---------------|---------|------------------------|----------|------------|
| 0 | 10.2016   | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 1.0     | Dry                    | 3184.764 | 0.0        |
| 1 | 10.2016   | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 2.0     | Frozen                 | 1582.941 | 0.0        |
| 2 | 10.2016   | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 3.0     | other                  | 47.205   | 0.0        |
| 3 | 10.2016   | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 4.0     | Fish                   | 1623.852 | 0.0        |
| 4 | 10.2016   | 1.0           | United<br>Kingdom | 88253.0 | London<br>(I) | 5.0     | Fruits &<br>Vegetables | 1759.173 | 0.0        |
| 4 |           |               |                   |         |               |         |                        |          | •          |

## In [75]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 14 columns):

| Column        | Non-Null Count   | Dtype   |
|---------------|--|---|
|               |  |   |
| MonthYear     | 500 non-null   | object  |
| Time index    | 500 non-null   | float64   |
| Country       | 500 non-null   | object  |
| StoreID       | 500 non-null   | float64   |
| City          | 500 non-null   | object  |
| Dept_ID       | 500 non-null   | float64   |
| Dept. Name    | 500 non-null   | object  |
| HoursOwn      | 500 non-null   | object  |
| HoursLease    | 500 non-null   | float64   |
| Sales units   | 500 non-null   | float64   |
| Turnover      | 500 non-null   | float64   |
| Customer      | 0 non-null   | float64   |
| Area (m2)     | 500 non-null   | object  |
| Opening hours | 500 non-null   | object  |
|               | MonthYear Time index Country StoreID City Dept_ID Dept. Name HoursOwn HoursLease Sales units Turnover Customer Area (m2) | MonthYear 500 non-null Time index 500 non-null Country 500 non-null StoreID 500 non-null City 500 non-null Dept_ID 500 non-null Dept. Name 500 non-null HoursOwn 500 non-null HoursLease 500 non-null Sales units 500 non-null Turnover 500 non-null Customer 0 non-null Area (m2) 500 non-null |

dtypes: float64(7), object(7)

memory usage: 54.8+ KB

## In [76]:

```
#to display summary of statistics
df.describe()
```

### Out[76]:

|       | Time index | StoreID      | Dept_ID    | HoursLease  | Sales units  | Turnover     | Customer |
|-------|------------|--------------|------------|-------------|--------------|--------------|----------|
| count | 500.0      | 500.000000   | 500.000000 | 500.000000  | 5.000000e+02 | 5.000000e+02 | 0.0      |
| mean  | 1.0        | 57412.764000 | 9.406000   | 31.520000   | 9.397837e+05 | 3.153113e+06 | NaN      |
| std   | 0.0        | 32104.273482 | 5.350366   | 142.134408  | 1.486945e+06 | 5.165524e+06 | NaN      |
| min   | 1.0        | 15552.000000 | 1.000000   | 0.000000    | 0.000000e+00 | 0.000000e+00 | NaN      |
| 25%   | 1.0        | 20891.000000 | 5.000000   | 0.000000    | 5.200250e+04 | 2.345122e+05 | NaN      |
| 50%   | 1.0        | 71991.000000 | 9.000000   | 0.000000    | 2.555375e+05 | 7.053345e+05 | NaN      |
| 75%   | 1.0        | 88253.000000 | 14.000000  | 0.000000    | 8.903900e+05 | 2.542147e+06 | NaN      |
| max   | 1.0        | 96857.000000 | 18.000000  | 1896.000000 | 7.476680e+06 | 2.571973e+07 | NaN      |

## In [77]:

```
#to display cloumn heading
df.columns
```

### Out[77]:

# **EDA and VISUALIZATION**

## In [78]:

df1=df[['MonthYear', 'Time index', 'Country', 'StoreID', 'City', 'Dept\_ID', 'Dept. Name']]
df1

## Out[78]:

|     | MonthYear | Time index | Country        | StoreID | City       | Dept_ID | Dept. Name          |
|-----|-----------|------------|----------------|---------|------------|---------|---------------------|
| 0   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 1.0     | Dry                 |
| 1   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 2.0     | Frozen              |
| 2   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 3.0     | other               |
| 3   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 4.0     | Fish                |
| 4   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 5.0     | Fruits & Vegetables |
|     |           |            |                |         |            |         | ***                 |
| 495 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 3.0     | other               |
| 496 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 4.0     | Fish                |
| 497 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 5.0     | Fruits & Vegetables |
| 498 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 6.0     | Meat                |
| 499 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 13.0    | Food                |

500 rows × 7 columns

## In [79]:

df1.fillna(1)

## Out[79]:

|     | MonthYear | Time index | Country        | StoreID | City       | Dept_ID | Dept. Name          |
|-----|-----------|------------|----------------|---------|------------|---------|---------------------|
| 0   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 1.0     | Dry                 |
| 1   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 2.0     | Frozen              |
| 2   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 3.0     | other               |
| 3   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 4.0     | Fish                |
| 4   | 10.2016   | 1.0        | United Kingdom | 88253.0 | London (I) | 5.0     | Fruits & Vegetables |
|     |           |            |                |         |            |         |                     |
| 495 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 3.0     | other               |
| 496 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 4.0     | Fish                |
| 497 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 5.0     | Fruits & Vegetables |
| 498 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 6.0     | Meat                |
| 499 | 10.2016   | 1.0        | Italy          | 64983.0 | Milano     | 13.0    | Food                |

500 rows × 7 columns

## In [80]:

## df1.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 500 entries, 0 to 499
Data columns (total 7 columns):

| # | Column     | Non-Null Count | Dtype   |
|---|------------|----------------|---------|
|   |            |                |         |
| 0 | MonthYear  | 500 non-null   | object  |
| 1 | Time index | 500 non-null   | float64 |
| 2 | Country    | 500 non-null   | object  |
| 3 | StoreID    | 500 non-null   | float64 |
| 4 | City       | 500 non-null   | object  |
| 5 | Dept_ID    | 500 non-null   | float64 |
| 6 | Dept. Name | 500 non-null   | object  |
|   |            |                |         |

dtypes: float64(3), object(4)

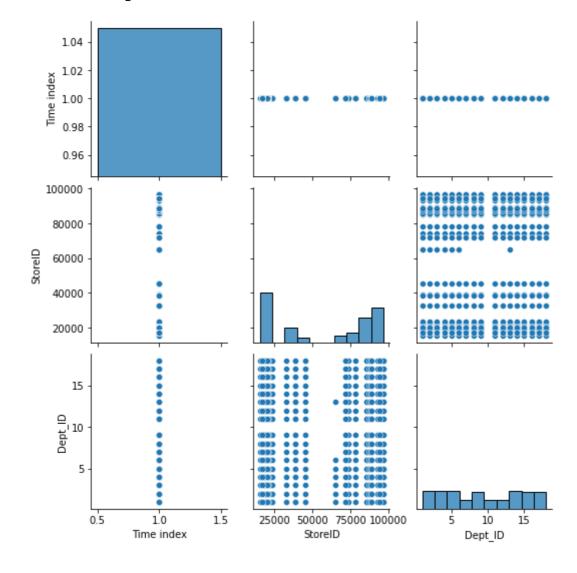
memory usage: 27.5+ KB

## In [81]:

sns.pairplot(df1)

### Out[81]:

<seaborn.axisgrid.PairGrid at 0x165be45dac0>



#### In [83]:

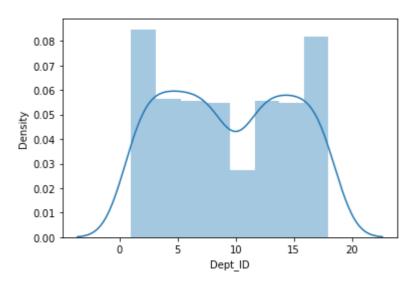
```
sns.distplot(df['Dept_ID'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure -level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

#### Out[83]:

<AxesSubplot:xlabel='Dept\_ID', ylabel='Density'>

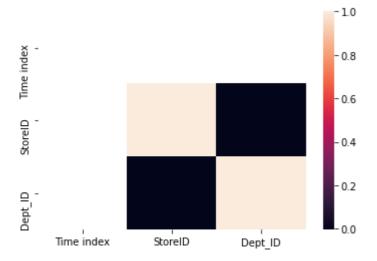


#### In [84]:

```
data=df1[[ 'Time index', 'StoreID', 'Dept_ID']]
sns.heatmap(data.corr())
```

#### Out[84]:

#### <AxesSubplot:>



# to Train the model-Model buliding

we are going to split our data into two variable where x is a independent and y is dependent on x

```
In [85]:
```

```
x=data[['Time index', 'StoreID', 'Dept_ID']]
y=data['Dept_ID']
```

## In [86]:

```
# to split my dataset into test and train data
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3)
```

## In [87]:

```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

#### Out[87]:

LinearRegression()

#### In [88]:

```
print(lr.intercept_)
```

#### 1.0658141036401503e-14

#### In [89]:

```
coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-effecient'])
coeff
```

# Out[89]:

#### Co-effecient

Time index 0.000000e+00

StoreID -1.247199e-19

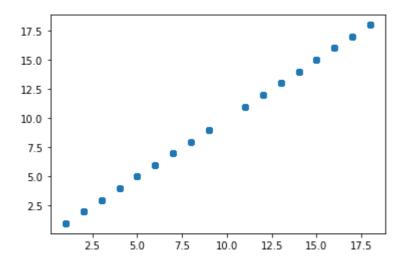
**Dept\_ID** 1.000000e+00

## In [90]:

prediction=lr.predict(x\_test)
plt.scatter(y\_test,prediction)

## Out[90]:

<matplotlib.collections.PathCollection at 0x165bec49730>



## In [91]:

print(lr.score(x\_test,y\_test))

### 1.0

## In [ ]: