Problem statement

Data collection

Importing libraries

In [1]:

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

Importing dataset

In [2]:

data=pd.read_csv(r"C:\Users\user\Downloads\6_Salesworkload1 - 6_Salesworkload1.csv")
data

Out[2]:

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	
0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.0	
1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.0	
2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.0	
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.0	
•••	•••								•••	
7653	6.2017	9.0	Sweden	29650.0	Gothenburg	12.0	Checkout	6322.323	0.0	3
7654	6.2017	9.0	Sweden	29650.0	Gothenburg	16.0	Customer Services	4270.479	0.0	
7655	6.2017	9.0	Sweden	29650.0	Gothenburg	11.0	Delivery	0	0.0	
7656	6.2017	9.0	Sweden	29650.0	Gothenburg	17.0	others	2224.929	0.0	
7657	6.2017	9.0	Sweden	29650.0	Gothenburg	18.0	all	39652.2	0.0	3

7658 rows × 14 columns

head

In [3]:

to display first 8 dataset values
da=data.head(8)

Out[3]:

0		MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	Sale unit
	0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.0	398560.
	1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.0	82725.
	2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.0	438400.
	3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	309425.
	4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.0	165515.
	5	10.2016	1.0	United Kingdom	88253.0	London (I)	6.0	Meat	8270.316	0.0	1713310.
	6	10.2016	1.0	United Kingdom	88253.0	London (I)	13.0	Food	16468.251	0.0	3107935.
	7	10.2016	1.0	United Kingdom	88253.0	London (I)	7.0	Clothing	4698.471	0.0	213680.
	4 ∥										>

info

In [4]:

to identify missing values
data.info()

Column	Non-Null Count	Dtype
MonthYear	7658 non-null	object
Time index	7650 non-null	float64
Country	7650 non-null	object
StoreID	7650 non-null	float64
City	7650 non-null	object
Dept_ID	7650 non-null	float64
Dept. Name	7650 non-null	object
HoursOwn	7650 non-null	object
HoursLease	7650 non-null	float64
Sales units	7650 non-null	float64
Turnover	7650 non-null	float64
Customer	0 non-null	float64
Area (m2)	7650 non-null	object
	MonthYear Time index Country StoreID City Dept_ID Dept. Name HoursOwn HoursLease Sales units Turnover Customer	MonthYear 7658 non-null Time index 7650 non-null Country 7650 non-null StoreID 7650 non-null City 7650 non-null Dept_ID 7650 non-null Dept. Name 7650 non-null HoursOwn 7650 non-null HoursLease 7650 non-null Sales units 7650 non-null Turnover 7650 non-null Customer 0 non-null

13 Opening hours 7650 non-null object dtypes: float64(7), object(7) memory usage: 837.7+ KB

describe

```
In [5]: # to display summary of the dataset
    data.describe()
```

Out[5]:		Time index	StoreID	Dept_ID	HoursLease	Sales units	Turnover	Customer
	count	7650.000000	7650.000000	7650.000000	7650.000000	7.650000e+03	7.650000e+03	0.0
	mean	5.000000	61995.220000	9.470588	22.036078	1.076471e+06	3.721393e+06	NaN
	std	2.582158	29924.581631	5.337429	133.299513	1.728113e+06	6.003380e+06	NaN
	min	1.000000	12227.000000	1.000000	0.000000	0.000000e+00	0.000000e+00	NaN
	25%	3.000000	29650.000000	5.000000	0.000000	5.457125e+04	2.726798e+05	NaN
	50%	5.000000	75400.500000	9.000000	0.000000	2.932300e+05	9.319575e+05	NaN
	75%	7.000000	87703.000000	14.000000	0.000000	9.175075e+05	3.264432e+06	NaN
	max	9.000000	98422.000000	18.000000	3984.000000	1.124296e+07	4.271739e+07	NaN

columns

Out[7]:		MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	Sale unit
	0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.0	398560.
	1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.0	82725.1
	2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.0	438400.
	3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	309425.
	4	10.2016	1.0	United	88253.0	London	5.0	Fruits &	1759.173	0.0	165515.

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	Sale unit
			Kingdom		(I)		Vegetables			
5	10.2016	1.0	United Kingdom	88253.0	London (I)	6.0	Meat	8270.316	0.0	1713310.
6	10.2016	1.0	United Kingdom	88253.0	London (I)	13.0	Food	16468.251	0.0	3107935.
7	10.2016	1.0	United Kingdom	88253.0	London (I)	7.0	Clothing	4698.471	0.0	213680.

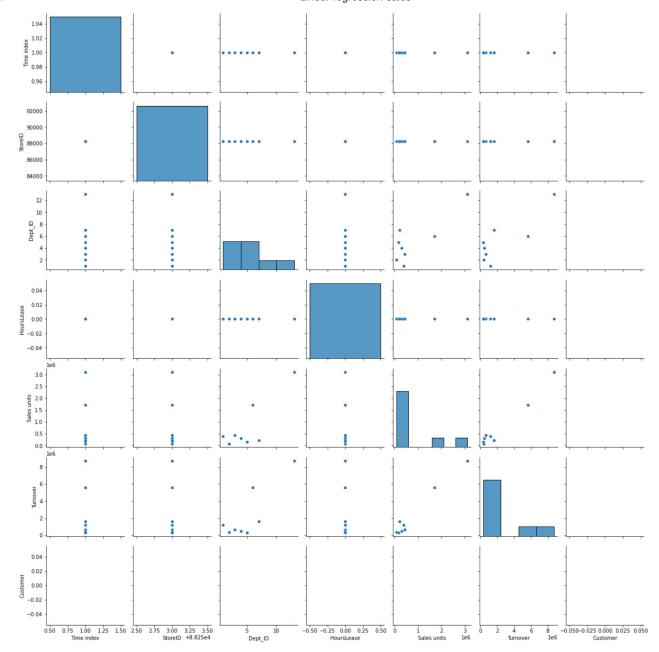
```
In [8]: a.columns

Out[8]: Index(['MonthYear', 'Time index', 'Country', 'StoreID', 'City', 'Dept ID',
```

EDA and Visualization

```
In [9]: sns.pairplot(da)
```

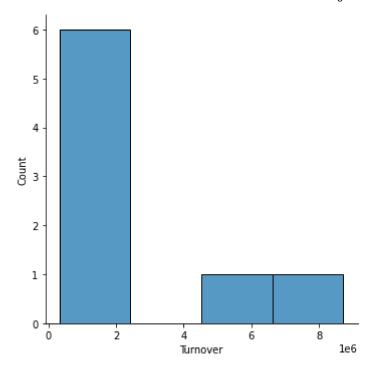
Out[9]: <seaborn.axisgrid.PairGrid at 0x21954640730>



distribution plot

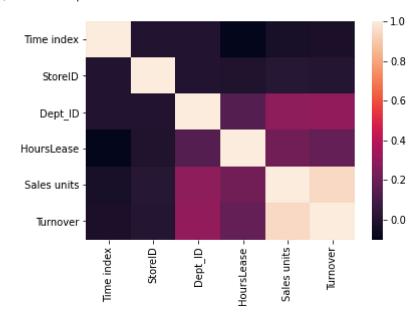
```
In [10]: sns.displot(a['Turnover'])
```

Out[10]: <seaborn.axisgrid.FacetGrid at 0x2195b0b6280>



correlation

Out[11]: <AxesSubplot:>



To train the model-Model Building

```
In [14]:
           # to split my dataset into training and test 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 [15]:
           from sklearn.linear model import LinearRegression
           lr= LinearRegression()
          lr.fit(x_train,y_train)
Out[15]: LinearRegression()
In [16]:
           print(lr.intercept_)
          6.4
In [17]:
           coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
           coeff
Out[17]:
                  Co-efficient
          StoreID
                         0.0
In [18]:
           prediction=lr.predict(x_test)
           plt.scatter(y_test,prediction)
Out[18]: <matplotlib.collections.PathCollection at 0x2195d3804c0>
          6.7
          6.6
          6.5
          6.4
          6.3
          6.2
          6.1
              2.00
                    2.25
                         2.50
                               2.75
                                    3.00
                                          3.25
                                                3.50
                                                      3.75
                                                           4.00
In [19]:
           print(lr.score(x_test,y_test))
          -17.3400000000000003
```