Deena 20104016

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as pp
```

Problem Statement

LINEAR REGRESSION

In [2]: a = pd.read_csv("Salesworkload.csv")

Out[2]:

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

7658 rows × 14 columns

HEAD

In [3]: d=a.head(8)

Out[3]:

	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	۷
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	3
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.0	1
5	10.2016	1.0	United Kingdom	88253.0	London (I)	6.0	Meat	8270.316	0.0	17
6	10.2016	1.0	United Kingdom	88253.0	London (I)	13.0	Food	16468.251	0.0	31

Data Cleaning and Preprocessing

In [4]: b=d.dropna(axis=1)

Out[4]:

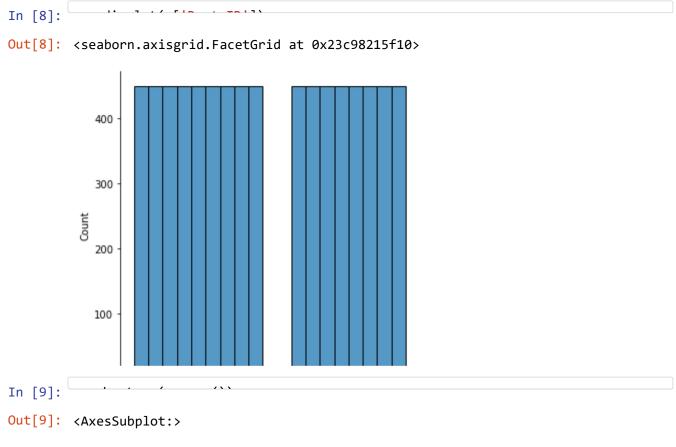
	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	
0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.0	:
1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.0	
2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.0	۷
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	3
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.0	1
5	10.2016	1.0	United Kingdom	88253.0	London (I)	6.0	Meat	8270.316	0.0	17
6	10.2016	1.0	United Kingdom	88253.0	London (I)	13.0	Food	16468.251	0.0	31

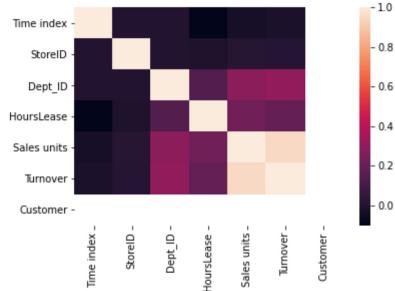
In [5]:

Out[5]:

	Time index	StoreID	Dept_ID	HoursLease	Sales units	Turnover	Custor
count	7650.000000	7650.000000	7650.000000	7650.000000	7.650000e+03	7.650000e+03	
mean	5.000000	61995.220000	9.470588	22.036078	1.076471e+06	3.721393e+06	١
std	2.582158	29924.581631	5.337429	133.299513	1.728113e+06	6.003380e+06	١
min	1.000000	12227.000000	1.000000	0.000000	0.000000e+00	0.000000e+00	١
25%	3.000000	29650.000000	5.000000	0.000000	5.457125e+04	2.726798e+05	١
50%	5.000000	75400.500000	9.000000	0.000000	2.932300e+05	9.319575e+05	١
75%	7.000000	87703.000000	14.000000	0.000000	9.175075e+05	3.264432e+06	١
max	9.000000	98422.000000	18.000000	3984.000000	1.124296e+07	4.271739e+07	١

To display heading





TO TRAIN THE MODEL - MODEL BUILDING

```
In [10]: x=b[['Dept_ID']]
In [11]: # 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 [12]: | from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
Out[12]: LinearRegression()
In [13]: coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
Out[13]:
                  Co-efficient
          Dept_ID
                         1.0
In [14]: prediction= lr.predict(x_test)
Out[14]: <matplotlib.collections.PathCollection at 0x23c9d314820>
          6
          5
          4
          3
In [15]: L
Out[15]: 1.0
```

RIDGE & LASSO

Out[19]: 0.6493652418247469

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