

```
In [2]: #PROBLEM STATEMENTS
#• What is the Highest and Lowest Sales units in the given dataset?
#• What is the Highest and Lowest Turnover in the given dataset?
#• What is the correlation between Turnover and sales unit?
#• Remove empty column in the given data set.

import numpy as np
import pandas as pd
```

```
In [3]: df=pd.read_csv(r"C:\Users\Jayadeep\Downloads\Salesworkload1 (1).csv")
df
```

Out[3]:

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	Sales units	Turnover	Customer	Area (m2)
0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.0	398560.0	1226244.0	NaN	953.04
1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.0	82725.0	387810.0	NaN	720.48
2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.0	438400.0	654657.0	NaN	966.72
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	309425.0	499434.0	NaN	1053.36
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.0	165515.0	329397.0	NaN	1053.36
...	...	...	...	...	...	...	...	...	...	...	...	...	...
7653	06.2017	9.0	Sweden	29650.0	Gothenburg	12.0	Checkout	6322.323	0.0	3886530.0	14538825.0	NaN	#NV
7654	06.2017	9.0	Sweden	29650.0	Gothenburg	16.0	Customer Services	4270.479	0.0	245.0	0.0	NaN	#NV
7655	06.2017	9.0	Sweden	29650.0	Gothenburg	11.0	Delivery	0	0.0	0.0	0.0	NaN	#NV
7656	06.2017	9.0	Sweden	29650.0	Gothenburg	17.0	others	2224.929	0.0	245.0	0.0	NaN	#NV
7657	06.2017	9.0	Sweden	29650.0	Gothenburg	18.0	all	39652.2	0.0	3886530.0	15056214.0	NaN	#NV

7658 rows × 14 columns



In [7]: `df.head(20)`

Out[7]:

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	Sales units	Turnover	Customer	Area (m2)	O1
0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.0	398560.0	1226244.0	NaN	953.04	
1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.0	82725.0	387810.0	NaN	720.48	
2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.0	438400.0	654657.0	NaN	966.72	
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.0	309425.0	499434.0	NaN	1053.36	
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.0	165515.0	329397.0	NaN	1053.36	
5	10.2016	1.0	United Kingdom	88253.0	London (I)	6.0	Meat	8270.316	0.0	1713310.0	5617137.0	NaN	11735.16	
6	10.2016	1.0	United Kingdom	88253.0	London (I)	13.0	Food	16468.251	0.0	3107935.0	8714679.0	NaN	19865.64	
7	10.2016	1.0	United Kingdom	88253.0	London (I)	7.0	Clothing	4698.471	0.0	213680.0	1615341.0	NaN	8513.52	
8	10.2016	1.0	United Kingdom	88253.0	London (I)	8.0	Household	1183.272	0.0	54915.0	290400.0	NaN	4842.72	
9	10.2016	1.0	United Kingdom	88253.0	London (I)	9.0	Hardware	2029.815	0.0	59260.0	450015.0	NaN	5608.8	
10	10.2016	1.0	United Kingdom	88253.0	London (I)	14.0	Non Food	7911.558	0.0	327855.0	2355756.0	NaN	19238.64	
11	10.2016	1.0	United Kingdom	88253.0	London (I)	15.0	Admin	4308.243	0.0	0.0	0.0	NaN	0	
12	10.2016	1.0	United Kingdom	88253.0	London (I)	12.0	Checkout	5825.097	0.0	3435790.0	11070435.0	NaN	39104.28	
13	10.2016	1.0	United Kingdom	88253.0	London (I)	16.0	Customer Services	3320.085	0.0	0.0	0.0	NaN	0	
14	10.2016	1.0	United Kingdom	88253.0	London (I)	11.0	Delivery	0	0.0	0.0	0.0	NaN	0	
15	10.2016	1.0	United Kingdom	88253.0	London (I)	17.0	others	2253.252	0.0	0.0	0.0	NaN	0	

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	Sales units	Turnover	Customer	Area (m2)	OI
<b>16</b>	10.2016	1.0	United Kingdom	88253.0	London (I)	18.0	all	40086.486	0.0	3435790.0	11070435.0	NaN	39104.28	
<b>17</b>	10.2016	1.0	United Kingdom	38976.0	Manchester	1.0	Dry	2583.687	0.0	754600.0	2648175.0	NaN	1404.48	
<b>18</b>	10.2016	1.0	United Kingdom	38976.0	Manchester	2.0	Frozen	5145.345	0.0	216925.0	1291830.0	NaN	1057.92	
<b>19</b>	10.2016	1.0	United Kingdom	38976.0	Manchester	3.0	other	47.205	0.0	1111835.0	1822749.0	NaN	1235.76	

```
In [8]: df.tail(20)
```

```
Out[8]:
```

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLease	Sales units	Turnover	Customer	Area (m2)	Open h
<b>7638</b>	06.2017	9.0	Sweden	90992.0	Malmö	11.0	Delivery	0	0.0	0.0	0.0	NaN	#NV	Tj
<b>7639</b>	06.2017	9.0	Sweden	90992.0	Malmö	17.0	others	1935.405	0.0	240.0	0.0	NaN	#NV	Tj
<b>7640</b>	06.2017	9.0	Sweden	90992.0	Malmö	18.0	all	40133.691	0.0	4363130.0	18666582.0	NaN	#NV	Tj
<b>7641</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	1.0	Dry	1929.111	0.0	454145.0	1815402.0	NaN	#NV	Tj
<b>7642</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	2.0	Frozen	2457.807	0.0	79500.0	466167.0	NaN	#NV	Tj
<b>7643</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	3.0	other	47.205	0.0	511770.0	692709.0	NaN	#NV	Tj
<b>7644</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	4.0	Fish	1689.939	0.0	363010.0	632217.0	NaN	#NV	Tj
<b>7645</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	5.0	Fruits & Vegetables	2567.952	0.0	270165.0	514305.0	NaN	#NV	Tj
<b>7646</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	6.0	Meat	8210.523	0.0	1895090.0	7750254.0	NaN	#NV	Tj
<b>7647</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	13.0	Food	16902.537	0.0	3573680.0	12223593.0	NaN	#NV	Tj
<b>7648</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	7.0	Clothing	3587.58	0.0	183700.0	1693797.0	NaN	#NV	Tj
<b>7649</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	8.0	Household	1312.299	0.0	58045.0	575979.0	NaN	#NV	Tj
<b>7650</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	9.0	Hardware	1598.676	0.0	71105.0	582849.0	NaN	#NV	Tj
<b>7651</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	14.0	Non Food	6498.555	0.0	312850.0	2855895.0	NaN	#NV	Tj
<b>7652</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	15.0	Admin	3433.377	0.0	245.0	0.0	NaN	#NV	Tj
<b>7653</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	12.0	Checkout	6322.323	0.0	3886530.0	14538825.0	NaN	#NV	Tj
<b>7654</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	16.0	Customer Services	4270.479	0.0	245.0	0.0	NaN	#NV	Tj
<b>7655</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	11.0	Delivery	0	0.0	0.0	0.0	NaN	#NV	Tj
<b>7656</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	17.0	others	2224.929	0.0	245.0	0.0	NaN	#NV	Tj
<b>7657</b>	06.2017	9.0	Sweden	29650.0	Gothenburg	18.0	all	39652.2	0.0	3886530.0	15056214.0	NaN	#NV	Tj

```
In [9]: df.shape
```

```
Out[9]: (7658, 14)
```

In [10]: `df.describe`



```

Out[10]: <bound method NDFrame.describe of
0      10.2016      1.0  United Kingdom  88253.0  London (I)      1.0
1      10.2016      1.0  United Kingdom  88253.0  London (I)      2.0
2      10.2016      1.0  United Kingdom  88253.0  London (I)      3.0
3      10.2016      1.0  United Kingdom  88253.0  London (I)      4.0
4      10.2016      1.0  United Kingdom  88253.0  London (I)      5.0
...      ...      ...      ...      ...      ...
7653    06.2017      9.0      Sweden  29650.0  Gothenburg      12.0
7654    06.2017      9.0      Sweden  29650.0  Gothenburg      16.0
7655    06.2017      9.0      Sweden  29650.0  Gothenburg      11.0
7656    06.2017      9.0      Sweden  29650.0  Gothenburg      17.0
7657    06.2017      9.0      Sweden  29650.0  Gothenburg      18.0

      Dept. Name  HoursOwn  HoursLease  Sales units  Turnover \
0              Dry  3184.764          0.0    398560.0  1226244.0
1             Frozen  1582.941          0.0     82725.0   387810.0
2             other    47.205          0.0   438400.0   654657.0
3              Fish  1623.852          0.0   309425.0   499434.0
4  Fruits & Vegetables  1759.173          0.0   165515.0   329397.0
...      ...      ...      ...      ...      ...
7653          Checkout  6322.323          0.0  3886530.0  14538825.0
7654    Customer Services  4270.479          0.0      245.0      0.0
7655          Delivery      0          0.0      0.0      0.0
7656          others  2224.929          0.0      245.0      0.0
7657             all  39652.2          0.0  3886530.0  15056214.0

      Customer Area (m2) Opening hours
0      NaN    953.04      Type A
1      NaN    720.48      Type A
2      NaN    966.72      Type A
3      NaN   1053.36      Type A
4      NaN   1053.36      Type A
...      ...      ...      ...
7653    NaN    #NV      Type A
7654    NaN    #NV      Type A
7655    NaN    #NV      Type A
7656    NaN    #NV      Type A
7657    NaN    #NV      Type A

```

[7658 rows x 14 columns]>

```
In [11]: df.isna().any()
```

```
Out[11]: MonthYear      False  
Time index      True  
Country          True  
StoreID          True  
City             True  
Dept_ID          True  
Dept. Name       True  
HoursOwn         True  
HoursLease       True  
Sales units      True  
Turnover         True  
Customer         True  
Area (m2)        True  
Opening hours    True  
dtype: bool
```

In [12]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7658 entries, 0 to 7657
Data columns (total 14 columns):
#   Column          Non-Null Count  Dtype
---  -
0   MonthYear       7658 non-null   object
1   Time index      7650 non-null   float64
2   Country         7650 non-null   object
3   StoreID         7650 non-null   float64
4   City            7650 non-null   object
5   Dept_ID        7650 non-null   float64
6   Dept. Name      7650 non-null   object
7   HoursOwn        7650 non-null   object
8   HoursLease      7650 non-null   float64
9   Sales units     7650 non-null   float64
10  Turnover        7650 non-null   float64
11  Customer        0 non-null      float64
12  Area (m2)       7650 non-null   object
13  Opening hours   7650 non-null   object
dtypes: float64(7), object(7)
memory usage: 837.7+ KB
```

In [13]: `print(df.sum())`

```
MonthYear      10.201610.201610.201610.201610.201610.201610.2...
Time index      38250.0
StoreID         474263433.0
Dept_ID         72450.0
HoursLease      168576.0
Sales units     8235000965.0
Turnover       28468656015.0
Customer        0.0
dtype: object
```

C:\Users\Jayadeep\AppData\Local\Temp\ipykernel\_11736\760981365.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.

```
print(df.sum())
```

```
In [15]: print(df.mean())
```

```
Time index      5.000000e+00  
StoreID         6.199522e+04  
Dept_ID         9.470588e+00  
HoursLease      2.203608e+01  
Sales units     1.076471e+06  
Turnover        3.721393e+06  
Customer        NaN  
dtype: float64
```

C:\Users\Jayadeep\AppData\Local\Temp\ipykernel\_11736\2807316344.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.  
print(df.mean())

```
In [16]: print(df.max())
```

```
MonthYear      12.2016  
Time index      9.0  
StoreID        98422.0  
Dept_ID        18.0  
HoursLease     3984.0  
Sales units    11242955.0  
Turnover       42717390.0  
Customer        NaN  
dtype: object
```

C:\Users\Jayadeep\AppData\Local\Temp\ipykernel\_11736\2376050674.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.  
print(df.max())

```
In [ ]: #the highest sales units 11242955.0  
        #the highest turnover rate 42717390.0
```

```
In [17]: print(df.min())
```

```
MonthYear      - - - -  
Time index      1.0  
StoreID      12227.0  
Dept_ID        1.0  
HoursLease      0.0  
Sales units     0.0  
Turnover        0.0  
Customer       NaN  
dtype: object
```

C:\Users\Jayadeep\AppData\Local\Temp\ipykernel\_11736\3863265735.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.  
print(df.min())

```
In [ ]: #the lowest sales unit 0.0  
# the lowest turnover 0.0
```

```
In [18]: print(df.mode())
```

	MonthYear	Time index	Country	StoreID	City	Dept_ID \
0	01.2017	1.0	France	12227.0	Aalborg (I)	1.0
1	02.2017	2.0	Germany	15552.0	Aalborg (II)	2.0
2	03.2017	3.0	United Kingdom	16927.0	Amsterdam	3.0
3	04.2017	4.0	NaN	17647.0	Antwerp	4.0
4	05.2017	5.0	NaN	18808.0	Barcelona (I)	5.0
5	06.2017	6.0	NaN	19000.0	Barcelona (II)	6.0
6	10.2016	7.0	NaN	19340.0	Berlin (I)	7.0
7	11.2016	8.0	NaN	19769.0	Berlin (II)	8.0
8	12.2016	9.0	NaN	20166.0	Bilbao	9.0
9	NaN	NaN	NaN	20891.0	Birmingham	11.0
10	NaN	NaN	NaN	22117.0	Bologna	12.0
11	NaN	NaN	NaN	23623.0	Bordeaux	13.0
12	NaN	NaN	NaN	29650.0	Brno	14.0
13	NaN	NaN	NaN	32949.0	Brussels (I)	15.0
14	NaN	NaN	NaN	34378.0	Brussels (II)	16.0
15	NaN	NaN	NaN	38560.0	Cologne	17.0
16	NaN	NaN	NaN	38976.0	Copenhagen (I)	18.0
17	NaN	NaN	NaN	42367.0	Copenhagen (II)	NaN
18	NaN	NaN	NaN	45500.0	Copenhagen (III)	NaN

```
In [19]: df[['Turnover', 'Sales units']]
```

```
Out[19]:
```

	Turnover	Sales units
0	1226244.0	398560.0
1	387810.0	82725.0
2	654657.0	438400.0
3	499434.0	309425.0
4	329397.0	165515.0
...	...	...
7653	14538825.0	3886530.0
7654	0.0	245.0
7655	0.0	0.0
7656	0.0	245.0
7657	15056214.0	3886530.0

7658 rows × 2 columns

```
In [20]: df2=df[['Turnover', 'Sales units']]
print(df2.corr())
```

	Turnover	Sales units
Turnover	1.000000	0.947374
Sales units	0.947374	1.000000

```
In [ ]: #The correlation relation between Turnover and sales unit is 1.0 0.9
          0.0 1.0
```

```
In [16]: del df['Customer']
```

```
In [17]: print(df)
```



	MonthYear	Time index	Country	StoreID	City	Dept_ID	\
0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	
1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	
2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	
...	...	...	...	...	...	...	
7653	06.2017	9.0	Sweden	29650.0	Gothenburg	12.0	
7654	06.2017	9.0	Sweden	29650.0	Gothenburg	16.0	
7655	06.2017	9.0	Sweden	29650.0	Gothenburg	11.0	
7656	06.2017	9.0	Sweden	29650.0	Gothenburg	17.0	
7657	06.2017	9.0	Sweden	29650.0	Gothenburg	18.0	

	Dept. Name	HoursOwn	HoursLease	Sales units	Turnover	\
0	Dry	3184.764	0.0	398560.0	1226244.0	
1	Frozen	1582.941	0.0	82725.0	387810.0	
2	other	47.205	0.0	438400.0	654657.0	
3	Fish	1623.852	0.0	309425.0	499434.0	
4	Fruits & Vegetables	1759.173	0.0	165515.0	329397.0	
...	...	...	...	...	...	
7653	Checkout	6322.323	0.0	3886530.0	14538825.0	
7654	Customer Services	4270.479	0.0	245.0	0.0	
7655	Delivery	0	0.0	0.0	0.0	
7656	others	2224.929	0.0	245.0	0.0	
7657	all	39652.2	0.0	3886530.0	15056214.0	

	Area (m2)	Opening hours
0	953.04	Type A
1	720.48	Type A
2	966.72	Type A
3	1053.36	Type A
4	1053.36	Type A
...	...	...
7653	#NV	Type A
7654	#NV	Type A
7655	#NV	Type A
7656	#NV	Type A
7657	#NV	Type A

[7658 rows x 13 columns]

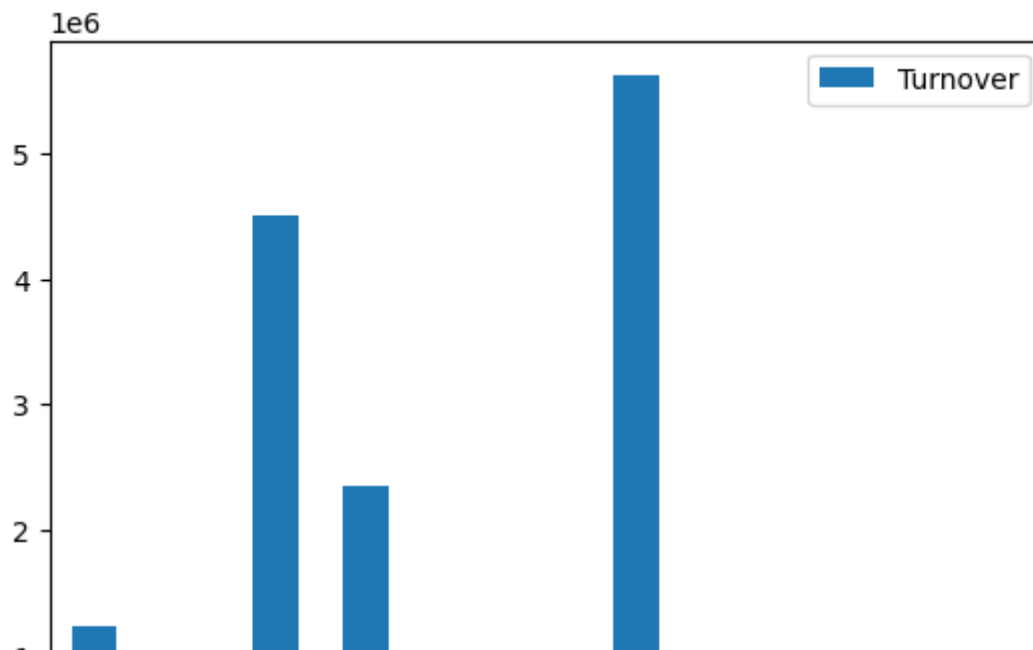
```
In [ ]: #Deleting empty column in the data set
```

```
In [18]: df.shape
```

```
Out[18]: (7658, 13)
```

```
In [6]: import pandas as pd  
import matplotlib.pyplot as plt
```

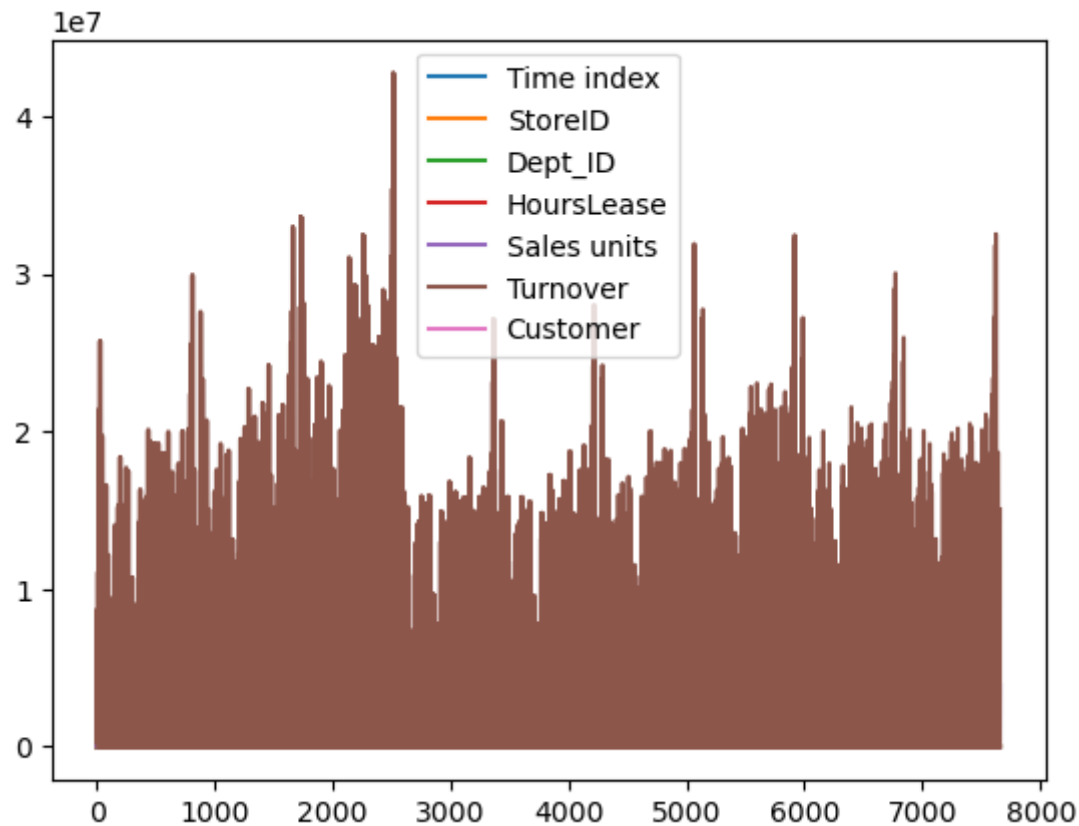
```
In [7]: df=pd.DataFrame({"Countries":["united kingdom",'poland','the nether lands','czech republic','den mark','spain','italy',  
                                "Turnover":[1226244,499439,4500150,2355756,329397,654657,5617137,290400,0.0,0.0,34523]})  
df.plot(x="Countries",y="Turnover",kind="bar")  
plt.show()
```



```
In [4]: df.plot()  
plt.show()
```

```
-----  
NameError                                Traceback (most recent call last)  
~\AppData\Local\Temp\ipykernel_16612\2546976976.py in <module>  
      1 df.plot()  
----> 2 plt.show()
```

NameError: name 'plt' is not defined



Answers:

1.The highest sales unit in the given data set is [11242955.0](#),and the lowest sales unit is 0.0

- 2.The highest turnover rate [42717390.0](#) and the lowest sales unit is 0.0
- 3.The correlation relation between Turnover and sales unit is

	Turnover	Sales units
Turnover	1.000000	0.947374
Sales units	0.947374	1.000000