

```
In [1]: import numpy as np
import pandas as pd
from numpy import linalg as lg
from numpy import cov
```

5 Data Sets

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

```
Out[2]:
```

	MonthYear	Time index	Country	StoreID	City	Dept_ID	Dept. Name	HoursOwn	HoursLeas
0	10.2016	1.0	United Kingdom	88253.0	London (I)	1.0	Dry	3184.764	0.
1	10.2016	1.0	United Kingdom	88253.0	London (I)	2.0	Frozen	1582.941	0.
2	10.2016	1.0	United Kingdom	88253.0	London (I)	3.0	other	47.205	0.
3	10.2016	1.0	United Kingdom	88253.0	London (I)	4.0	Fish	1623.852	0.
4	10.2016	1.0	United Kingdom	88253.0	London (I)	5.0	Fruits & Vegetables	1759.173	0.
...
7653	06.2017	9.0	Sweden	29650.0	Gothenburg	12.0	Checkout	6322.323	0.
7654	06.2017	9.0	Sweden	29650.0	Gothenburg	16.0	Customer Services	4270.479	0.
7655	06.2017	9.0	Sweden	29650.0	Gothenburg	11.0	Delivery	0	0.
7656	06.2017	9.0	Sweden	29650.0	Gothenburg	17.0	others	2224.929	0.
7657	06.2017	9.0	Sweden	29650.0	Gothenburg	18.0	all	39652.2	0.

7658 rows × 14 columns



a) Find mean, median, mode and describe

```
In [3]: print(a.mean)
print("Median:")
print(a.median)
print("Mode:")
print(a.mode)
print("Describe")
print(a.describe())
```

```
<bound method NDFrame._add_numeric_operations.<locals>.mean of
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

	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]>

Median:

	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

	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]>

Mode:

```

<bound method DataFrame.mode of      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

```

```

      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]>

Describe

```

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

      Turnover  Customer
count  7.650000e+03      0.0
mean    3.721393e+06      NaN
std     6.003380e+06      NaN
min     0.000000e+00      NaN
25%     2.726798e+05      NaN
50%     9.319575e+05      NaN

```

75%3.264432e+06NaN

max4.271739e+07NaN

b) Find sum(), cumsum(), count, min and max values

In [4]:

print(a.sum())

MonthYear10.201610.201610.201610.201610.201610.201610.2...

Time index38250.0

StoreID474263433.0

Dept_ID72450.0

HoursLease168576.0

Sales units8235000965.0

Turnover28468656015.0

Customer0.0

dtype: object

In [10]:

b=a.head(90)

print(b.cumsum())

MonthYearTime index \

010.20161.0

110.201610.20162.0

210.201610.201610.20163.0

310.201610.201610.201610.20164.0

410.201610.201610.201610.20165.0

.....

8510.201610.201610.201610.201610.201610.201610.2...86.0

8610.201610.201610.201610.201610.201610.201610.2...87.0

8710.201610.201610.201610.201610.201610.201610.2...88.0

8810.201610.201610.201610.201610.201610.201610.2...89.0

8910.201610.201610.201610.201610.201610.201610.2...90.0

CountryStoreID \

0United Kingdom88253.0

1United KingdomUnited Kingdom176506.0

2United KingdomUnited KingdomUnited Kingdom264759.0

3United KingdomUnited KingdomUnited KingdomUnit...353012.0

4United KingdomUnited KingdomUnited KingdomUnit...441265.0

.....

85United KingdomUnited KingdomUnited KingdomUnit...4114822.0

86United KingdomUnited KingdomUnited KingdomUnit...4133630.0

87United KingdomUnited KingdomUnited KingdomUnit...4152438.0

88United KingdomUnited KingdomUnited KingdomUnit...4171246.0

89United KingdomUnited KingdomUnited KingdomUnit...4190054.0

CityDept_ID \

0London (I)1.0

1London (I)London (I)3.0

2London (I)London (I)London (I)6.0

3London (I)London (I)London (I)London (I)10.0

4London (I)London (I)London (I)London (I)London...15.0

.....

85London (I)London (I)London (I)London (I)London...806.0

86London (I)London (I)London (I)London (I)London...808.0

87London (I)London (I)London (I)London (I)London...811.0

88London (I)London (I)London (I)London (I)London...815.0

89London (I)London (I)London (I)London (I)London...820.0

Dept. Name \

0Dry

1DryFrozen

2DryFrozenother

```

3                               DryFrozenotherFish
4                               DryFrozenotherFishFruits & Vegetables
..                               ...
85 DryFrozenotherFishFruits & VegetablesMeatFoodC...
86 DryFrozenotherFishFruits & VegetablesMeatFoodC...
87 DryFrozenotherFishFruits & VegetablesMeatFoodC...
88 DryFrozenotherFishFruits & VegetablesMeatFoodC...
89 DryFrozenotherFishFruits & VegetablesMeatFoodC...

                                HoursOwn  HoursLease  \
0                                3184.764         0.0
1                                3184.7641582.941       0.0
2                                3184.7641582.94147.205   0.0
3                                3184.7641582.94147.2051623.852 0.0
4                                3184.7641582.94147.2051623.8521759.173 0.0
..                               ...
85 3184.7641582.94147.2051623.8521759.1738270.316... 6288.0
86 3184.7641582.94147.2051623.8521759.1738270.316... 6288.0
87 3184.7641582.94147.2051623.8521759.1738270.316... 6288.0
88 3184.7641582.94147.2051623.8521759.1738270.316... 6288.0
89 3184.7641582.94147.2051623.8521759.1738270.316... 6288.0

    Sales units      Turnover  Customer  \
0      398560.0      1226244.0      NaN
1      481285.0      1614054.0      NaN
2      919685.0      2268711.0      NaN
3     1229110.0      2768145.0      NaN
4     1394625.0      3097542.0      NaN
..           ...           ...
85  96145695.0  322567602.0      NaN
86  96238930.0  322624806.0      NaN
87  96728045.0  323392302.0      NaN
88  97071970.0  324047673.0      NaN
89  97227435.0  324399396.0      NaN

                                Area (m2)  \
0                                953.04
1                                953.04720.48
2                                953.04720.48966.72
3                                953.04720.48966.721053.36
4                                953.04720.48966.721053.361053.36
..                               ...
85 953.04720.48966.721053.361053.3611735.1619865....
86 953.04720.48966.721053.361053.3611735.1619865....
87 953.04720.48966.721053.361053.3611735.1619865....
88 953.04720.48966.721053.361053.3611735.1619865....
89 953.04720.48966.721053.361053.3611735.1619865....

                                Opening hours
0                                Type A
1                                Type AType A
2                                Type AType AType A
3                                Type AType AType AType A
4                                Type AType AType AType AType A
..                               ...
85 Type AType AType AType AType AType AType AType AType...
86 Type AType AType AType AType AType AType AType AType...
87 Type AType AType AType AType AType AType AType AType...
88 Type AType AType AType AType AType AType AType AType...
89 Type AType AType AType AType AType AType AType AType...

```

[90 rows x 14 columns]

In [6]:

```

print(a.count())
print(a.min())
print(a.max())

```

```

MonthYear      7658
Time index     7650
Country        7650
StoreID        7650
City           7650
Dept_ID        7650
Dept. Name     7650
HoursOwn       7650
HoursLease     7650
Sales units    7650
Turnover       7650
Customer       0
Area (m2)      7650
Opening hours  7650
dtype: int64
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
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) Find covariance and correlation (spearman and pearsons)

```
In [7]: print(a.cov())
```

```

              Time index      StoreID      Dept_ID      HoursLease  \
Time index  6.667538e+00  2.094019e-10 -1.430561e-14 -3.438253e+01
StoreID     2.094019e-10  8.954806e+08  1.814186e-11 -4.798921e+04
Dept_ID     -1.430561e-14  1.814186e-11  2.848815e+01  9.613363e+01
HoursLease  -3.438253e+01 -4.798921e+04  9.613363e+01  1.776876e+04
Sales units -1.552564e+05  6.869293e+08  2.645877e+06  4.927821e+07
Turnover    -3.152632e+05  1.653862e+09  9.630443e+06  1.495198e+08
Customer              NaN              NaN              NaN              NaN

              Sales units      Turnover      Customer
Time index -1.552564e+05 -3.152632e+05          NaN
StoreID     6.869293e+08  1.653862e+09          NaN
Dept_ID     2.645877e+06  9.630443e+06          NaN
HoursLease  4.927821e+07  1.495198e+08          NaN
Sales units 2.986375e+12  9.828553e+12          NaN
Turnover    9.828553e+12  3.604057e+13          NaN
Customer              NaN              NaN          NaN

```

```
In [19]: from scipy.stats import pearsonr
print(pearsonr(b['StoreID'],b['MonthYear']))
```

```
(nan, nan)
```

```
In [11]: from scipy.stats import spearmanr
print(spearmanr(a['Time index'],a['MonthYear']))
```

```
SpearmanrResult(correlation=-0.3601703830417159, pvalue=2.5854375801641448e-233)
```

In []: