

```
In [1]: import pandas as pd
```

```
In [2]: hsp=pd.read_csv("HousePrices.csv")
```

In [3]: hsp.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2073 entries, 0 to 2072
Data columns (total 81 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Id                     2073 non-null   int64
1   Dwell_Type             2073 non-null   int64
2   Zone_Class             2073 non-null   object
3   LotFrontage            1753 non-null   float64
4   LotArea                2073 non-null   int64
5   Road_Type              2073 non-null   object
6   Alley                  129 non-null    object
7   Property_Shape         2073 non-null   object
8   LandContour            2073 non-null   object
9   Utilities              2073 non-null   object
10  LotConfig              2073 non-null   object
11  LandSlope              2073 non-null   object
12  Neighborhood            2073 non-null   object
13  Condition1             2073 non-null   object
14  Condition2             2073 non-null   object
15  Dwelling_Type          2073 non-null   object
16  HouseStyle             2073 non-null   object
17  OverallQual            2073 non-null   int64
18  OverallCond            2073 non-null   int64
19  YearBuilt              2073 non-null   int64
20  YearRemodAdd           2073 non-null   int64
21  RoofStyle              2073 non-null   object
22  RoofMatl               2073 non-null   object
23  Exterior1st            2073 non-null   object
24  Exterior2nd            2073 non-null   object
25  MasVnrType             2059 non-null   object
26  MasVnrArea             2059 non-null   float64
27  ExterQual              2073 non-null   object
28  ExterCond              2073 non-null   object
29  Foundation              2073 non-null   object
30  BsmtQual               2014 non-null   object
31  BsmtCond               2014 non-null   object
32  BsmtExposure           2012 non-null   object
33  BsmtFinType1           2014 non-null   object
```

34	BsmtFinSF1	2073	non-null	int64
35	BsmtFinType2	2013	non-null	object
36	BsmtFinSF2	2073	non-null	int64
37	BsmtUnfSF	2073	non-null	int64
38	TotalBsmtSF	2073	non-null	int64
39	Heating	2073	non-null	object
40	HeatingQC	2073	non-null	object
41	CentralAir	2073	non-null	object
42	Electrical	2072	non-null	object
43	1stFlrSF	2073	non-null	int64
44	2ndFlrSF	2073	non-null	int64
45	LowQualFinSF	2073	non-null	int64
46	GrLivArea	2073	non-null	int64
47	BsmtFullBath	2073	non-null	int64
48	BsmtHalfBath	2073	non-null	int64
49	FullBath	2073	non-null	int64
50	HalfBath	2073	non-null	int64
51	BedroomAbvGr	2073	non-null	int64
52	KitchenAbvGr	2073	non-null	int64
53	KitchenQual	2073	non-null	object
54	TotRmsAbvGrd	2073	non-null	int64
55	Functional	2073	non-null	object
56	Fireplaces	2073	non-null	int64
57	FireplaceQu	1085	non-null	object
58	GarageType	1960	non-null	object
59	GarageYrBlt	1960	non-null	float64
60	GarageFinish	1960	non-null	object
61	GarageCars	2073	non-null	int64
62	GarageArea	2073	non-null	int64
63	GarageQual	1960	non-null	object
64	GarageCond	1960	non-null	object
65	PavedDrive	2073	non-null	object
66	WoodDeckSF	2073	non-null	int64
67	OpenPorchSF	2073	non-null	int64
68	EnclosedPorch	2073	non-null	int64
69	3SsnPorch	2073	non-null	int64
70	ScreenPorch	2073	non-null	int64
71	PoolArea	2073	non-null	int64
72	PoolQC	8	non-null	object
73	Fence	404	non-null	object
74	MiscFeature	80	non-null	object
75	MiscVal	2073	non-null	int64

```

76  MoSold          2073 non-null  int64
77  YrSold          2073 non-null  int64
78  SaleType        2073 non-null  object
79  SaleCondition    2073 non-null  object
80  Property_Sale_Price 2073 non-null  int64
dtypes: float64(3), int64(35), object(43)
memory usage: 1.3+ MB

```

```
In [4]: cl_to_drop=["LotFrontage","MasVnrArea","Alley","BsmtExposure","BsmtFinType1","BsmtFinType2","BsmtQual","BsmtCond","PoolQual"]
```

```
In [5]: hsp.drop(cl_to_drop,axis=1,inplace=True)
```

```
In [6]: hsp
```

Out[6]:

	Id	Dwell_Type	Zone_Class	LotArea	Road_Type	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	...	EnclosedPorch	3SsnPor
0	1	60	RL	8450	Pave	Lvl	AllPub	Inside	Gtl	CollgCr	...	0	
1	2	20	RL	9600	Pave	Lvl	AllPub	FR2	Gtl	Veenker	...	0	
2	3	60	RL	11250	Pave	Lvl	AllPub	Inside	Gtl	CollgCr	...	0	
3	4	70	RL	9550	Pave	Lvl	AllPub	Corner	Gtl	Crawfor	...	272	
4	5	60	RL	14260	Pave	Lvl	AllPub	FR2	Gtl	NoRidge	...	0	
...	
2068	942	60	RL	8755	Pave	Lvl	AllPub	FR2	Gtl	Gilbert	...	0	1
2069	943	90	RL	7711	Pave	Lvl	AllPub	Inside	Gtl	Edwards	...	0	
2070	944	90	RL	25000	Pave	Low	AllPub	Inside	Gtl	Mitchel	...	0	
2071	945	20	RL	14375	Pave	Lvl	NoSeWa	CulDSac	Gtl	Timber	...	0	
2072	946	50	RM	8820	Pave	Lvl	AllPub	Corner	Gtl	OldTown	...	244	

2073 rows × 63 columns

```
In [7]: import sklearn  
        from sklearn.preprocessing import LabelEncoder
```

```
In [8]: col_name_objects = hsp.select_dtypes(include=['object']).columns.values
LE = LabelEncoder()
for col in col_name_objects:
    hsp[col] = LE.fit_transform(hsp[col])
hsp.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2073 entries, 0 to 2072
Data columns (total 63 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Id                    2073 non-null   int64
1   Dwell_Type            2073 non-null   int64
2   Zone_Class            2073 non-null   int32
3   LotArea               2073 non-null   int64
4   Road_Type            2073 non-null   int32
5   LandContour          2073 non-null   int32
6   Utilities            2073 non-null   int32
7   LotConfig            2073 non-null   int32
8   LandSlope            2073 non-null   int32
9   Neighborhood         2073 non-null   int32
10  Condition1           2073 non-null   int32
11  Condition2           2073 non-null   int32
12  Dwelling_Type        2073 non-null   int32
13  HouseStyle           2073 non-null   int32
14  OverallQual          2073 non-null   int64
15  OverallCond          2073 non-null   int64
16  YearBuilt            2073 non-null   int64
17  YearRemodAdd         2073 non-null   int64
18  RoofStyle            2073 non-null   int32
19  RoofMatl            2073 non-null   int32
20  Exterior1st          2073 non-null   int32
21  Exterior2nd          2073 non-null   int32
22  MasVnrType           2073 non-null   int32
23  ExterQual            2073 non-null   int32
24  ExterCond            2073 non-null   int32
25  Foundation           2073 non-null   int32
26  BsmtFinSF1           2073 non-null   int64
27  BsmtFinSF2           2073 non-null   int64
28  BsmtUnfSF            2073 non-null   int64
29  TotalBsmtSF          2073 non-null   int64
```

30	Heating	2073	non-null	int32
31	HeatingQC	2073	non-null	int32
32	CentralAir	2073	non-null	int32
33	Electrical	2073	non-null	int32
34	1stFlrSF	2073	non-null	int64
35	2ndFlrSF	2073	non-null	int64
36	LowQualFinSF	2073	non-null	int64
37	GrLivArea	2073	non-null	int64
38	BsmtFullBath	2073	non-null	int64
39	BsmtHalfBath	2073	non-null	int64
40	FullBath	2073	non-null	int64
41	HalfBath	2073	non-null	int64
42	BedroomAbvGr	2073	non-null	int64
43	KitchenAbvGr	2073	non-null	int64
44	KitchenQual	2073	non-null	int32
45	TotRmsAbvGrd	2073	non-null	int64
46	Functional	2073	non-null	int32
47	Fireplaces	2073	non-null	int64
48	GarageCars	2073	non-null	int64
49	GarageArea	2073	non-null	int64
50	PavedDrive	2073	non-null	int32
51	WoodDeckSF	2073	non-null	int64
52	OpenPorchSF	2073	non-null	int64
53	EnclosedPorch	2073	non-null	int64
54	3SsnPorch	2073	non-null	int64
55	ScreenPorch	2073	non-null	int64
56	PoolArea	2073	non-null	int64
57	MiscVal	2073	non-null	int64
58	MoSold	2073	non-null	int64
59	YrSold	2073	non-null	int64
60	SaleType	2073	non-null	int32
61	SaleCondition	2073	non-null	int32
62	Property_Sale_Price	2073	non-null	int64

dtypes: int32(28), int64(35)

memory usage: 793.7 KB

```
In [9]: hsp.iloc[:,22:]
```

```
Out[9]:
```

	MasVnrType	ExterQual	ExterCond	Foundation	BsmtFinSF1	BsmtFinSF2	BsmtUnfSF	TotalBsmtSF	Heating	HeatingQC	...	EnclosedPorch
0	1	2	4	2	706	0	150	856	1	0	...	0
1	2	3	4	1	978	0	284	1262	1	0	...	0
2	1	2	4	2	486	0	434	920	1	0	...	0
3	2	3	4	0	216	0	540	756	1	2	...	272
4	1	2	4	2	655	0	490	1145	1	0	...	0
...
2068	1	2	4	2	772	0	220	992	1	0	...	0
2069	2	3	4	2	1440	0	0	1440	1	4	...	0
2070	2	3	4	1	0	0	1632	1632	1	4	...	0
2071	1	3	4	1	111	354	354	819	1	2	...	0
2072	2	3	4	0	1088	0	0	1088	1	4	...	244

2073 rows × 41 columns



In [10]: hsp.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2073 entries, 0 to 2072
Data columns (total 63 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Id                     2073 non-null   int64
1   Dwell_Type             2073 non-null   int64
2   Zone_Class             2073 non-null   int32
3   LotArea                2073 non-null   int64
4   Road_Type              2073 non-null   int32
5   LandContour            2073 non-null   int32
6   Utilities              2073 non-null   int32
7   LotConfig              2073 non-null   int32
8   LandSlope              2073 non-null   int32
9   Neighborhood           2073 non-null   int32
10  Condition1             2073 non-null   int32
11  Condition2             2073 non-null   int32
12  Dwelling_Type          2073 non-null   int32
13  HouseStyle             2073 non-null   int32
14  OverallQual            2073 non-null   int64
15  OverallCond            2073 non-null   int64
16  YearBuilt              2073 non-null   int64
17  YearRemodAdd           2073 non-null   int64
18  RoofStyle              2073 non-null   int32
19  RoofMatl               2073 non-null   int32
20  Exterior1st            2073 non-null   int32
21  Exterior2nd            2073 non-null   int32
22  MasVnrType             2073 non-null   int32
23  ExterQual              2073 non-null   int32
24  ExterCond              2073 non-null   int32
25  Foundation             2073 non-null   int32
26  BsmtFinSF1             2073 non-null   int64
27  BsmtFinSF2             2073 non-null   int64
28  BsmtUnfSF              2073 non-null   int64
29  TotalBsmtSF            2073 non-null   int64
30  Heating                2073 non-null   int32
31  HeatingQC              2073 non-null   int32
32  CentralAir             2073 non-null   int32
33  Electrical             2073 non-null   int32
```

```
34 1stFlrSF          2073 non-null int64
35 2ndFlrSF          2073 non-null int64
36 LowQualFinSF      2073 non-null int64
37 GrLivArea         2073 non-null int64
38 BsmtFullBath       2073 non-null int64
39 BsmtHalfBath       2073 non-null int64
40 FullBath           2073 non-null int64
41 HalfBath           2073 non-null int64
42 BedroomAbvGr      2073 non-null int64
43 KitchenAbvGr       2073 non-null int64
44 KitchenQual        2073 non-null int32
45 TotRmsAbvGrd       2073 non-null int64
46 Functional         2073 non-null int32
47 Fireplaces         2073 non-null int64
48 GarageCars         2073 non-null int64
49 GarageArea         2073 non-null int64
50 PavedDrive         2073 non-null int32
51 WoodDeckSF         2073 non-null int64
52 OpenPorchSF        2073 non-null int64
53 EnclosedPorch      2073 non-null int64
54 3SsnPorch          2073 non-null int64
55 ScreenPorch        2073 non-null int64
56 PoolArea           2073 non-null int64
57 MiscVal            2073 non-null int64
58 MoSold             2073 non-null int64
59 YrSold             2073 non-null int64
60 SaleType           2073 non-null int32
61 SaleCondition      2073 non-null int32
62 Property_Sale_Price 2073 non-null int64
```

dtypes: int32(28), int64(35)

memory usage: 793.7 KB

```
In [11]: c_to_mean=["MasVnrType"]
```

```
In [12]: hsp["Electrical"]=hsp["Electrical"].fillna(0)
```

```
In [13]: import numpy as np
```

```
In [14]: hsp[c_to_mean]=hsp[c_to_mean].fillna(np.mean)
```

In [15]: hsp.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2073 entries, 0 to 2072
Data columns (total 63 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Id                     2073 non-null   int64
1   Dwell_Type             2073 non-null   int64
2   Zone_Class             2073 non-null   int32
3   LotArea                2073 non-null   int64
4   Road_Type              2073 non-null   int32
5   LandContour            2073 non-null   int32
6   Utilities              2073 non-null   int32
7   LotConfig              2073 non-null   int32
8   LandSlope              2073 non-null   int32
9   Neighborhood           2073 non-null   int32
10  Condition1             2073 non-null   int32
11  Condition2             2073 non-null   int32
12  Dwelling_Type          2073 non-null   int32
13  HouseStyle             2073 non-null   int32
14  OverallQual            2073 non-null   int64
15  OverallCond            2073 non-null   int64
16  YearBuilt              2073 non-null   int64
17  YearRemodAdd           2073 non-null   int64
18  RoofStyle              2073 non-null   int32
19  RoofMatl               2073 non-null   int32
20  Exterior1st            2073 non-null   int32
21  Exterior2nd            2073 non-null   int32
22  MasVnrType             2073 non-null   int32
23  ExterQual              2073 non-null   int32
24  ExterCond              2073 non-null   int32
25  Foundation             2073 non-null   int32
26  BsmtFinSF1             2073 non-null   int64
27  BsmtFinSF2             2073 non-null   int64
28  BsmtUnfSF              2073 non-null   int64
29  TotalBsmtSF            2073 non-null   int64
30  Heating                2073 non-null   int32
31  HeatingQC              2073 non-null   int32
32  CentralAir             2073 non-null   int32
33  Electrical             2073 non-null   int32
```

34	1stFlrSF	2073	non-null	int64
35	2ndFlrSF	2073	non-null	int64
36	LowQualFinSF	2073	non-null	int64
37	GrLivArea	2073	non-null	int64
38	BsmtFullBath	2073	non-null	int64
39	BsmtHalfBath	2073	non-null	int64
40	FullBath	2073	non-null	int64
41	HalfBath	2073	non-null	int64
42	BedroomAbvGr	2073	non-null	int64
43	KitchenAbvGr	2073	non-null	int64
44	KitchenQual	2073	non-null	int32
45	TotRmsAbvGrd	2073	non-null	int64
46	Functional	2073	non-null	int32
47	Fireplaces	2073	non-null	int64
48	GarageCars	2073	non-null	int64
49	GarageArea	2073	non-null	int64
50	PavedDrive	2073	non-null	int32
51	WoodDeckSF	2073	non-null	int64
52	OpenPorchSF	2073	non-null	int64
53	EnclosedPorch	2073	non-null	int64
54	3SsnPorch	2073	non-null	int64
55	ScreenPorch	2073	non-null	int64
56	PoolArea	2073	non-null	int64
57	MiscVal	2073	non-null	int64
58	MoSold	2073	non-null	int64
59	YrSold	2073	non-null	int64
60	SaleType	2073	non-null	int32
61	SaleCondition	2073	non-null	int32
62	Property_Sale_Price	2073	non-null	int64

dtypes: int32(28), int64(35)

memory usage: 793.7 KB

```
In [16]: hsp.isnull().sum().head(50)
```

```
Out[16]: Id                0
Dwell_Type                0
Zone_Class                0
LotArea                   0
Road_Type                 0
LandContour              0
Utilities                 0
LotConfig                 0
LandSlope                 0
Neighborhood              0
Condition1                0
Condition2                0
Dwelling_Type             0
HouseStyle                0
OverallQual               0
OverallCond               0
YearBuilt                 0
YearRemodAdd              0
RoofStyle                 0
RoofMatl                  0
Exterior1st               0
Exterior2nd               0
MasVnrType                0
ExterQual                 0
ExterCond                 0
Foundation                0
BsmtFinSF1                0
BsmtFinSF2                0
BsmtUnfSF                 0
TotalBsmtSF               0
Heating                   0
HeatingQC                 0
CentralAir                0
Electrical                0
1stFlrSF                  0
2ndFlrSF                  0
LowQualFinSF              0
GrLivArea                 0
BsmtFullBath              0
```

```

BsmtHalfBath    0
FullBath        0
HalfBath        0
BedroomAbvGr    0
KitchenAbvGr    0
KitchenQual     0
TotRmsAbvGrd    0
Functional      0
Fireplaces      0
GarageCars      0
GarageArea      0
dtype: int64

```

In [17]: hsp.describe()

Out[17]:

	Id	Dwell_Type	Zone_Class	LotArea	Road_Type	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	...
count	2073.000000	2073.000000	2073.000000	2073.000000	2073.000000	2073.000000	2073.000000	2073.000000	2073.000000	2073.000000	...
mean	916.132176	60.556199	3.027014	10717.853353	0.995176	2.768934	0.001447	2.992282	0.062229	12.273517	...
std	493.014670	159.924810	0.651138	9215.982306	0.069303	0.719395	0.038023	1.639268	0.268138	5.984291	...
min	1.000000	20.000000	0.000000	1300.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	...
25%	519.000000	20.000000	3.000000	7620.000000	1.000000	3.000000	0.000000	1.000000	0.000000	7.000000	...
50%	932.000000	50.000000	3.000000	9492.000000	1.000000	3.000000	0.000000	4.000000	0.000000	12.000000	...
75%	1302.000000	70.000000	3.000000	11601.000000	1.000000	3.000000	0.000000	4.000000	0.000000	17.000000	...
max	1820.000000	7080.000000	4.000000	215245.000000	1.000000	3.000000	1.000000	4.000000	2.000000	24.000000	...

8 rows × 63 columns



In [18]: hsp.shape

Out[18]: (2073, 63)

```
In [19]: hsp_corr=hsp.corr()
hsp_corr
```

Out[19]:

	Id	Dwell_Type	Zone_Class	LotArea	Road_Type	LandContour	Utilities	LotConfig	LandSlope	Neighborhood	...	E
Id	1.000000	0.043264	0.008184	0.004201	-0.009714	-0.023591	0.002230	0.015822	0.005695	-0.004485	...	
Dwell_Type	0.043264	1.000000	0.002442	-0.033609	-0.005419	0.010011	-0.009657	0.029538	0.068747	-0.020327	...	
Zone_Class	0.008184	0.002442	1.000000	-0.038714	0.056364	-0.000062	-0.001580	-0.017439	-0.017926	-0.242801	...	
LotArea	0.004201	-0.033609	-0.038714	1.000000	-0.144528	-0.141765	0.015111	-0.103496	0.357898	0.029985	...	
Road_Type	-0.009714	-0.005419	0.056364	-0.144528	1.000000	0.161557	0.002650	0.042154	-0.191610	0.007838	...	
...	
MoSold	0.000234	-0.020589	-0.062572	-0.001707	0.007443	0.007161	-0.074844	0.005100	-0.002735	0.013191	...	
YrSold	0.043219	0.018891	0.018558	-0.006478	-0.049654	0.033331	0.032878	-0.017574	-0.012042	0.055070	...	
SaleType	0.001829	0.011964	0.100362	0.005604	0.021432	-0.031928	-0.177783	0.022360	0.061987	-0.048993	...	
SaleCondition	-0.011822	0.005838	-0.007771	0.032000	-0.003179	0.023873	-0.125765	0.074189	-0.034126	0.006096	...	
Property_Sale_Price	-0.002491	-0.017399	-0.168772	0.211572	0.054667	0.027606	-0.021040	-0.061008	0.038989	0.199664	...	

63 rows × 63 columns



```
In [20]: x_ind=hsp.drop("Property_Sale_Price",axis=1)
y_dep=hsp.Property_Sale_Price
```

```
In [21]: from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x_ind, y_dep, test_size=0.2, random_state=2)
```

```
In [22]: from sklearn.preprocessing import StandardScaler
SE=StandardScaler()
```

```
In [23]: x_norm=SE.fit_transform(x_ind)
```



```
In [24]: from sklearn.decomposition import PCA
pca=PCA()
x_new=pca.fit_transform(x_norm)
```

```
In [25]: x_new
```

```
Out[25]: array([[ 1.84824349e+00,  5.59011719e-01, -1.99890156e+00, ...,
                 -1.16623086e-01, -3.21656412e-15, -1.02708580e-15],
                [-1.93829457e-01, -1.04516637e+00,  9.96206545e-01, ...,
                 1.53225426e-01,  7.00247657e-15,  6.70217178e-16],
                [ 2.07410097e+00,  4.42450697e-01, -1.81999917e+00, ...,
                 -1.90513304e-01, -6.92494848e-15, -7.69594997e-16],
                ...,
                [-2.31264441e-01,  1.58238102e+00,  9.21368060e-01, ...,
                 -1.88347921e-01,  8.50609627e-16, -1.67741439e-16],
                [-1.14325672e+00,  6.15411931e-01,  3.77225825e+00, ...,
                 -1.80843607e-02, -3.56923496e-16, -1.55034579e-17],
                [-1.66425357e+00,  4.45186117e-01,  2.16840995e+00, ...,
                 4.40974907e-01, -5.80273190e-16,  9.62128311e-17]])
```

```
In [27]: explained_variance=pca.explained_variance_ratio_  
         explained_variance
```

```
Out[27]: array([1.22313074e-01, 5.57268366e-02, 4.67845196e-02, 3.97304538e-02,  
                3.23012404e-02, 2.94044488e-02, 2.61980741e-02, 2.46682179e-02,  
                2.32654149e-02, 2.18114541e-02, 2.10036444e-02, 2.06357227e-02,  
                2.01021651e-02, 1.94196000e-02, 1.89834735e-02, 1.86974724e-02,  
                1.85140980e-02, 1.77101397e-02, 1.70785697e-02, 1.62798682e-02,  
                1.61895761e-02, 1.60302672e-02, 1.58778655e-02, 1.53591971e-02,  
                1.50214476e-02, 1.45889165e-02, 1.44702034e-02, 1.41937645e-02,  
                1.38122019e-02, 1.36574382e-02, 1.32763433e-02, 1.25941685e-02,  
                1.23348912e-02, 1.21659705e-02, 1.21055209e-02, 1.18597834e-02,  
                1.13574220e-02, 1.12031582e-02, 1.05312864e-02, 9.93820648e-03,  
                9.74987575e-03, 9.52897588e-03, 9.31929282e-03, 9.15011412e-03,  
                8.70182024e-03, 8.32817261e-03, 7.73209575e-03, 7.48727425e-03,  
                7.39776704e-03, 7.02149984e-03, 6.56854336e-03, 5.67697205e-03,  
                5.04655562e-03, 4.53141383e-03, 4.27295141e-03, 3.50144536e-03,  
                2.92655264e-03, 2.31556858e-03, 2.06138582e-03, 1.48560650e-03,  
                2.41220005e-32, 8.94768612e-34])
```

```
In [28]: ev=pd.DataFrame(explained_variance)
```

```
In [29]: ev.iloc[0:50].sum()
```

```
Out[29]: 0    0.961613  
         dtype: float64
```

```
In [30]: pca1=PCA(n_components=50)  
         x_new_model=pca1.fit_transform(x_norm)
```

```
In [31]: from sklearn.model_selection import train_test_split  
         x_train,x_test,y_train,y_test=train_test_split(x_new_model,y_dep,train_size=0.8,random_state=3)
```

```
In [32]: from sklearn import linear_model  
         from sklearn.linear_model import LinearRegression  
         model_li=LinearRegression()
```

```
In [33]: y_pred_li=model_li.fit(x_train,y_train).predict(x_test)
```

```
In [34]: model_li.score(x_test,y_test)
```

```
Out[34]: 0.8586939425814789
```

By using PCA with Linear regression,got the accuracy of 85%

```
In [ ]:
```