```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
```

In [2]: #already data is standardized with z score
df=pd.read\_csv("D:\python programs\Scaler classes\cars24-car-price-clean.csv")

In [3]: d:

Out[3]:

		selling_price	year	km_driven	mileage	engine	max_power	age	make	model	Indi
	0	-1.111046	-0.801317	1.195828	0.045745	-1.310754	-1.157780	0.801317	-0.433854	-1.125683	1.2
	1	-0.223944	0.450030	-0.737872	-0.140402	-0.537456	-0.360203	-0.450030	-0.327501	-0.333227	1.2
	2	-0.915058	-1.426990	0.035608	-0.582501	-0.537456	-0.404885	1.426990	-0.327501	-0.789807	1.2
	3	-0.892365	-0.801317	-0.409143	0.329620	-0.921213	-0.693085	0.801317	-0.433854	-0.905265	1.2
	4	-0.182683	0.137194	-0.544502	0.760085	0.042999	0.010435	-0.137194	-0.246579	-0.013096	-0.8
	•••										
198	815	-0.017641	0.762867	0.218923	0.950886	-0.215410	-0.694202	-0.762867	0.724475	0.091865	-0.8
198	816	0.549692	1.388540	-0.776546	-0.466159	-0.198054	-0.156899	-1.388540	-0.433854	0.112857	-0.8
198	817	-0.481822	0.137194	0.170967	0.380810	0.042999	0.120577	-0.137194	-0.240799	-0.448684	-0.8
198	818	1.168601	0.450030	72.355997	-0.815185	1.356256	0.935579	-0.450030	0.123346	0.328027	-0.8
198	819	1.117025	1.388540	-0.873231	-0.349818	0.041071	0.435139	-1.388540	-0.177218	1.036514	-0.8

19820 rows × 18 columns

## In [4]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 19820 entries, 0 to 19819
Data columns (total 18 columns):

#	Column	Non-Null Count	Dtype
0	selling_price	19820 non-null	float64
1	year	19820 non-null	float64
2	km_driven	19820 non-null	float64
3	mileage	19820 non-null	float64
4	engine	19820 non-null	float64
5	max_power	19820 non-null	float64
6	age	19820 non-null	float64
7	make	19820 non-null	float64
8	model	19820 non-null	float64
9	Individual	19820 non-null	float64
10	Trustmark Dealer	19820 non-null	float64
11	Diesel	19820 non-null	float64
12	Electric	19820 non-null	float64
13	LPG	19820 non-null	float64
14	Petrol	19820 non-null	float64
15	Manual	19820 non-null	float64
16	5	19820 non-null	float64
17	>5	19820 non-null	float64

```
dtypes: float64(18)
         memory usage: 2.7 MB
 In [5]: # model:x1
         # price:y
         # y=f(x)
         x=df['model'].values
         array([-1.12568266, -0.3332271 , -0.78980745, ..., -0.4486842 ,
Out[5]:
                 0.32802721, 1.03651397])
In [6]: y=df['selling_price'].values
         У
         array([-1.11104589, -0.22394353, -0.91505816, ..., -0.48182212,
Out[6]:
                 1.16860087, 1.11702515])
         Model=LinearRegression()
In [7]:
In [8]:
         type (Model)
         sklearn.linear model. base.LinearRegression
Out[8]:
In [9]:
         array([-1.12568266, -0.3332271 , -0.78980745, ..., -0.4486842 ,
Out[9]:
                 0.32802721, 1.03651397])
In [10]:
         array([-1.11104589, -0.22394353, -0.91505816, ..., -0.48182212,
Out[10]:
                 1.16860087, 1.11702515])
In [11]: | print(x.shape)
         print(y.shape)
         (19820,)
         (19820,)
In [12]: x=x.reshape(x.size,1)
         y=y.reshape(y.size,1)
In [13]: Model.fit(x,y)
Out[13]:
         ▼ LinearRegression
         LinearRegression()
In [14]: y hat=Model.predict(x)
         y_hat
         array([[-1.08634131],
Out[14]:
                [-0.32158118],
                [-0.76220457],
                [-0.43300319],
                [ 0.31656303],
                [ 1.00028896]])
In [15]:
         array([[-1.11104589],
Out[15]:
                [-0.22394353],
```

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[-0.48182212],
                   [ 1.16860087],
                   [ 1.11702515]])
          Model.intercept
In [16]:
          array([-1.18731936e-16])
Out[16]:
          Model.coef
In [17]:
          array([[0.96505112]])
Out[17]:
In [18]:
          # price=0.9650*model-1.1873
                                              written as manually
          fig=plt.figure()
          plt.scatter(x, y, label='data')
          plt.scatter(x,y hat,color='orange',label='prediction')
          plt.legend()
          plt.show()
            3
            2
            1
           0
                                                        data
          -1
                                                        prediction
          #R^2 coefficient of determination reliable method to compare two models quickly.
In [19]:
          Model.score(x,y)
          0.9313236629576508
Out[19]:
          df.head()
In [20]:
Out[20]:
             selling_price
                                   km_driven
                                                                                                    model Individua
                              year
                                               mileage
                                                          engine max_power
                                                                                   age
                                                                                           make
          0
                         -0.801317
                -1.111046
                                     1.195828
                                               0.045745
                                                       -1.310754
                                                                    -1.157780
                                                                              0.801317
                                                                                        -0.433854
                                                                                                 -1.125683
                                                                                                             1.24889
                -0.223944
                          0.450030
                                    -0.737872
                                              -0.140402
                                                       -0.537456
                                                                              -0.450030
                                                                                                 -0.333227
                                                                    -0.360203
                                                                                        -0.327501
                                                                                                             1.24889
          2
                -0.915058
                         -1.426990
                                     0.035608
                                              -0.582501
                                                        -0.537456
                                                                    -0.404885
                                                                               1.426990
                                                                                        -0.327501
                                                                                                 -0.789807
                                                                                                             1.24889
                -0.892365
                         -0.801317
                                    -0.409143
                                               0.329620
                                                        -0.921213
                                                                    -0.693085
                                                                              0.801317
                                                                                        -0.433854
                                                                                                  -0.905265
                                                                                                             1.24889
          4
                -0.182683
                          0.137194
                                    -0.544502
                                               0.760085
                                                         0.042999
                                                                    0.010435
                                                                              -0.137194
                                                                                       -0.246579
                                                                                                 -0.013096
                                                                                                            -0.80071
In [21]:
          y=df['selling price']
          x=df[df.columns.drop('selling price')]
In [22]:
In [23]:
          x.shape
```

[-0.91505816],

```
Out[23]: (19820, 17)
In [24]:
         y.shape
         (19820,)
Out[24]:
In [25]:
         x=x.to numpy()
         y=y.to_numpy()
         y=y.reshape(y.size,1)
In [26]:
         Model.fit(x,y)
In [27]:
Out[27]:
         ▼ LinearRegression
         LinearRegression()
         y hat=Model.predict(x)
In [28]:
         y hat
         array([[-1.18348944],
Out[28]:
                [-0.29402792],
                [-0.89529669],
                . . . ,
                [-0.32373367],
                [-0.86281723],
                [ 1.04516761]])
In [29]: print(Model.coef)
          [[\ 2.63965047e+10\ -1.82342529e-02\ -4.86898422e-02\ \ 3.13931704e-02\ \ ] 
            2.73611546e-02 2.63965047e+10 6.07826114e-02 7.94148445e-01
           -1.69569254e-02 -3.47542763e-03 1.40804052e-02 1.27513111e-02
            2.83169746e-03 -2.25827694e-02 -1.29337311e-02 -2.46414095e-02
           -3.26581001e-02]]
         print(Model.intercept)
In [30]:
         [3.55622928e-05]
         Model.score(x, y)
In [31]:
         0.9421886817589384
Out[31]:
In [32]:
                                                     Traceback (most recent call last)
         NameError
         Input In [32], in <cell line: 1>()
         ---> 1 pyppeteer-install
         NameError: name 'pyppeteer' is not defined
```

In [ ]: