Regression Model

```
Supervised learning has two type of model:
    1)Regression
    2)classification

1. Regression (CH-5)
Regression is continuous:
    -> LR(linear regression)
    -> Polynomial Regression

2. Classification (CH-6)
Classification is discrete:
    -> KNN
    -> Decision Tree
```

LR(linear regression)

```
simple linear regression(SLR):
    y=mx+c
    where, y=dependent variable
           x=independent variable
           m=co-efficient
           c=intercept

multiple linear regression(MLR):
```

```
Step of Supervised learning process-

step1 : Define the problem
step2 : Collect and Prepare data
step3 : Data prepocessing
step4 : Feature Engineering
step5 : Select a model
step6 : Split the data in test/train
step7 : Train the model
step8 : Validate hyper parameters
step9 : Validate the model on the basis of tesing data
step10: Iterate the model for improvement
step11: Deploy a model
```

```
# Simple linear regression with the help of excel
```

Simple linear regression model

In [5]:
```
import pandas as pd
import numpy as np

df=pd.read_csv("placement_03-06.csv")
df
```

Out[5]:

	cgpa	package
0	6.89	3.26
1	5.12	1.98
2	7.82	3.25
3	7.42	3.67
4	6.94	3.57
...	...	...
195	6.93	2.46
196	5.89	2.57
197	7.21	3.24
198	7.63	3.96
199	6.22	2.33

200 rows × 2 columns

In [6]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 2 columns):
 #   Column   Non-Null Count  Dtype  
---  ------   --------------  -----  
 0   cgpa     200 non-null    float64
 1   package  200 non-null    float64
dtypes: float64(2)
memory usage: 3.2 KB
```

In [7]:
```
from sklearn.model_selection import train_test_split
y=df['package']
y
```

Out[7]:
```
0      3.26
1      1.98
2      3.25
3      3.67
4      3.57
       ...
195    2.46
196    2.57
197    3.24
198    3.96
199    2.33
Name: package, Length: 200, dtype: float64
```

In [11]:
```
x=df.drop('package',axis=1)
x
```

Out[11]:

	cgpa
0	6.89
1	5.12
2	7.82
3	7.42
4	6.94
...	...
195	6.93
196	5.89
197	7.21
198	7.63
199	6.22

200 rows × 1 columns

In [12]:
```
type(y)
```

Out[12]:
```
pandas.core.series.Series
```

In [13]:
```
type(x)
```

Out[13]:
```
pandas.core.frame.DataFrame
```

In [14]:
```
y.shape
```

Out[14]:
```
(200,)
```

In [15]:
```
x.shape
```

Out[15]:
```
(200, 1)
```

In [49]:
```
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.8,random_state=
```

In [50]:
```
print(y_train.shape)
print(y_test.shape)
print(x_train.shape)
print(x_test.shape)
```

```
(160,)
(40,)
(160, 1)
(40, 1)
```

In [51]:
```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
model=lr.fit(x_train,y_train)
```

In [52]:
```
print("m = ",model.coef_)
```

```
m =  [0.58154877]
```

In [53]:
```
print("c =" ,model.intercept_)
```

```
c = -1.0859839580358024
```

In [54]:
```
# y=mx+c
y_pred=model.predict(x_test)
print(y_pred)
```

```
[2.9383335  4.36894346 3.18258398 1.89736121 3.49662031 3.35123312
 2.76968435 2.94996447 3.07208971 3.94441286 3.57222165 2.94996447
 2.75805338 2.64755911 3.67108494 3.2174769  3.97930579 2.90925606
 2.19395108 3.31052471 4.29915761 2.8918096  1.87409926 2.30444534
 3.62456104 2.12998071 3.9269664  2.36841571 1.5716939  2.06601035
 2.31026083 3.6885314  3.5024358  3.03719679 2.57195777 2.39167766
 3.170953   3.82228762 3.15932203 2.94414898]
```

In [55]:
```
# y_pred-y_test
diff=pd.DataFrame({"actual":y_test,
                   "predicted":y_pred})
diff
```

Out[55]:

	actual	predicted
58	3.09	2.938333
40	4.02	4.368943
34	3.42	3.182584
102	1.37	1.897361
184	3.14	3.496620
198	3.96	3.351233
95	2.79	2.769684
4	3.57	2.949964
29	3.49	3.072090
168	3.52	3.944413
171	3.76	3.572222
18	2.98	2.949964
11	2.60	2.758053
89	2.72	2.647559
110	3.76	3.671085
118	2.88	3.217477
159	4.08	3.979306
35	2.87	2.909256
136	2.10	2.193951
59	3.31	3.310525
51	3.79	4.299158
16	2.35	2.891810
44	1.86	1.874099
94	2.42	2.304445
31	3.89	3.624561
162	2.55	2.129981
38	4.36	3.926966
28	2.24	2.368416
193	1.94	1.571694
27	2.16	2.066010
47	3.26	2.310261
165	4.08	3.688531
194	3.67	3.502436
177	3.64	3.037197
176	3.23	2.571958

	actual	predicted
97	2.84	2.391678
174	2.99	3.170953
73	4.03	3.822288
69	2.94	3.159322
172	2.51	2.944149

In [56]:
```
# (mean absolute error)
# mae=1/n |y_act-y_pred|
```

In [57]:
```
from sklearn.metrics import mean_absolute_error
mae=mean_absolute_error(y_test,y_pred)
print(mae)
```

```
0.29931188593316804
```

In [58]:
```
from sklearn.metrics import mean_squared_error
mse=mean_squared_error(y_test,y_pred)
print(mse)
```

```
0.1370062519255722
```

In [59]:
```
from sklearn.metrics import r2_score
rs=r2_score(y_test,y_pred)
print(rs)
```

```
0.7283345498058083
```

Multiple linear regression model-1

In [67]:
```
import pandas as pd
df=pd.read_csv("Advertising_03-06.csv")
df
```

Out[67]:

	Unnamed: 0	TV	Radio	Newspaper	Sales
0	1	230.1	37.8	69.2	22.1
1	2	44.5	39.3	45.1	10.4
2	3	17.2	45.9	69.3	9.3
3	4	151.5	41.3	58.5	18.5
4	5	180.8	10.8	58.4	12.9
...	...	...	...	...	...
195	196	38.2	3.7	13.8	7.6
196	197	94.2	4.9	8.1	9.7
197	198	177.0	9.3	6.4	12.8
198	199	283.6	42.0	66.2	25.5
199	200	232.1	8.6	8.7	13.4

200 rows × 5 columns

In [68]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
 #   Column      Non-Null Count  Dtype  
---  ------      --------------  -----  
 0   Unnamed: 0  200 non-null    int64  
 1   TV          200 non-null    float64
 2   Radio       200 non-null    float64
 3   Newspaper   200 non-null    float64
 4   Sales       200 non-null    float64
dtypes: float64(4), int64(1)
memory usage: 7.9 KB
```

In [69]:
```
df.drop("Unnamed: 0",axis=1,inplace=True)
df
```

Out[69]:

	TV	Radio	Newspaper	Sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	9.3
3	151.5	41.3	58.5	18.5
4	180.8	10.8	58.4	12.9
...	...	...	...	...
195	38.2	3.7	13.8	7.6
196	94.2	4.9	8.1	9.7
197	177.0	9.3	6.4	12.8
198	283.6	42.0	66.2	25.5
199	232.1	8.6	8.7	13.4

200 rows × 4 columns

In [70]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 4 columns):
 #   Column     Non-Null Count  Dtype  
---  ------     --------------  -----  
 0   TV         200 non-null    float64
 1   Radio      200 non-null    float64
 2   Newspaper  200 non-null    float64
 3   Sales      200 non-null    float64
dtypes: float64(4)
memory usage: 6.4 KB
```

In [71]:
```
from sklearn.model_selection import train_test_split
y=df['Sales']
y
```

Out[71]:
```
0      22.1
1      10.4
2       9.3
3      18.5
4      12.9
       ...
195     7.6
196     9.7
197    12.8
198    25.5
199    13.4
Name: Sales, Length: 200, dtype: float64
```

In [72]:
```
x=df.drop('Sales',axis=1)
x
```

Out[72]:

	TV	Radio	Newspaper
0	230.1	37.8	69.2
1	44.5	39.3	45.1
2	17.2	45.9	69.3
3	151.5	41.3	58.5
4	180.8	10.8	58.4
...	...	...	...
195	38.2	3.7	13.8
196	94.2	4.9	8.1
197	177.0	9.3	6.4
198	283.6	42.0	66.2
199	232.1	8.6	8.7

200 rows × 3 columns

In [73]:
```
print(type(x))
print(type(y))
```

```
<class 'pandas.core.frame.DataFrame'>
<class 'pandas.core.series.Series'>
```

In [74]:
```
print(x.shape)
print(y.shape)
```

```
(200, 3)
(200,)
```

In [77]:
```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.15,random_state=
```

In [78]:
```
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(170, 3)
(30, 3)
(170,)
(30,)
```

In [80]:
```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
model=lr.fit(x_train,y_train)
print("m =" ,model.coef_)
print("c =" ,model.intercept_)
```

```
m = [ 0.04666462  0.18481678 -0.00121229]
c = 2.898009326357265
```

In [81]:
```
y_pred=model.predict(x_test)
print(y_pred)
```

```
[21.8561195  16.43070013  7.61358579 17.81414167 18.64171793 23.81414579
 16.29442126 13.26019548  9.10013819 17.24141586 14.3795469   9.89375986
 17.34765633 16.79371444 14.88188303 15.48747719 12.40242285 17.2108581
 11.28920355 18.17106497  9.35301379 12.68796292  8.76659008 10.48401019
 11.33546207 15.00377232  9.8013108  19.48893945 18.43960804 17.16086278]
```

In [83]:
```
diff=pd.DataFrame({"actual":y_test,"predicted":y_pred})
diff
```

Out[83]:

	actual	predicted
58	23.8	21.856120
40	16.6	16.430700
34	9.5	7.613586
102	14.8	17.814142
184	17.6	18.641718
198	25.5	23.814146
95	16.9	16.294421
4	12.9	13.260195
29	10.5	9.100138
168	17.1	17.241416
171	14.5	14.379547
18	11.3	9.893760
11	17.4	17.347656
89	16.7	16.793714
110	13.4	14.881883
118	15.9	15.487477
159	12.9	12.402423
35	12.8	17.210858
136	9.5	11.289204
59	18.4	18.171065
51	10.7	9.353014
16	12.5	12.687963
44	8.5	8.766590
94	11.5	10.484010
31	11.9	11.335462
162	14.9	15.003772
38	10.1	9.801311
28	18.9	19.488939
193	19.6	18.439608
27	15.9	17.160863

In [84]:
```
from sklearn.metrics import mean_absolute_error
mae=mean_absolute_error(y_test,y_pred)
print(mae)
```

```
0.984560464503767
```

In [85]:
```
from sklearn.metrics import mean_squared_error
mse=mean_squared_error(y_test,y_pred)
print(mse)
```

```
1.8945245763596135
```

In [86]:
```
from sklearn.metrics import r2_score
rs=r2_score(y_test,y_pred)
print(rs)
```

```
0.882855181551423
```

Multiple linear regression model-2

In [91]:
```
import pandas as pd
df=pd.read_csv("insurance_03-06.csv")
df
```

Out[91]:

	age	sex	bmi	children	smoker	region	expenses
0	19	female	27.9	0	yes	southwest	16884.92
1	18	male	33.8	1	no	southeast	1725.55
2	28	male	33.0	3	no	southeast	4449.46
3	33	male	22.7	0	no	northwest	21984.47
4	32	male	28.9	0	no	northwest	3866.86
...	...	...	...	...	...	...	...
1333	50	male	31.0	3	no	northwest	10600.55
1334	18	female	31.9	0	no	northeast	2205.98
1335	18	female	36.9	0	no	southeast	1629.83
1336	21	female	25.8	0	no	southwest	2007.95
1337	61	female	29.1	0	yes	northwest	29141.36

1338 rows × 7 columns

In [92]:
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1338 entries, 0 to 1337
Data columns (total 7 columns):
 #   Column    Non-Null Count  Dtype  
---  ------    --------------  -----  
 0   age       1338 non-null   int64  
 1   sex       1338 non-null   object 
 2   bmi       1338 non-null   float64
 3   children  1338 non-null   int64  
 4   smoker    1338 non-null   object 
 5   region    1338 non-null   object 
 6   expenses  1338 non-null   float64
dtypes: float64(2), int64(2), object(3)
memory usage: 73.3+ KB
```

In [93]:
```
df.drop(["sex","children","region"],axis=1,inplace=True)
df
```

Out[93]:

	age	bmi	smoker	expenses
0	19	27.9	yes	16884.92
1	18	33.8	no	1725.55
2	28	33.0	no	4449.46
3	33	22.7	no	21984.47
4	32	28.9	no	3866.86
...	...	...	...	...
1333	50	31.0	no	10600.55
1334	18	31.9	no	2205.98
1335	18	36.9	no	1629.83
1336	21	25.8	no	2007.95
1337	61	29.1	yes	29141.36

1338 rows × 4 columns

In [97]:
```
df1=pd.get_dummies(df,drop_first=True)
df1
```

Out[97]:

	age	bmi	expenses	smoker_yes
0	19	27.9	16884.92	1
1	18	33.8	1725.55	0
2	28	33.0	4449.46	0
3	33	22.7	21984.47	0
4	32	28.9	3866.86	0
...	...	...	...	...
1333	50	31.0	10600.55	0
1334	18	31.9	2205.98	0
1335	18	36.9	1629.83	0
1336	21	25.8	2007.95	0
1337	61	29.1	29141.36	1

1338 rows × 4 columns

In [103]:
```
from sklearn.model_selection import train_test_split
y=df1['expenses']
y
```

Out[103]:
```
0       16884.92
1        1725.55
2        4449.46
3       21984.47
4        3866.86
          ...
1333    10600.55
1334     2205.98
1335     1629.83
1336     2007.95
1337    29141.36
Name: expenses, Length: 1338, dtype: float64
```

In [104]:
```
x=df1.drop('expenses',axis=1)
x
```

Out[104]:

	age	bmi	smoker_yes
0	19	27.9	1
1	18	33.8	0
2	28	33.0	0
3	33	22.7	0
4	32	28.9	0
...	...	...	...
1333	50	31.0	0
1334	18	31.9	0
1335	18	36.9	0
1336	21	25.8	0
1337	61	29.1	1

1338 rows × 3 columns

In [105]:
```
print(type(x))
print(type(y))
```

```
<class 'pandas.core.frame.DataFrame'>
<class 'pandas.core.series.Series'>
```

In [106]:
```
print(x.shape)
print(y.shape)
```

```
(1338, 3)
(1338,)
```

In [110]:
```
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.85,random_state
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(1137, 3)
(201, 3)
(1137,)
(201,)
```

In [111]:
```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
model=lr.fit(x_train,y_train)
print("m =" ,model.coef_)
print("c =" ,model.intercept_)
```

```
m = [  260.25098608   313.77333855 23758.11752343]
c = -11417.098577656794
```

In [112]:
```
y_pred=model.predict(x_test)
print(y_pred)
```

```
[ 4.66662368e+03  1.33010678e+04  1.34653152e+04  1.28193039e+04
  1.04531758e+03  3.12026024e+04  1.31016999e+04  1.21308741e+04
  3.76778485e+03  3.02539216e+04  1.18854074e+04  1.72896696e+04
  8.86756871e+03  8.61286966e+03  4.06122202e+03  1.05509036e+04
  4.29383875e+03  6.38320741e+03  1.53904365e+04  1.51357374e+04
  1.25646049e+04  3.24170861e+04  9.24596826e+03  9.94182153e+03
  2.71199666e+03  8.16435467e+03  8.37108330e+03  1.15236009e+04
  7.49432680e+03  4.39165114e+03  1.43900422e+04  5.90331509e+03
  3.32255987e+04  2.72951572e+04  3.29597957e+04  1.00008958e+04
  3.12247474e+04  2.58702645e+04  1.59459962e+04  3.36224001e+04
  6.22444920e+03  1.44712615e+04  1.03183495e+04  1.55787005e+04
  4.05573278e+03  1.31718781e+04  4.86411999e+03  2.95746614e+04
  7.53499914e+03  1.24538798e+04  1.43734491e+04  1.22157738e+04
  2.26535321e+03  8.35817060e+03  2.56229262e+04  1.07797145e+04
  3.37608222e+04  1.52114048e+04  2.16198890e+03  6.75967273e+03
  7.26913354e+03  1.49696184e+04  2.72693318e+04  3.38570492e+03
  1.58075741e+04  1.12098276e+04  1.03404946e+04  1.09477678e+04
  1.24649429e+03  2.50193961e+04  3.66992504e+04  3.23635637e+04
  2.59571964e+03  1.03552788e+04  1.41316627e+04  3.45138782e+04
  3.16606358e+03  4.88445616e+03  1.10954221e+04  9.81631220e+03
 -6.80435782e+02  1.34081125e+04  1.01577825e+04  4.02248391e+03
  3.30428239e+04  3.31239805e+04  7.45190830e+03  3.75667119e+04
  1.17838519e+04  9.63908936e+03  3.02354569e+04  3.25611227e+04
  1.45838581e+04  1.11784503e+04  6.18947609e+02  1.16325172e+04
  9.95105385e+03  1.49234568e+04  1.51301855e+04  5.06535939e+03
  1.40855011e+04  2.64092311e+04  2.78046180e+04  2.82050999e+04
  3.62138688e+04  2.72065771e+04  1.27787162e+03  9.77015062e+03
  4.97116470e+03  1.22471511e+04  6.12295643e+03  4.73124988e+03
  1.43475829e+03  1.79375524e+04  3.23805056e+03  2.53296497e+03
  1.04751735e+04  1.28544243e+04  1.01484875e+04  3.79729064e+03
  9.62804819e+03  1.22452796e+04  8.10528039e+03  7.55159223e+03
  3.68930663e+04  1.18373743e+04  1.08665485e+04  2.91058729e+04
  3.59480032e+04  1.16749357e+04  2.90116782e+04  8.55329335e+01
  7.19533771e+03  3.21623870e+04  9.21459093e+03 -8.59906805e+00
  1.56026762e+03  5.28493804e+03  7.85800479e+03  1.23117147e+04
  1.46650774e+04  8.45411145e+03  2.91280179e+04  1.65348047e+04
  1.36554508e+04  1.15568498e+04  2.49235533e+03  9.48220269e+03
  4.31224069e+03  5.55448403e+03  1.19887090e+04  5.04502322e+03
  1.38824529e+04  1.30629618e+04  1.36886370e+04  7.95394564e+03
  1.19905805e+04  1.05472232e+04  9.67233823e+03  5.36247695e+03
  6.67851612e+03  4.06066956e+04  1.37182054e+04  4.76262722e+03
  7.32820782e+03  5.62640832e+03  3.27548760e+04  1.17986362e+04
  1.16915915e+04  6.64345840e+03  6.26137847e+03  6.78924121e+03
  3.31905409e+04  3.51322552e+04  2.35580483e+03  7.03845104e+03
  5.61349562e+03  1.41039658e+04  1.29820779e+03  1.12780715e+04
  1.34856514e+04  1.13612250e+04  1.07170226e+04  1.24722817e+04
  2.14907619e+03  2.85299771e+04  3.12545393e+03  1.50286300e+04
  6.79853622e+03  8.54088267e+03  1.45930904e+04  3.89897958e+04
  3.05165810e+03  1.21511695e+03  4.59101900e+03  8.31943249e+03
  7.05136374e+03  4.57249167e+03  9.51177117e+03  9.37147760e+03
  9.98424003e+03]
```

In [113]:
```
diff=pd.DataFrame({"actual":y_test,"predicated":y_pred})
diff
```

Out[113]:

	actual	predicated
559	1646.43	4666.623676
1087	11353.23	13301.067793
1020	8798.59	13465.315239
460	10381.48	12819.303931
802	2103.08	1045.317580
...	...	...
891	7243.81	7051.363739
414	2134.90	4572.491675
258	11520.10	9511.771173
538	8233.10	9371.477595
929	6289.75	9984.240030

201 rows × 2 columns

In [114]:
```
from sklearn.metrics import mean_absolute_error
mae=mean_absolute_error(y_test,y_pred)
print(mae)
```

```
4019.341739524793
```

In [115]:
```
from sklearn.metrics import mean_squared_error
mse=mean_squared_error(y_test,y_pred)
print(mse)
```

```
36946790.51390431
```

In [116]:
```
from sklearn.metrics import r2_score
rs=r2_score(y_test,y_pred)
print(rs)
```

```
0.739443551121639
```

Model-3

In [1]:
```
import pandas as pd
car=pd.read_csv("car data_03-06.csv")
car
```

Out[1]:

	Car_Name	Year	Selling_Price	Present_Price	Driven_kms	Fuel_Type	Selling_type	Transmi
0	ritz	2014	3.35	5.59	27000	Petrol	Dealer	M
1	sx4	2013	4.75	9.54	43000	Diesel	Dealer	M
2	ciaz	2017	7.25	9.85	6900	Petrol	Dealer	M
3	wagon r	2011	2.85	4.15	5200	Petrol	Dealer	M
4	swift	2014	4.60	6.87	42450	Diesel	Dealer	M
...	...	...	...	...	...	...	...	
296	city	2016	9.50	11.60	33988	Diesel	Dealer	M
297	brio	2015	4.00	5.90	60000	Petrol	Dealer	M
298	city	2009	3.35	11.00	87934	Petrol	Dealer	M
299	city	2017	11.50	12.50	9000	Diesel	Dealer	M
300	brio	2016	5.30	5.90	5464	Petrol	Dealer	M

301 rows × 9 columns

In [2]:
```
car.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 301 entries, 0 to 300
Data columns (total 9 columns):
 #   Column         Non-Null Count  Dtype  
---  ------         --------------  -----  
 0   Car_Name       301 non-null    object 
 1   Year           301 non-null    int64  
 2   Selling_Price  301 non-null    float64
 3   Present_Price  301 non-null    float64
 4   Driven_kms     301 non-null    int64  
 5   Fuel_Type      301 non-null    object 
 6   Selling_type   301 non-null    object 
 7   Transmission   301 non-null    object 
 8   Owner          301 non-null    int64  
dtypes: float64(2), int64(3), object(4)
memory usage: 21.3+ KB
```

In [3]:
```
car.drop("Car_Name",axis=1,inplace=True)
car
```

Out[3]:

	Year	Selling_Price	Present_Price	Driven_kms	Fuel_Type	Selling_type	Transmission	Own
0	2014	3.35	5.59	27000	Petrol	Dealer	Manual	
1	2013	4.75	9.54	43000	Diesel	Dealer	Manual	
2	2017	7.25	9.85	6900	Petrol	Dealer	Manual	
3	2011	2.85	4.15	5200	Petrol	Dealer	Manual	
4	2014	4.60	6.87	42450	Diesel	Dealer	Manual	
...	...	...	...	...	...	...	...	...
296	2016	9.50	11.60	33988	Diesel	Dealer	Manual	
297	2015	4.00	5.90	60000	Petrol	Dealer	Manual	
298	2009	3.35	11.00	87934	Petrol	Dealer	Manual	
299	2017	11.50	12.50	9000	Diesel	Dealer	Manual	
300	2016	5.30	5.90	5464	Petrol	Dealer	Manual	

301 rows × 8 columns

In [4]:
```
car['Age']=2024-car['Year']
car
```

Out[4]:

	Year	Selling_Price	Present_Price	Driven_kms	Fuel_Type	Selling_type	Transmission	Own
0	2014	3.35	5.59	27000	Petrol	Dealer	Manual	
1	2013	4.75	9.54	43000	Diesel	Dealer	Manual	
2	2017	7.25	9.85	6900	Petrol	Dealer	Manual	
3	2011	2.85	4.15	5200	Petrol	Dealer	Manual	
4	2014	4.60	6.87	42450	Diesel	Dealer	Manual	
...	...	...	...	...	...	...	...	...
296	2016	9.50	11.60	33988	Diesel	Dealer	Manual	
297	2015	4.00	5.90	60000	Petrol	Dealer	Manual	
298	2009	3.35	11.00	87934	Petrol	Dealer	Manual	
299	2017	11.50	12.50	9000	Diesel	Dealer	Manual	
300	2016	5.30	5.90	5464	Petrol	Dealer	Manual	

301 rows × 9 columns

In [5]:
```
car.drop("Year",axis=1,inplace=True)
car
```

Out[5]:

	Selling_Price	Present_Price	Driven_kms	Fuel_Type	Selling_type	Transmission	Owner	Ag
0	3.35	5.59	27000	Petrol	Dealer	Manual	0	1
1	4.75	9.54	43000	Diesel	Dealer	Manual	0	1
2	7.25	9.85	6900	Petrol	Dealer	Manual	0	
3	2.85	4.15	5200	Petrol	Dealer	Manual	0	1
4	4.60	6.87	42450	Diesel	Dealer	Manual	0	1
...	...	...	...	...	...	...	...	.
296	9.50	11.60	33988	Diesel	Dealer	Manual	0	
297	4.00	5.90	60000	Petrol	Dealer	Manual	0	
298	3.35	11.00	87934	Petrol	Dealer	Manual	0	1
299	11.50	12.50	9000	Diesel	Dealer	Manual	0	
300	5.30	5.90	5464	Petrol	Dealer	Manual	0	

301 rows × 8 columns

In [8]:
```
car1=pd.get_dummies(car,drop_first=True)
car1
```

Out[8]:

	Selling_Price	Present_Price	Driven_kms	Owner	Age	Fuel_Type_Diesel	Fuel_Type_Petrol
0	3.35	5.59	27000	0	10	0	1
1	4.75	9.54	43000	0	11	1	0
2	7.25	9.85	6900	0	7	0	1
3	2.85	4.15	5200	0	13	0	1
4	4.60	6.87	42450	0	10	1	0
...	...	...	...	...	...	...	...
296	9.50	11.60	33988	0	8	1	0
297	4.00	5.90	60000	0	9	0	1
298	3.35	11.00	87934	0	15	0	1
299	11.50	12.50	9000	0	7	1	0
300	5.30	5.90	5464	0	8	0	1

301 rows × 9 columns

In [9]:
```
from sklearn.model_selection import train_test_split
y=car1['Selling_Price']
y
```

Out[9]:
```
0       3.35
1       4.75
2       7.25
3       2.85
4       4.60
       ...
296     9.50
297     4.00
298     3.35
299    11.50
300     5.30
Name: Selling_Price, Length: 301, dtype: float64
```

In [10]:
```
x=car1.drop("Selling_Price",axis=1)
x
```

Out[10]:

	Present_Price	Driven_kms	Owner	Age	Fuel_Type_Diesel	Fuel_Type_Petrol	Selling_type_In
0	5.59	27000	0	10	0	1	
1	9.54	43000	0	11	1	0	
2	9.85	6900	0	7	0	1	
3	4.15	5200	0	13	0	1	
4	6.87	42450	0	10	1	0	
...	...	...	...	...	...	...	
296	11.60	33988	0	8	1	0	
297	5.90	60000	0	9	0	1	
298	11.00	87934	0	15	0	1	
299	12.50	9000	0	7	1	0	
300	5.90	5464	0	8	0	1	

301 rows × 8 columns

In [11]:
```
print(type(x))
print(type(y))
```

```
<class 'pandas.core.frame.DataFrame'>
<class 'pandas.core.series.Series'>
```

In [12]:
```
print(x.shape)
print(y.shape)
```

```
(301, 8)
(301,)
```

In [13]:
```
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.85,random_state
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(255, 8)
(46, 8)
(255,)
(46,)
```

In [14]:
```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
model=lr.fit(x_train,y_train)
print("m =" ,model.coef_)
print("c =" ,model.intercept_)
```

```
m = [ 4.35302568e-01 -5.74656638e-06  3.09388161e-01 -3.92843790e-01
  2.25837305e+00  5.10872356e-01 -1.21081820e+00 -1.83023303e+00]
c = 6.805277248730253
```

In [15]:
```
y_pred=model.predict(x_test)
y_pred
```

Out[15]:
```
array([ 7.78693357,  2.99462971, -0.50679037,  4.18535814,  0.52057204,
        5.76800596,  1.91861631,  2.61742261,  7.67505414,  0.9729874 ,
        8.07083232,  3.5281268 ,  4.88005217,  4.60002113, -2.04896272,
        3.07278595,  7.92318893,  6.74538423,  6.87869784,  7.95766833,
        4.25721114,  4.07722942, 11.29456886,  8.02126452,  9.47602518,
        3.45113708,  3.87626941,  1.05764871, -0.56181678, -0.54720019,
        0.03389518, -1.19234707,  4.26949393, 20.55308353, 18.62280514,
        4.27026138,  3.52636571,  1.64807387, -0.02518064,  5.76668222,
        7.97846986,  9.80428476,  0.40823392,  6.05156627,  5.83961851,
        4.32779026])
```

In [17]:
```
diff=pd.DataFrame({"actual":y_test,"predicated":y_pred})
diff
```

Out[17]:

	actual	predicated
285	7.40	7.786934
248	4.00	2.994630
150	0.50	-0.506790
217	3.15	4.185358
107	1.25	0.520572
206	5.75	5.768006
132	0.75	1.918616
73	2.65	2.617423
288	8.40	7.675054
157	0.48	0.972987
267	8.35	8.070832
88	3.45	3.528127
300	5.30	4.880052
58	4.10	4.600021
192	0.20	-2.048963
177	0.35	3.072786
11	6.85	7.923189
230	6.15	6.745384
224	5.11	6.878698
29	7.45	7.957668
27	6.00	4.257211
293	3.25	4.077229
78	5.25	11.294569
12	7.50	8.021265
85	2.50	9.476025
18	3.25	3.451137
298	3.35	3.876269
139	0.60	1.057649
180	0.30	-0.561817
176	0.35	-0.547200
182	0.30	0.033895
197	0.16	-1.192347
202	4.40	4.269494
59	19.99	20.553084
51	23.00	18.622805

	actual	predicated
89	4.75	4.270261
246	3.75	3.526366
120	1.05	1.648074
191	0.20	-0.025181
221	4.50	5.766682
256	10.25	7.978470
250	12.90	9.804285
193	0.20	0.408234
4	4.60	6.051566
70	3.95	5.839619
294	3.75	4.327790

In [18]:
```
from sklearn.metrics import mean_absolute_error
mae=mean_absolute_error(y_test,y_pred)
print(mae)
```

```
1.1919111738890504
```

In [19]:
```
from sklearn.metrics import mean_squared_error
mse=mean_squared_error(y_test,y_pred)
print(mse)
```

```
3.468310758909892
```

In [20]:
```
from sklearn.metrics import r2_score
rs=r2_score(y_test,y_pred)
print(rs)
```

```
0.8404169753392161
```

In [21]:
```
x.columns
```

Out[21]:
```
Index(['Present_Price', 'Driven_kms', 'Owner', 'Age', 'Fuel_Type_Diesel',
       'Fuel_Type_Petrol', 'Selling_type_Individual', 'Transmission_Manual'],
      dtype='object')
```

In [27]:
```
y_pred=model.predict([[425000,100000,0,8,0,1,0,0]])
print(y_pred)
```

```
[185007.1902717]
```

In [29]:
```
y_pred=model.predict([[80000,425000,0,8,0,1,0,0]])
print(y_pred)
```

```
[34825.93657286]
```

MODEL-4

In [31]:
```
import pandas as pd
import numpy as np
olm=pd.read_csv("olympic100m.csv")
olm
```

Out[31]:

	year	time
0	1896	12.00
1	1900	11.00
2	1904	11.00
3	1906	11.20
4	1908	10.80
5	1912	10.80
6	1920	10.80
7	1924	10.60
8	1928	10.80
9	1932	10.30
10	1936	10.30
11	1948	10.30
12	1952	10.40
13	1956	10.50
14	1960	10.20
15	1964	10.00
16	1968	9.95
17	1972	10.14
18	1976	10.06
19	1980	10.25
20	1984	9.99
21	1988	9.92
22	1992	9.96
23	1996	9.84
24	2000	9.87
25	2004	9.85
26	2008	9.69
27	2012	9.63
28	2016	9.81

In [32]:
```
olm.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 29 entries, 0 to 28
Data columns (total 2 columns):
 #   Column  Non-Null Count  Dtype  
---  ------  --------------  -----  
 0   year    29 non-null     int64  
 1   time    29 non-null     float64
dtypes: float64(1), int64(1)
memory usage: 592.0 bytes
```

In [33]:
```
from sklearn.model_selection import train_test_split
y=olm['time']
y
```

Out[33]:
```
0     12.00
1     11.00
2     11.00
3     11.20
4     10.80
5     10.80
6     10.80
7     10.60
8     10.80
9     10.30
10    10.30
11    10.30
12    10.40
13    10.50
14    10.20
15    10.00
16     9.95
17    10.14
18    10.06
19    10.25
20     9.99
21     9.92
22     9.96
23     9.84
24     9.87
25     9.85
26     9.69
27     9.63
28     9.81
Name: time, dtype: float64
```

In [37]:
```
x=olm[["year"]]
x
```

Out[37]:

	year
0	1896
1	1900
2	1904
3	1906
4	1908
5	1912
6	1920
7	1924
8	1928
9	1932
10	1936
11	1948
12	1952
13	1956
14	1960
15	1964
16	1968
17	1972
18	1976
19	1980
20	1984
21	1988
22	1992
23	1996
24	2000
25	2004
26	2008
27	2012
28	2016

In [38]:
```
print(type(x))
print(type(y))
```

```
<class 'pandas.core.frame.DataFrame'>
<class 'pandas.core.series.Series'>
```

In [39]:
```
print(x.shape)
print(y.shape)
```

```
(29, 1)
(29,)
```

In [40]:
```
x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.85,random_state
print(x_train.shape)
print(x_test.shape)
print(y_train.shape)
print(y_test.shape)
```

```
(24, 1)
(5, 1)
(24,)
(5,)
```

In [41]:
```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
model=lr.fit(x_train,y_train)
print("m = " ,model.coef_)
print("c = " ,model.intercept_)
```

```
m =  [-0.0129917]
c =  35.764216687315766
```

In [42]:
```
y_pred=model.predict([[2024]])
print(y_pred)
```

```
[9.46902066]
```

In [ ]:
```

```