```
In [1]:
          import pandas as pd
          import numpy as np
          import seaborn as sns
          import matplotlib.pyplot as plt
In [2]:
          cars=pd.read csv("C:/Users/USER/Desktop/Datasets/Cars.csv")
          cars
                      MPG VOL
                                        SP
                                                  WT
Out[2]:
             HP
                                104.185353 28.762059
          0
              49
                  53.700681
                             89
          1
              55
                  50.013401
                             92 105.461264 30.466833
          2
              55 50.013401
                                105.461264 30.193597
          3
              70 45.696322
                                 113.461264 30.632114
          4
              53 50.504232
                             92
                                 104.461264 29.889149
                                         ...
         76
             322 36.900000
                             50
                                 169.598513 16.132947
         77
             238 19.197888
                            115
                                150.576579 37.923113
         78
             263 34.000000
                             50
                                 151.598513 15.769625
             295 19.833733
         79
                            119
                                 167.944460 39.423099
             236 12.101263
                            107 139.840817 34.948615
         80
        81 rows × 5 columns
In [3]:
          cols =['HP','VOL','SP','WT']
          x=cars[cols]
          y=cars.MPG
In [4]:
Out[4]:
             HP VOL
                              SP
                                        WT
          0
              49
                   89 104.185353 28.762059
          1
              55
                       105.461264 30.466833
                   92
          2
              55
                      105.461264 30.193597
                   92
          3
              70
                       113.461264 30.632114
          4
              53
                   92
                       104.461264 29.889149
         76
             322
                   50 169.598513 16.132947
         77 238
                  115 150.576579 37.923113
```

	HP	VOL	SP	WT
78	263	50	151.598513	15.769625
79	295	119	167.944460	39.423099
80	236	107	139.840817	34.948615

81 rows × 4 columns

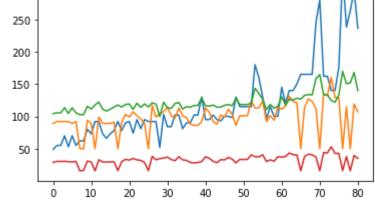
```
In [5]:
                53.700681
Out[5]:
                50.013401
         1
         2
                50.013401
         3
                45.696322
         4
                50.504232
         76
                36.900000
         77
                19.197888
         78
                34.000000
         79
                19.833733
         80
                12.101263
         Name: MPG, Length: 81, dtype: float64
In [6]:
          x.describe()
Out[6]:
                       HP
                                  VOL
                                              SP
                                                        WT
                 81.000000
                             81.000000
                                        81.000000
         count
                                                  81.000000
                117.469136
                             98.765432
                                       121.540272 32.412577
         mean
                 57.113502
                             22.301497
                                        14.181432
           std
                                                   7.492813
                 49.000000
                             50.000000
                                        99.564907
                                                  15.712859
           min
          25%
                 84.000000
                             89.000000
                                       113.829145
                                                  29.591768
          50%
                100.000000
                            101.000000
                                       118.208698
                                                  32.734518
          75%
                140.000000
                           113.000000
                                       126.404312 37.392524
          max 322.000000 160.000000 169.598513 52.997752
In [7]:
          x.median()
         HP
                 100.000000
Out[7]:
         VOL
                 101.000000
                 118.208698
         SP
                  32.734518
         dtype: float64
In [8]:
          x.mode()
Out[8]:
              HP VOL
                                SP
                                          WT
```

	HP	VOL	SP	WT
0	92.0	50.0	118.288996	15.712859
1	NaN	NaN	NaN	15.753535
2	NaN	NaN	NaN	15.769625
3	NaN	NaN	NaN	15.823060
4	NaN	NaN	NaN	15.847758
•••	•••			
76	NaN	NaN	NaN	42.778219
77	NaN	NaN	NaN	43.353123
78	NaN	NaN	NaN	43.390988
79	NaN	NaN	NaN	44.013139
80	NaN	NaN	NaN	52.997752

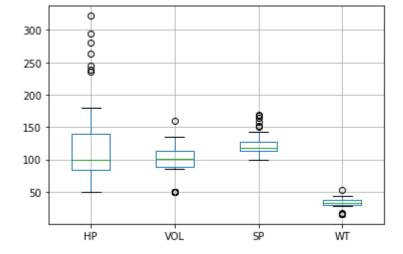
81 rows × 4 columns

```
In [9]:
           x.var()
                  3261.952160
          {\sf HP}
 Out[9]:
          VOL
                   497.356790
          SP
                   201.113002
          WT
                    56.142247
          dtype: float64
In [10]:
           x.std()
                  57.113502
          ΗP
Out[10]:
          VOL
                  22.301497
          SP
                  14.181432
          WT
                   7.492813
          dtype: float64
In [11]:
           x.skew()
                  1.716216
          ΗP
Out[11]:
          VOL
                 -0.590197
          SP
                  1.611450
          WT
                 -0.614753
          dtype: float64
In [12]:
           x.kurtosis()
          {\sf HP}
                  2.960025
Out[12]:
          VOL
                  0.920229
          SP
                  2.977329
                  0.950291
          WT
          dtype: float64
```

```
Multiple Linear Regression
           x.isnull().sum()
In [13]:
                 0
Out[13]:
          VOL
                 0
          SP
                 0
          WT
                  0
          dtype: int64
In [14]:
           plt.plot(x)
          [<matplotlib.lines.Line2D at 0x206d8e07190>,
Out[14]:
           <matplotlib.lines.Line2D at 0x206d8e071c0>,
           <matplotlib.lines.Line2D at 0x206d8e072e0>,
           <matplotlib.lines.Line2D at 0x206d8e07400>]
          300
          250
          200
```





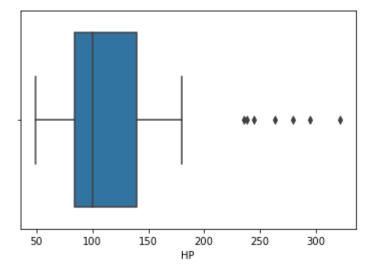


```
In [16]:
          sns.boxplot(x["HP"])
```

C:\Users\USER\anaconda3\lib\site-packages\seaborn\\_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will re sult in an error or misinterpretation.

warnings.warn(

<AxesSubplot:xlabel='HP'> Out[16]:



```
from feature_engine.outliers import Winsorizer
win=Winsorizer(capping_method="iqr", tail="both", fold=1.5, variables=["HP"])
New_HP=win.fit_transform(x[["HP"]])
```

```
In [18]: x.insert(loc=1,column='New_HP', value=New_HP)
```

```
In [19]: del x["HP"] x
```

0

ut[19]:		New_HP	VOL	SP	WT
	0	49.0	89	104.185353	28.762059
	1	55.0	92	105.461264	30.466833
	2	55.0	92	105.461264	30.193597
	3	70.0	92	113.461264	30.632114
	4	53.0	92	104.461264	29.889149
	•••		•••		
	76	224.0	50	169.598513	16.132947
	77	224.0	115	150.576579	37.923113
	78	224.0	50	151.598513	15.769625
	79	224.0	119	167.944460	39.423099
	80	224.0	107	139.840817	34.948615

81 rows × 4 columns

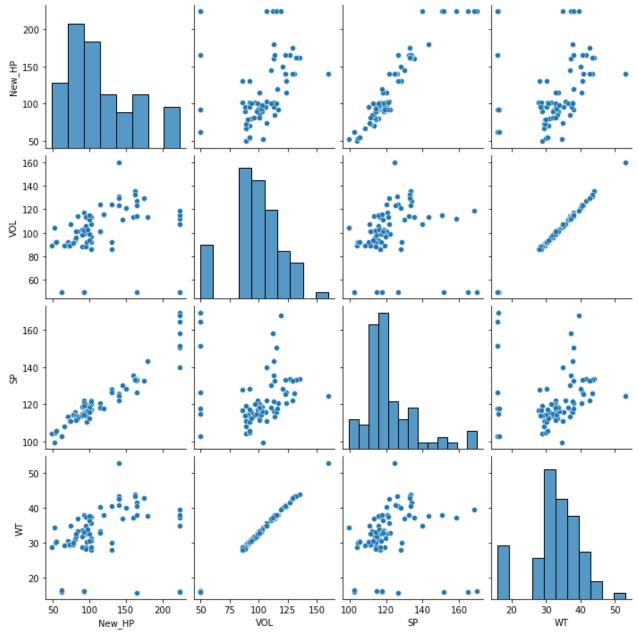
## boxplot=x.boxplot()

```
In [20]: x.corr()
Out[20]: New_HP VOL SP WT
```

	New_HP	VOL	SP	WT
New_HP	1.000000	0.183756	0.948771	0.182321
VOL	0.183756	1.000000	0.102170	0.999203
SP	0.948771	0.102170	1.000000	0.102439
WT	0.182321	0.999203	0.102439	1.000000

In [21]: sns.pairplot(x)

Out[21]: <seaborn.axisgrid.PairGrid at 0x206d902b2b0>



In [22]: names=x.columns names

Out[22]: Index(['New\_HP', 'VOL', 'SP', 'WT'], dtype='object')

```
In [25]:
          from sklearn.preprocessing import MinMaxScaler
          scale=MinMaxScaler()
          S_x=scale.fit(x)
          S_x=scale.transform(x)
          S_x=pd.DataFrame(S_x, columns=names)
          S x
Out[25]:
             New_HP
                         VOL
                                   SP
                                           WT
          0 0.000000 0.354545 0.065975 0.349986
             0.034286  0.381818  0.084193  0.395709
             0.120000 0.381818 0.198424 0.400142
             0.022857  0.381818  0.069914  0.380215
             1.000000 0.000000
                             1.000000 0.011267
             1.000000 0.590909 0.728388 0.595690
         78
             1.000000 0.000000 0.742981 0.001523
             1.000000 0.627273 0.976382 0.635921
            1.000000 0.518182 0.575094 0.515913
        81 rows × 4 columns
In [26]:
          from sklearn.model_selection import train_test_split
          x train, x test, y train, y test = train test split(S x,y, test size=0.2)
In [27]:
          x train
Out[27]:
             New_HP
                         VOL
                                   SP
                                           WT
         28 0.017143 0.490909 0.000000 0.503430
         67 0.662857 0.700000 0.487128 0.693608
             0.645714  0.772727  0.483355  0.759028
             0.634286  0.572727  0.511579  0.577755
             0.022857  0.381818  0.069914  0.380215
             1.000000 0.000000 0.742981 0.001523
         78
             0.280000 0.481818 0.241427 0.488058
```

```
        New_HP
        VOL
        SP
        WT

        25
        0.245714
        0.000000
        0.214663
        0.008859

        22
        0.262857
        0.463636
        0.295916
        0.454956
```

64 rows × 4 columns

```
In [28]:
          from sklearn.linear model import LinearRegression
          rgsr=LinearRegression()
          rgsr.fit(x_train, y_train)
          y_pred=rgsr.predict(x_test)
          y_pred
         array([22.87126169, 39.44267825, 18.15228001, 14.20081617, 38.48251104,
Out[28]:
                 38.9758946 , 36.45555826, 43.91231198, 21.74910438, 27.66116864,
                 17.77210199, 41.7611509, 45.63690802, 37.84592873, 39.11881947,
                 35.79710839, 32.99787181])
In [29]:
          from sklearn.metrics import mean_squared_error
In [30]:
          RMSE=mean_squared_error(y_test, y_pred)
In [31]:
          RMSE
         8.91223610053179
Out[31]:
```