import pandas as pd
import numpy as np
from sklearn import linear_model
import matplotlib.pyplot as plt

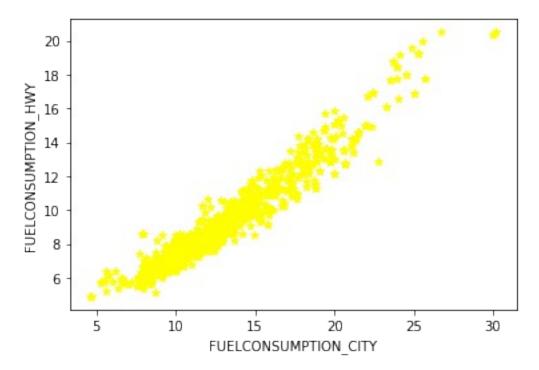
#Read the Database file

df = pd.read_csv('FuelConsumption.csv')
df

| | | MAKE | MODEL | VEHICLECLASS | ENGINESIZE |
|---|-----------|-------|------------|----------------|------------|
| CYLINDERS 0 4 1 4 2 4 3 6 4 6 | \ 2014 | ACURA | ILX | COMPACT | 2.0 |
| | 2014 | ACURA | ILX | COMPACT | 2.4 |
| | 2014 | ACURA | ILX HYBRID | COMPACT | 1.5 |
| | 2014 | ACURA | MDX 4WD | SUV - SMALL | 3.5 |
| | 2014 | ACURA | RDX AWD | SUV - SMALL | 3.5 |
| | | | | | |
| 1062 6 | 2014 | V0LV0 | XC60 AWD | SUV - SMALL | 3.0 |
| 1063 6 | 2014 | V0LV0 | XC60 AWD | SUV - SMALL | 3.2 |
| 1064 6 | 2014 | V0LV0 | XC70 AWD | SUV - SMALL | 3.0 |
| 1065 6 | 2014 | V0LV0 | XC70 AWD | SUV - SMALL | 3.2 |
| 1066 6 | 2014 | V0LV0 | XC90 AWD | SUV - STANDARD | 3.2 |

| | TRANSMISSION | FUELTYPE | FUELCONSUMPTION_CITY | FUELCONSUMPTION_HWY |
|------|--------------|----------|----------------------|---------------------|
| 0 | AS5 | Z | 9.9 | 6.7 |
| 1 | M6 | Z | 11.2 | 7.7 |
| 2 | AV7 | Z | 6.0 | 5.8 |
| 3 | AS6 | Z | 12.7 | 9.1 |
| 4 | AS6 | Z | 12.1 | 8.7 |
| | | | | • • • |
| 1062 | AS6 | Χ | 13.4 | 9.8 |

```
1063
               AS6
                          Χ
                                               13.2
                                                                       9.5
1064
               AS6
                           Χ
                                               13.4
                                                                       9.8
1065
               AS6
                           Χ
                                               12.9
                                                                       9.3
                           Χ
                                               14.9
1066
               AS6
                                                                      10.2
      FUELCONSUMPTION COMB
                              FUELCONSUMPTION COMB MPG CO2EMISSIONS
0
                        8.5
                                                      33
                                                                    196
                        9.6
                                                      29
1
                                                                    221
2
                        5.9
                                                      48
                                                                    136
3
                       11.1
                                                      25
                                                                    255
4
                       10.6
                                                      27
                                                                    244
                        . . .
                                                     . . .
                                                                    . . .
1062
                       11.8
                                                      24
                                                                    271
                                                      25
1063
                       11.5
                                                                    264
1064
                       11.8
                                                      24
                                                                    271
                       11.3
1065
                                                      25
                                                                    260
1066
                       12.8
                                                      22
                                                                    294
[1067 rows x 13 columns]
%matplotlib inline
plt.xlabel('FUELCONSUMPTION CITY')
plt.ylabel('FUELCONSUMPTION HWY')
plt.scatter(df.FUELCONSUMPTION_CITY,df.FUELCONSUMPTION_HWY,color='yell
ow', marker='*');
```



```
new_df=df
new_df.drop(df.iloc[:,0:5], inplace = True, axis = 1)
new_df
```

| | FUELCONSUMPTION COMB MPG | CO2EMISSIONS |
|------|--------------------------|--------------|
| 0 | 33 | 196 |
| 1 | 29 | 221 |
| 2 | 48 | 136 |
| 3 | 25 | 255 |
| 4 | 27 | 244 |
| | | |
| 1062 | 24 | 271 |
| 1063 | 25 | 264 |
| 1064 | 24 | 271 |
| 1065 | 25 | 260 |
| 1066 | 22 | 294 |

[1067 rows x 2 columns]

new_df=new_df.drop('CO2EMISSIONS',axis='columns')
new_df

| | FUELCONSUMPTION COMB MPG |
|---|--------------------------|
| 0 | 33 |
| 1 | 29 |
| 2 | 48 |
| 3 | 25 |
| 4 | 27 |
| | |

```
1062
                              24
1063
                              25
1064
                              24
1065
                              25
1066
                              22
[1067 rows x 1 columns]
JanTemp = df.C02EMISSIONS
JanTemp
0
        196
1
        221
2
        136
3
        255
4
        244
1062
        271
1063
        264
1064
        271
1065
        260
1066
        294
Name: CO2EMISSIONS, Length: 1067, dtype: int64
print(new_df)
      FUELCONSUMPTION COMB MPG
0
                              33
                              29
1
2
                              48
3
                              25
4
                              27
1062
                              24
1063
                              25
1064
                              24
1065
                              25
1066
                              22
[1067 rows x 1 columns]
#create linear regression object
reg = linear model.LinearRegression()
reg.fit(new_df,JanTemp)
LinearRegression()
#**(2) Predict FOR FC 26
reg.predict([[26]])
array([259.62358637])
```

```
reg.coef
array([-7.6907997])
reg.intercept
459.5843786973706
#Y = m * X+b (m is coefficient and b is intercept)
-7.6907997*26+459.5843786973706
259.6235864973706
#**(2) Predict FOR FC 60
reg.predict([[60]])
array([-1.86360359])
X=new df
Y=JanTemp
from sklearn.model selection import train test split
X train,X test,Y train,Y test =
train_test_split(X,Y,test_size=0.25,random_state=2)
from sklearn.linear model import LinearRegression
lr = LinearRegression()
lr.fit(X train,Y train)
Y pred = lr.predict(X test)
Y pred
array([289.28394791, 281.71206333, 289.28394791, 251.42452502,
       168.13379468, 160.5619101 , 304.42771707, 327.1433708 ,
       311.99960164, 266.56829418, 251.42452502, 205.99321756,
       296.85583249, 266.56829418, 311.99960164, 266.56829418,
       334.71525537, 205.99321756, 213.56510214, 342.28713995,
       221.13698672, 152.99002552, 289.28394791, 168.13379468,
       243.85264045, 327.1433708 , 266.56829418, 266.56829418,
       251.42452502, 236.28075587, 266.56829418, 274.14017876,
       266.56829418, 296.85583249, 357.43090911, 251.42452502,
       190.84944841, 213.56510214, 221.13698672, 327.1433708 ,
       213.56510214, 296.85583249, 198.42133298, 251.42452502,
       266.56829418, 289.28394791, 311.99960164, 289.28394791,
       327.1433708 , 296.85583249 , 327.1433708 , 266.56829418 ,
       289.28394791, 205.99321756, 251.42452502, 281.71206333,
       251.42452502, 183.27756383, 175.70567925, 274.14017876,
       289.28394791, 281.71206333, 281.71206333, 107.55871806,
       183.27756383, 62.12741059, 304.42771707, 304.42771707,
       198.42133298, 251.42452502, 327.1433708 , 274.14017876,
       296.85583249, 342.28713995, 251.42452502, 198.42133298,
       160.5619101 , 327.1433708 , 274.14017876, 243.85264045,
       274.14017876, 160.5619101 , 221.13698672, 175.70567925,
       349.85902453, 99.98683348, 152.99002552, 258.9964096,
```

```
228.70887129, 334.71525537, 160.5619101, 296.85583249,
319.57148622, 334.71525537, 205.99321756, 311.99960164,
228.70887129, 213.56510214, 251.42452502, 228.70887129,
251.42452502, 190.84944841, 334.71525537, 236.28075587,
296.85583249, 334.71525537,
                             69.69929517, 281.71206333,
205.99321756, 221.13698672, 243.85264045, 266.56829418,
289.28394791, 289.28394791, 304.42771707, 349.85902453,
327.1433708 , 281.71206333 , 281.71206333 , 243.85264045 ,
289.28394791, 281.71206333, 251.42452502, 327.1433708
198.42133298, 365.00279368, 266.56829418, 319.57148622,
115.13060263, 281.71206333, 289.28394791, 334.71525537,
213.56510214, 221.13698672, 258.9964096, 304.42771707,
221.13698672, 198.42133298, 258.9964096 , 236.28075587,
319.57148622, 251.42452502, 205.99321756, 221.13698672,
221.13698672, 266.56829418, 198.42133298, 243.85264045,
281.71206333, 319.57148622, 243.85264045, 228.70887129,
243.85264045, 311.99960164, 198.42133298, 281.71206333,
213.56510214, 251.42452502, 327.1433708 , 243.85264045,
190.84944841, 205.99321756, 183.27756383, 311.99960164,
221.13698672, 183.27756383, 311.99960164, 236.28075587,
258.9964096 , 243.85264045, 274.14017876, 175.70567925,
289.28394791, 319.57148622, 205.99321756, 228.70887129,
122.70248721, 221.13698672, 145.41814094, 221.13698672,
334.71525537, 221.13698672, 281.71206333, 304.42771707,
319.57148622, 160.5619101 , 190.84944841, 205.99321756,
258.9964096 , 243.85264045, 160.5619101 , 281.71206333,
334.71525537, 183.27756383, 289.28394791, 198.42133298,
183.27756383, 258.9964096, 296.85583249, 289.28394791,
258.9964096 , 304.42771707, 221.13698672, 334.71525537,
243.85264045, 304.42771707, 168.13379468, 228.70887129,
251.42452502, 243.85264045, 258.9964096 , 266.56829418,
274.14017876, 357.43090911, 334.71525537, 122.70248721,
304.42771707, 198.42133298, 342.28713995, 311.99960164,
296.85583249, 251.42452502, 289.28394791, 266.56829418,
251.42452502, 183.27756383, 319.57148622, 221.13698672,
289.28394791, 183.27756383, 304.42771707, 349.85902453,
266.56829418, 266.56829418, 311.99960164, 160.5619101
175.70567925, 228.70887129, 236.28075587, 243.85264045,
258.9964096 , 274.14017876, 327.1433708 , 160.5619101 ,
258.9964096 , 289.28394791, 183.27756383, 190.84944841,
221.13698672, 221.13698672, 266.56829418, 213.56510214,
327.1433708 , 213.56510214, 296.85583249, 319.57148622,
205.99321756, 198.42133298, 236.28075587, 152.99002552,
327.1433708 , 175.70567925, 236.28075587, 311.99960164,
304.42771707, 236.28075587, 258.9964096 ])
```

from sklearn.metrics import mean_absolute_error
print("MAE",mean_absolute_error(Y_test,Y_pred))

```
from sklearn.metrics import mean squared error
print("MSE", mean_squared_error(Y_test, Y_pred))
MSE 761.2896508697498
print("RMSE",np.sqrt(mean_squared_error(Y_test,Y_pred)))
RMSE 27.59147786672091
print("RMSE",np.log(np.sqrt(mean squared error(Y test,Y pred))))
RMSE 3.3175069520980287
print("RMSE",np.log(np.sqrt(mean squared error(Y test,Y pred))))
RMSE 3.3175069520980287
#c. Visualize simple regression model.
%matplotlib inline
plt.xlabel('CO2EMISSIONS')
plt.ylabel('FUELCONSUMPTION COMB MPG')
plt.scatter(df.CO2EMISSIONS,df.FUELCONSUMPTION COMB MPG,color=
'orange', marker='*')
plt.plot(X train, lr.predict(X train), color = "blue")
[<matplotlib.lines.Line2D at 0x13352709a00>]
```

