

Roll No. 412039

AIM: To Implement Multiple Linear Regression using Python.

Code & Output:

In [2]:

```
import pandas as pd
path_to_file = 'Downloads/petrol_consumption.csv'
df = pd.read_csv(path_to_file)
```

In [3]:

```
df.head()
```

Out[3]:

	Petrol_tax	Average_income	Paved_Highways	Population_Driver_licence(%)	Petrol_Consumption
0	9.0	3571	1976	0.525	541
1	9.0	4092	1250	0.572	524
2	9.0	3865	1586	0.580	561
3	7.5	4870	2351	0.529	414
4	8.0	4399	431	0.544	410

In [4]:

```
print(df.describe().round(2).T)
```

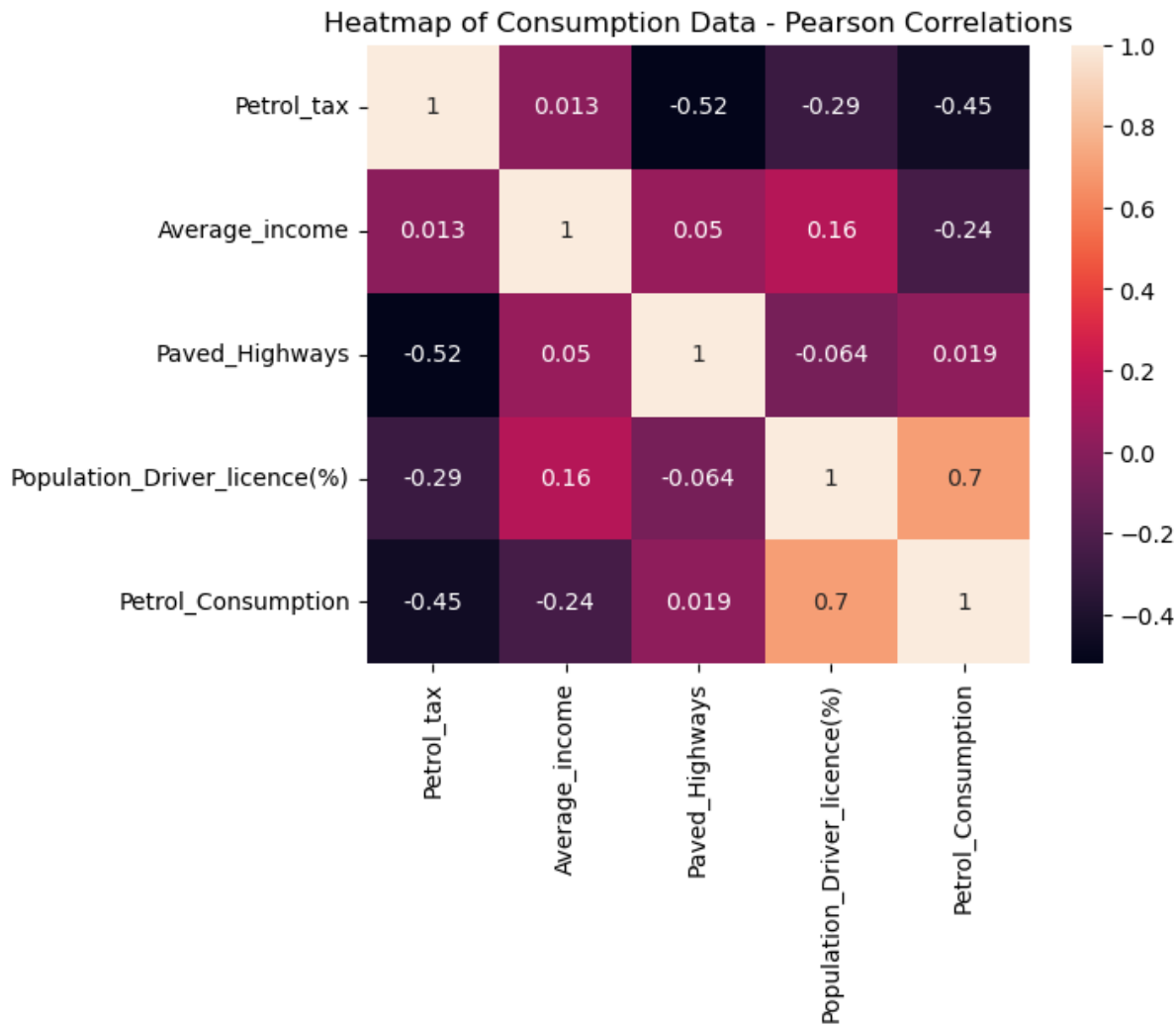
	count	mean	std	min	25%	\
Petrol_tax	48.0	7.67	0.95	5.00	7.00	
Average_income	48.0	4241.83	573.62	3063.00	3739.00	
Paved_Highways	48.0	5565.42	3491.51	431.00	3110.25	
Population_Driver_licence(%)	48.0	0.57	0.06	0.45	0.53	
Petrol_Consumption	48.0	576.77	111.89	344.00	509.50	
	50%	75%	max			
Petrol_tax	7.50	8.12	10.00			
Average_income	4298.00	4578.75	5342.00			
Paved_Highways	4735.50	7156.00	17782.00			
Population_Driver_licence(%)	0.56	0.60	0.72			
Petrol_Consumption	568.50	632.75	968.00			

In [7]:

```

correlations = df.corr()
# annot=True displays the correlation values
sns.heatmap(correlations, annot=True).set(title='Heatmap of Consumption Data - Pearson Correlations')

```



In [8]:

```

y = df['Petrol_Consumption']
X = df[['Average_income', 'Paved_Highways',
        'Population_Driver_licence(%)', 'Petrol_tax']]

```

In [9]:

```

from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)

```

In [11]:

```

from sklearn.linear_model import LinearRegression
regressor = LinearRegression()
regressor.fit(X_train, y_train)

```

Out[11]:

LinearRegression()

In [12]:

```
regressor.intercept_
```

Out[12]:

361.4508790666836

In [13]:

```
regressor.coef_
```

Out[13]:

```
array([-5.65355145e-02, -4.38217137e-03,  1.34686930e+03, -3.69937459e+01])
```

In [15]:

```
feature_names = X.columns
print(feature_names)
```

```
Index(['Average_income', 'Paved_Highways', 'Population_Driver_licence(%)',
      'Petrol_tax'],
      dtype='object')
```

In [16]:

```
feature_names = X.columns
model_coefficients = regressor.coef_

coefficients_df = pd.DataFrame(data = model_coefficients,
                               index = feature_names,
                               columns = ['Coefficient value'])
print(coefficients_df)
```

	Coefficient value
Average_income	-0.056536
Paved_Highways	-0.004382
Population_Driver_licence(%)	1346.869298
Petrol_tax	-36.993746

In [17]:

```
y_pred = regressor.predict(X_test)
```

In [18]:

```
results = pd.DataFrame({'Actual': y_test, 'Predicted': y_pred})
print(results)
```

	Actual	Predicted
27	631	606.692665
40	587	673.779442
26	577	584.991490
43	591	563.536910
24	460	519.058672
37	704	643.461003
12	525	572.897614
19	640	687.077036
4	410	547.609366
25	566	530.037630

In [21]:

```
from sklearn.metrics import mean_absolute_error, mean_squared_error
import numpy as np
mae = mean_absolute_error(y_test, y_pred)
mse = mean_squared_error(y_test, y_pred)
rmse = np.sqrt(mse)

print(f'Mean absolute error: {mae:.2f}')
print(f'Mean squared error: {mse:.2f}')
print(f'Root mean squared error: {rmse:.2f}')
```

Mean absolute error: 53.47
Mean squared error: 4083.26
Root mean squared error: 63.90

In [22]:

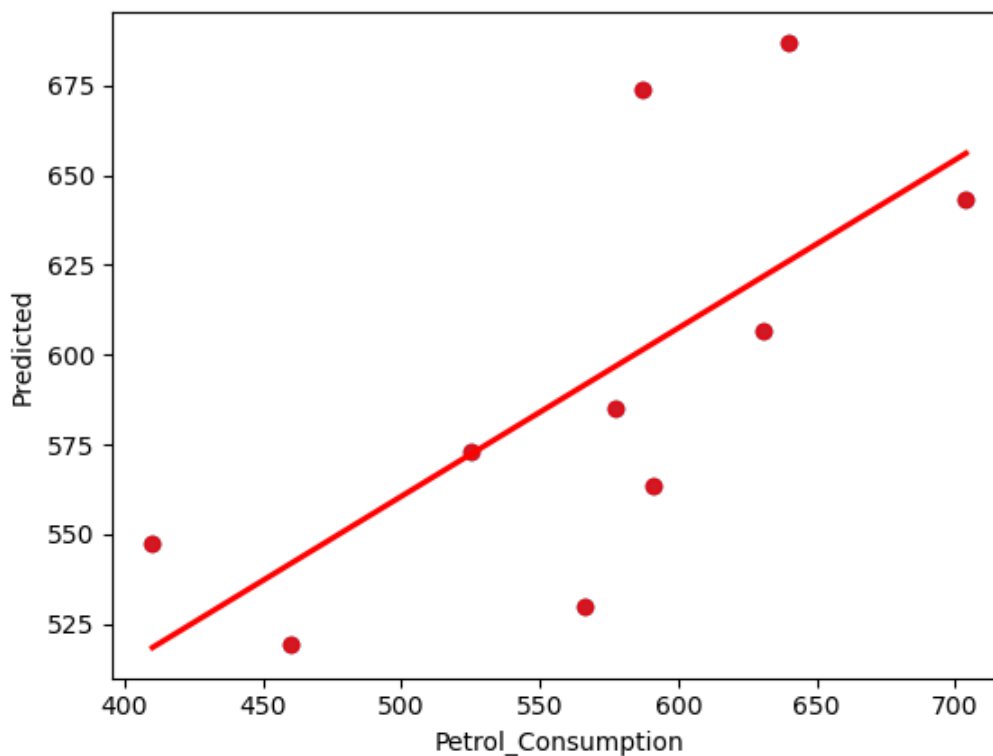
```
regressor.score(X_train, y_train)
```

Out[22]:

0.7068781342155135

In [23]:

```
plt.scatter(y_test,y_pred);
plt.xlabel('Actual');
plt.ylabel('Predicted');
sns.regplot(x=y_test,y=y_pred,ci=None,color='red');
```



Conclusion: Thus, we implemented multiple Linear Regression using sklearn library of python.

