Project Title:

"Price Prediction of European Motorbikes using Linear Regression and Standard Scaler"

Hypothesis:

"The price of European motorbikes is influenced by various factors such as mileage, power, fuel type, offer type, and gear system. By applying Linear Regression, we hypothesize that these features will help predict the price accurately."

Method:

1. **Data Collection**:

- The dataset includes features such as mileage, power, fuel type, offer type, and gear system.
- The target variable is the **price** of the motorbike.



2. Data Preprocessing:

- Missing data handling, if necessary.
- Convert categorical variables (like fuel type, offer type) into numerical form using techniques like one-hot encoding.
- Features like **mileage** and **power** vary in scale, so they are standardized using **Standard Scaler**.

3. Model Selection:

o **Linear Regression** is chosen as the model because it can map the relationship between multiple independent variables and the dependent variable (price) in a linear fashion.

4. Model Training:

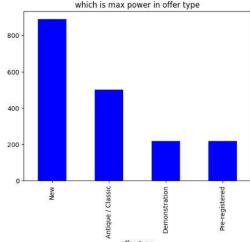
- o After preprocessing, the dataset is split into **training** and **test** sets.
- Linear Regression is trained on the training set, with standardized features, to learn the relationship between the predictors and the price.

5. Evaluation:

The model's performance is evaluated using metrics such as **Mean Squared Error (MSE)** and **R-squared** on the test

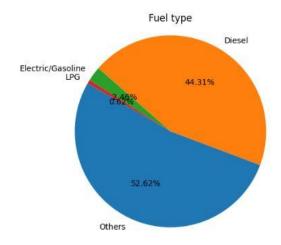
About the Figures:

1. Bar Plot (offertype vs MaxPower):



2. Pie Chart(Fuel type Distribution):

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Results:

1. Model Performance:

- The Linear Regression model with Standard Scaler resulted in an R-squared value of indicating that the model explains of the variability in the motorbike prices.
- The **Mean Squared Error (MSE)** on the test set was which shows the average squared difference between the actual price and the predicted price.

2. Conclusion:

- The model predicts motorbike prices reasonably well based on mileage, power, fuel type, offer type, and gear.
- Features like power and fuel type showed a strong positive correlation with price, while mileage showed a negative correlation.
- Further improvements could be made by incorporating more complex models or additional features like brand or year of manufacture

```
from sklearn.metrics import mean_absolute_error,mean_squared_error,r2_score

def modelresults(predictions):
    print("Mean error on model is {}".format(mean_absolute_error(y_test,predictions)))
    print("Root mean squared error on model is{}".format(np.sqrt(mean_squared_error(y_test,predictions))))
    print("r2scoreis{}".format (r2_score(y_test,predictions)))
✓ 0.0s
```