In [1]: **import** numpy **as** np import pandas as pd import math import matplotlib.pyplot as plt import seaborn as sns from sklearn.model_selection import train_test_split from sklearn.linear_model import LinearRegression from sklearn.tree import DecisionTreeRegressor from sklearn.metrics import mean_absolute_error, r2_score, mean_squared_error from sklearn.metrics import accuracy_score df = pd.read_csv('https://raw.githubusercontent.com/amankharwal/Website-data/master/CarPrice.csv') In [2]: df.head() carbody drivewheel enginelocation wheelbase ... enginesize fuelsystem boreratio stroke compressionratio hor Out[2]: car_ID symboling CarName fueltype aspiration doornumber alfa-romero 0 1 3 two convertible front 88.6 ... 130 2.68 9.0 std rwd mpfi 3.47 gas giulia alfa-romero 2 two convertible 88.6 ... 130 2.68 9.0 1 std rwd front mpfi 3.47 gas stelvio alfa-romero 2 3 hatchback front 152 9.0 std rwd 94.5 ... mpfi 2.68 3.47 gas two Quadrifoglio 3 audi 100 ls 10.0 std four sedan fwd front 99.8 ... 109 mpfi 3.19 3.40 5 2 audi 100ls 99.4 ... 136 3.40 8.0 4 gas std four sedan 4wd front mpfi 3.19 5 rows × 26 columns df.dtypes In [3]: int64 car_ID Out[3]: int64 symboling CarName object fueltype object aspiration object doornumber object carbody object drivewheel object object enginelocation wheelbase float64 carlength float64 carwidth float64 carheight float64 curbweight int64 enginetype object cylindernumber object enginesize int64 fuelsystem object boreratio float64 stroke float64 compressionratio float64 int64 horsepower peakrpm int64 citympg int64 highwaympg int64 float64 price dtype: object df.shape (205, 26)Out[4]: df.isnull().sum() In [5]: 0 car_ID Out[5]: symboling 0 CarName 0 fueltype 0 aspiration 0 doornumber 0 carbody 0 drivewheel 0 enginelocation 0 wheelbase 0 carlength 0 carwidth 0 carheight curbweight enginetype cylindernumber 0 enginesize 0 fuelsystem 0 boreratio 0 stroke 0 compressionratio 0 0 horsepower peakrpm 0 0 citympg highwaympg 0 price 0 dtype: int64 df.duplicated().sum() In [6]: Out[6]: df.drop(columns=['car_ID'], axis=1, inplace=True) In [8]: df CarName fueltype aspiration doornumber carbody drivewheel enginelocation wheelbase carlength ... enginesize fuelsystem boreratio stroke compressionratio Out[8]: symboling alfa-romero 0 3 std convertible front 88.6 168.8 ... 130 mpfi 3.47 2.68 9.0 gas two rwd giulia alfa-romero 1 std two convertible rwd front 88.6 168.8 ... 130 mpfi 3.47 2.68 9.0 gas stelvio alfa-romero 2 std hatchback front 94.5 171.2 ... 152 mpfi 2.68 3.47 9.0 gas two rwd Quadrifoglio 3 audi 100 ls std four sedan fwd front 99.8 176.6 ... 109 mpfi 3.19 3.40 10.0 gas 4 2 audi 100ls 99.4 176.6 ... 136 3.40 8.0 gas std four sedan 4wd front mpfi 3.19 volvo 145e 200 sedan front 109.1 188.8 ... 141 mpfi 3.78 3.15 9.5 std four rwd gas (sw) 201 -1 volvo 144ea turbo sedan 109.1 188.8 ... 141 3.15 8.7 gas four rwd front mpfi 3.78 188.8 ... 8.8 202 volvo 244dl gas std four sedan rwd front 109.1 173 mpfi 3.58 2.87 203 volvo 246 109.1 188.8 ... 145 idi 23.0 diesel turbo four sedan rwd front 3.01 3.40 204 109.1 188.8 ... 141 3.78 9.5 -1 volvo 264gl gas turbo four sedan rwd front mpfi 3.15 205 rows × 25 columns In [9]: df.describe() Out[9]: symboling wheelbase carlength carwidth carheight curbweight enginesize boreratio stroke compressionratio horsepower peakrpm citympg highwaympg 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 **count** 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 0.834146 98.756585 174.049268 30.751220 65.907805 53.724878 2555.565854 126.907317 3.329756 3.255415 10.142537 104.117073 5125.121951 25.219512 mean 1.245307 2.443522 0.270844 0.313597 6.542142 6.021776 12.337289 2.145204 520.680204 41.642693 3.972040 39.544167 476.985643 6.886443 std -2.000000 86.600000 141.100000 60.300000 48.000000 4150.000000 16.000000 min 47.800000 1488.000000 61.000000 2.540000 2.070000 7.000000 13.000000 0.000000 70.000000 4800.000000 25% 94.500000 166.300000 64.100000 52.000000 2145.000000 97.000000 3.150000 3.110000 8.600000 19.000000 25.000000 1.000000 97.000000 173.200000 3.310000 3.290000 9.000000 95.000000 5200.000000 30.000000 **75**% 2.000000 102.400000 183.100000 3.580000 3.410000 116.000000 5500.000000 30.000000 34.000000 66.900000 55.500000 2935.000000 141.000000 9.400000 3.000000 120.900000 208.100000 23.000000 72.300000 59.800000 4066.000000 326.000000 3.940000 4.170000 288.000000 6600.000000 49.000000 54.000000 max plt.figure(figsize = (10, 10)) In [10]: sns.jointplot(data = df) plt.show() <Figure size 1000x1000 with 0 Axes> symboling wheelbase 50000 carlength carwidth carheight 40000 curbweight enginesize boreratio stroke 30000 compressionratio horsepower peakrpm 20000 citympg highwaympg price 10000 0 -50 0 50 100 150 200 250 sns.set_style('darkgrid') In [13]: plt.figure(figsize=(15, 10)) sns.distplot(df.price , color ="Orange") plt.show() C:\Users\KIIT\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level fu nction for histograms). warnings.warn(msg, FutureWarning) 0.00010 0.00008 0.00006 Density 0.00004 0.00002 0.00000 0 30000 50000 10000 20000 40000 price plt.figure(figsize=(1000,800)) In [18]: sns.set_style('ticks') sns.pairplot(df) <seaborn.axisgrid.PairGrid at 0x1c153b27b20> Out[18]: <Figure size 100000x80000 with 0 Axes> 1 w 7 19 No. In [28]: plt.figure(figsize = (10,10)) sns.heatmap(df.corr(), annot = True, cmap = 'summer',linecolor='black',linewidth=1.0,fmt='0.2f') plt.show() - 1.00 -0.53 0.07 symboling 1.00 -0.36 -0.54 0.27 0.16 wheelbase 0.53 1.00 0.87 0.80 0.78 0.57 0.49 0.35 - 0.75 0.13 0.16 -0.36 0.87 0.68 0.55 carlength 1.00 0.84 0.49 0.88 0.61 -0.29 -0.67 -0.70 0.68 0.87 0.74 0.56 0.18 carwidth - 0.50 carheight 0.54 0.59 0.49 0.28 0.30 0.07 0.17 0.12 1.00 0.26 -0.32 0.78 0.85 0.65 0.17 curbweight 0.88 0.87 0.30 1.00 0.15 0.75 -0.27 -0.76 0.84 - 0.25 enginesize 0.57 0.68 0.74 0.07 0.85 1.00 0.58 0.20 0.81 -0.24 -0.65 -0.68 0.87 0.49 0.61 0.56 0.17 0.65 0.58 1.00 -0.06 0.57 -0.25 -0.58 -0.59 0.55 boreratio - 0.00 0.08 0.16 0.13 0.18 0.17 0.20 1.00 0.19 0.08 stroke compressionratio 0.25 0.16 0.18 0.26 0.15 0.19 -0.20 0.32 0.27 0.07 norsepower 1.00 -0.11 -0.05 peakrpm -0.36 -0.29 -0.22 -0.32 -0.27 -0.24 -0.25 -0.07 -0.44 0.13 -0.47 -0.67 -0.64 -0.05 -0.76 -0.65 -0.58 -0.04 -0.69 0.32 -0.80 -0.11 0.97 1.00 citympg - -0.50 -0.54 -0.70 -0.68 -0.11 -0.80 -0.68 -0.59 -0.04 0.27 -0.77 0.97 1.00 -0.70highwaympg 0.12 0.87 0.55 0.08 0.07 -0.69 1.00 0.58 0.68 0.76 0.84 0.81 -0.70 price - -0.75 stroke aitympg symboling compressionratio wheelbase carlength carheight curbweight enginesize boreratio peakrpm highwaympg carwidth horsepowel data = df[["symboling", "wheelbase", "carlength", "carwidth", "carheight", "curbweight", "enginesize", "boreratio", "stroke", "compressionratio", "l In [32]: Out[32]: symboling wheelbase carlength carwidth carheight curbweight enginesize boreratio stroke compressionratio horsepower peakrpm citympg highwaympg price 0 3 88.6 168.8 64.1 48.8 2548 130 3.47 2.68 9.0 5000 21 27 13495.0 111 3 88.6 168.8 64.1 48.8 2548 130 3.47 2.68 9.0 111 5000 21 27 16500.0 1 2 171.2 2823 152 2.68 3.47 9.0 154 5000 19 26 16500.0 1 94.5 65.5 52.4 2 10.0 30 13950.0 3 99.8 176.6 66.2 54.3 2337 109 3.19 3.40 102 5500 24 4 2 99.4 176.6 66.4 54.3 2824 136 3.19 3.40 8.0 115 5500 18 22 17450.0 200 109.1 188.8 68.9 55.5 2952 141 3.78 3.15 9.5 114 5400 23 28 16845.0 -1 -1 109.1 188.8 68.8 55.5 3049 141 3.78 3.15 8.7 160 5300 19 25 19045.0 201 202 -1 109.1 188.8 68.9 55.5 3012 173 3.58 2.87 134 5500 18 23 21485.0 8.8 203 -1 109.1 188.8 68.9 55.5 3217 145 3.40 23.0 106 4800 26 27 22470.0 3.01 204 -1 109.1 188.8 68.9 55.5 3062 141 3.78 3.15 9.5 5400 19 25 22625.0 114 205 rows × 15 columns In [33]: x = data.drop(["price"], 1)y = data["price"] xtrain, xtest, ytrain, $ytest = train_test_split(x, y, random_state = 16, test_size = 0.25, shuffle=True)$ C:\Users\KIIT\AppData\Local\Temp\ipykernel_12584\313162150.py:1: FutureWarning: In a future version of pandas all arguments of DataFrame.drop excep t for the argument 'labels' will be keyword-only. x = data.drop(["price"], 1) model = DecisionTreeRegressor() In [34]: In [35]: model.fit(xtrain, ytrain) model.score(xtrain, ytrain) 0.9997917694852244Out[35]: Prediction = model.predict(xtest) In [36]: print("Decision Tree Regression Model:") print("************************ print('R2_score:', r2_score(ytest,Prediction)) print('Mean Absolute Error:', mean_absolute_error(ytest,Prediction)) print('Mean Squared Error:', mean_squared_error(ytest,Prediction)) print('Root Mean Squared Error:', np.sqrt(mean_squared_error(ytest,Prediction))) Decision Tree Regression Model: R2_score: 0.9263969588782768 Mean Absolute Error: 1511.8782115384615 Mean Squared Error: 5158535.641382481 Root Mean Squared Error: 2271.2409914807545 In [37]: print("Accuracy of Decision Tree Regression Model: ", model.score(xtrain, ytrain)) Accuracy of Decision Tree Regression Model: 0.9997917694852244 In []: