In [1]: **import** pandas **as** pd from sklearn.model_selection import train_test_split from sklearn.linear_model import LinearRegression from sklearn.metrics import mean_squared_error data=pd.read_csv("downloads/CarPrice_assignment.csv") In [3]: data car_ID symboling CarName fueltype aspiration doornumber carbody drivewheel enginelocation wheelbase ... enginesize fuelsystem boreratio stroke compressionratio horsepower peakrpm cityl Out[3]: alfa-romero 0 3 two convertible 130 9.0 111 5000 1 std front 88.6 ... mpfi 3.47 2.68 gas rwd giulia alfa-romero 2 3 88.6 ... 130 2.68 9.0 111 5000 1 std two convertible front mpfi 3.47 gas rwd stelvio alfa-romero 2 3 152 9.0 5000 hatchback front 94.5 ... 3.47 154 gas std two rwd mpfi 2.68 Ouadrifoglio 5500 3 4 audi 100 ls std four sedan fwd front 99.8 ... 109 mpfi 3.19 3.40 10.0 102 5 4 2 audi 100ls gas std four sedan 4wd front 99.4 ... 136 mpfi 3.19 3.40 8.0 115 5500 volvo 145e 200 201 std sedan front 109.1 ... 141 mpfi 3.15 9.5 114 5400 gas four rwd 3.78 (sw) 201 202 -1 volvo 144ea turbo sedan front 109.1 ... 141 mpfi 3.78 3.15 8.7 160 5300 gas four rwd 203 173 5500 202 -1 volvo 244dl gas std four sedan rwd front 109.1 ... mpfi 3.58 2.87 8.8 134 203 204 volvo 246 front 109.1 ... 145 idi 3.01 3.40 23.0 106 4800 diesel turbo four sedan rwd 204 205 -1 volvo 264gl gas turbo four sedan rwd front 109.1 ... 141 mpfi 3.78 3.15 9.5 114 5400 205 rows × 26 columns In [4]: data.head(10) car_ID symboling CarName fueltype aspiration doornumber carbody drivewheel enginelocation wheelbase ... enginesize fuelsystem boreratio stroke compressionratio horsepower Out[4]: peakrpm citymr alfa-romero 0 3 convertible 88.6 ... 130 mpfi 3.47 2.68 9.0 111 5000 gas std two rwd front giulia alfa-romero 2 3 two convertible front 88.6 ... 130 3.47 2.68 9.0 111 5000 gas std rwd mpfi stelvio alfa-romero 2 3 hatchback 94.5 ... 152 2.68 3.47 9.0 154 5000 gas std two rwd front mpfi Quadrifoglio 3 audi 100 ls std sedan fwd front 99.8 ... 109 mpfi 3.19 3.40 10.0 102 5500 four gas 5 2 99.4 ... 8.0 4 audi 100ls gas std four sedan 4wd front 136 mpfi 3.19 3.40 115 5500 5 6 2 audi fox sedan fwd front 99.8 ... 136 3.40 8.5 110 5500 std two mpfi 3.19 gas 7 136 3.40 8.5 6 1 audi 100ls gas std four sedan fwd front 105.8 ... mpfi 3.19 110 5500 8 audi 5000 std four wagon fwd front 105.8 ... 136 mpfi 3.19 3.40 8.5 110 5500 gas 8 9 audi 4000 turbo four sedan fwd front 105.8 ... 131 mpfi 3.13 3.40 8.3 140 5500 gas audi 5000s two hatchback front 99.5 ... 131 3.40 7.0 160 5500 gas turbo mpfi 3.13 (diesel) 10 rows × 26 columns In [5]: data.tail() car_ID symboling CarName fueltype aspiration doornumber carbody drivewheel enginelocation wheelbase ... enginesize fuelsystem boreratio stroke compressionratio horsepower peakrpm citympg Out[5]: volvo 200 201 std four sedan rwd front 109.1 ... 141 3.78 3.15 9.5 114 5400 23 gas 145e (sw) volvo 201 202 turbo four sedan rwd front 109.1 ... 141 mpfi 3.78 3.15 8.7 160 5300 19 144ea VOIVC front 109.1 2.87 134 5500 18 gas sedan 244dl 109.1 ... 3.40 4800 26 -1 volvo 246 diesel turbo four sedan rwd front 3.01 106 volvo 204 205 four sedan 109.1 ... 141 3.78 3.15 9.5 114 5400 19 -1 gas turbo rwd front mpfi 264gl 5 rows × 26 columns data.columns In [6]: Index(['car_ID', 'symboling', 'CarName', 'fueltype', 'aspiration', 'doornumber', 'carbody', 'drivewheel', 'enginelocation', 'wheelbase', 'carlength', 'carwidth', 'carheight', 'curbweight', 'enginetype', 'cylindernumber', 'enginesize', 'fuelsystem', 'boreratio', 'stroke', 'compressionratio', 'horsepower', 'peakrpm', 'citympg', 'highwaympg', 'price'], dtype='object') data.describe() carwidth Out[7]: car_ID symboling wheelbase carlength carheight curbweight enginesize boreratio stroke compressionratio horsepower peakrpm citympg highwaympg pric **count** 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.000000 205.00000 mean 103.000000 0.834146 98.756585 174.049268 65.907805 53.724878 2555.565854 126.907317 3.329756 3.255415 10.142537 104.117073 5125.121951 25.219512 30.751220 13276.71057 0.270844 476.985643 6.542142 6.886443 7988.85233 std 59.322565 1.245307 6.021776 12.337289 2.145204 2.443522 520.680204 41.642693 0.313597 3.972040 39.544167 1.000000 -2.000000 86.600000 141.100000 60.300000 47.800000 1488.000000 61.000000 2.540000 2.070000 7.000000 48.000000 4150.000000 13.000000 16.000000 5118.00000 min 25.000000 25% 52.000000 0.000000 94.500000 166.300000 64.100000 52.000000 2145.000000 97.000000 3.150000 3.110000 8.600000 70.000000 4800.000000 19.000000 7788.00000 **50**% 103.000000 1.000000 97.000000 173.200000 65.500000 54.100000 2414.000000 120.000000 3.310000 3.290000 9.000000 95.000000 5200.000000 24.000000 30.000000 10295.00000 **75**% 154.000000 2.000000 102.400000 183.100000 66.900000 55.500000 2935.000000 141.000000 3.580000 3.410000 9.400000 116.000000 5500.000000 30.000000 34.000000 16503.00000 max 205.000000 3.000000 120.900000 208.100000 72.300000 59.800000 4066.000000 326.000000 3.940000 4.170000 23.000000 288.000000 6600.000000 49.000000 54.000000 45400.00000 In [8]: x=data[['horsepower', 'enginesize']] y=data['price'] x_train, x_test, y_train, y_test=train_test_split(x, y, test_size=0.2, random_state=42) In [9]: model=LinearRegression() In [10]: In [11]: model.fit(x_train,y_train) Out[11]: ▼ LinearRegression LinearRegression() y_pred=model.predict(x_test) In [12]:

In [13]: | mse=mean_squared_error(y_test,y_pred)

print('mean squared error :',mse)

mean squared error : 15096540.151434017