

```
Car Price Prediction.py > ...
168
169 # Deploy the model of car Price Prediction
170
171 import streamlit as st
172 import pandas as pd
173 import pickle
174 import numpy as np
175 import matplotlib.pyplot as plt
176 import seaborn as sns
177 from sklearn.linear_model import LinearRegression
178 from sklearn.preprocessing import MinMaxScaler
179 from sklearn.model_selection import train_test_split
180
181 # Convert categorical columns to dummy variables
182 df = pd.get_dummies(df, drop_first=True) # Drop first to avoid dummy variable trap
183
```

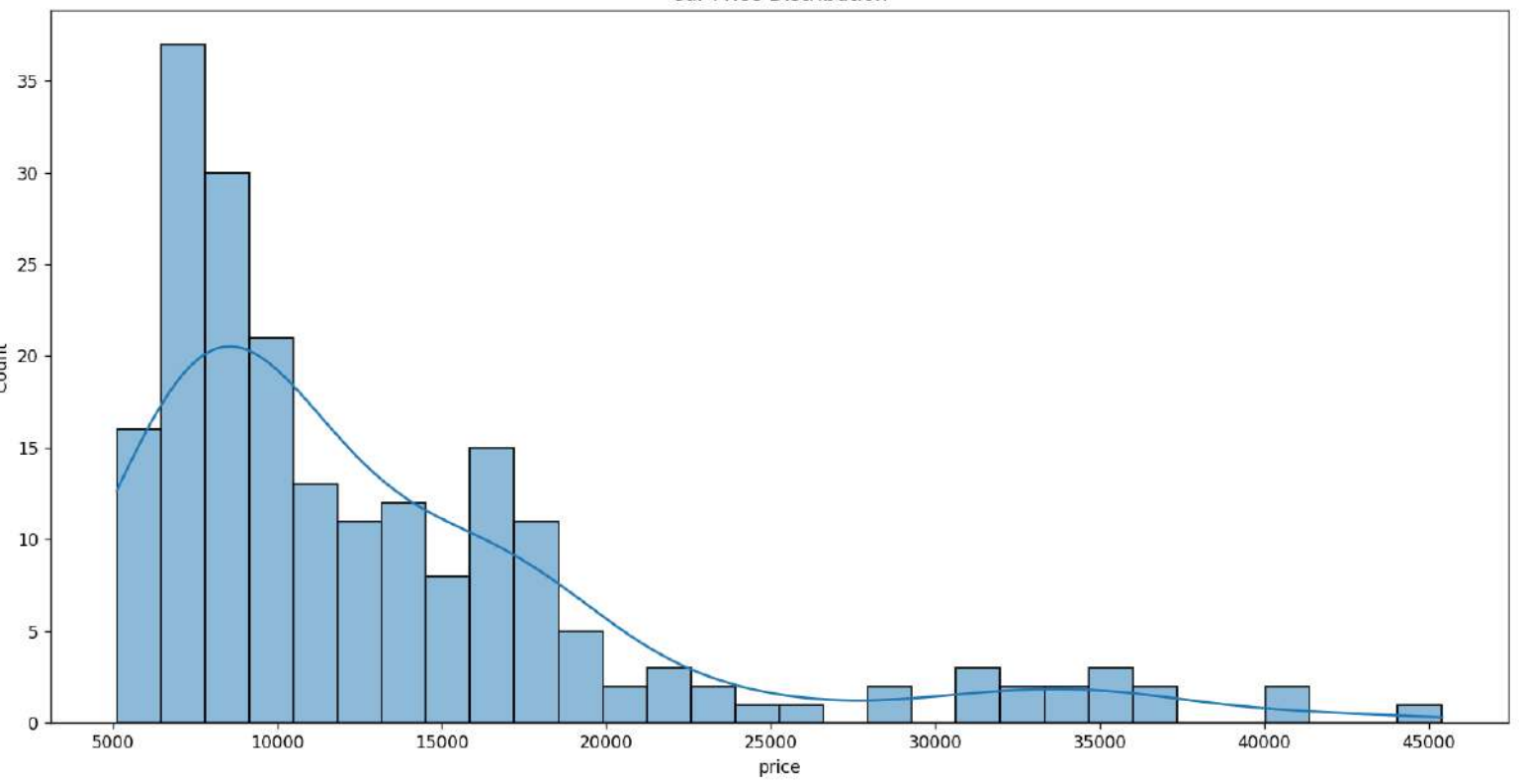
PROBLEMS 15 OUTPUT DEBUG CONSOLE TERMINAL PORTS

OLS Regression Results

Dep. Variable:	price	R-squared (uncentered):	0.842
Model:	OLS	Adj. R-squared (uncentered):	0.832
Method:	Least Squares	F-statistic:	82.09
Date:	Wed, 26 Feb 2025	Prob (F-statistic):	1.64e-56
Time:	16:08:31	Log-Likelihood:	-1661.8
No. Observations:	164	AIC:	3344.
Df Residuals:	154	BIC:	3375.
Df Model:	10		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
carbody_hatchback	6282.1172	1195.344	5.255	0.000	3920.730	8643.505
carbody_sedan	6709.4432	1182.427	5.674	0.000	4373.572	9045.314
cylindernumber_four	3822.3409	1018.977	3.751	0.000	1809.365	5835.317
companyname_bmw	1.483e+04	2737.947	5.416	0.000	9420.161	2.02e+04
companyname_buick	3.321e+04	3195.508	10.391	0.000	2.69e+04	3.95e+04
companyname_jaguar	2.789e+04	3814.069	7.313	0.000	2.04e+04	3.54e+04

Car Price Distribution



Enter Car Details

Enter car\_ID:

1.00

-

+

Enter symboling:

-2.00

-

+

Enter wheelbase:

86.60

-

+

Enter carlength:

141.10

-

+

Enter carwidth:

60.30

-

+

Enter carheight:

47.80

-

+

Enter curbweight:

1488.00

-

+

	Features	VIF
0	carbody_hatchback	0.04
1	carbody_sedan	0.04
2	cylindernumber_four	0.03
3	companyname_bmw	0.19
4	companyname_buick	0.26
5	companyname_jaguar	0.37
6	companyname_porcshce	1
7	companyname_porsche	0.26
8	companyname_saab	0.21
9	companyname_volvo	0.14

Car Price Prediction 🚗💰