# In [7]: import warnings warnings.filterwarnings("ignore") import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns from sklearn.model\_selection import train\_test\_split from sklearn.preprocessing import StandardScaler, OneHotEncoder from sklearn.impute import SimpleImputer from sklearn.compose import ColumnTransformer from sklearn.prepline import Pipeline from sklearn.linear\_model import LinearRegression from sklearn.ensemble import RandomForestRegressor from sklearn.metrics import mean\_squared\_error, r2\_score

# In [8]: df =pd.read\_csv("CarPricesPrediction.csv") df

### Out[8]:

	Unnamed: 0	Make	Model	Year	Mileage	Condition	Price
0	0	Ford	Silverado	2022	18107	Excellent	19094.75
1	1	Toyota	Silverado	2014	13578	Excellent	27321.10
2	2	Chevrolet	Civic	2016	46054	Good	23697.30
3	3	Ford	Civic	2022	34981	Excellent	18251.05
4	4	Chevrolet	Civic	2019	63565	Excellent	19821.85
995	995	Nissan	Camry	2010	149032	Excellent	24548.50
996	996	Chevrolet	F-150	2014	20608	Excellent	26969.70
997	997	Ford	Altima	2016	109851	Good	20507.55
998	998	Toyota	Silverado	2010	11704	Good	31414.90
999	999	Nissan	Silverado	2017	128390	Excellent	18580.60

1000 rows × 7 columns

```
In [25]: df.drop("Unnamed: 0",axis=1,inplace=True)
df
```

## Out[25]:

	Make	Model	Year	Mileage	Condition	Price
0	Ford	Silverado	2022	18107	Excellent	19094.75
1	Toyota	Silverado	2014	13578	Excellent	27321.10
2	Chevrolet	Civic	2016	46054	Good	23697.30
3	Ford	Civic	2022	34981	Excellent	18251.05
4	Chevrolet	Civic	2019	63565	Excellent	19821.85
995	Nissan	Camry	2010	149032	Excellent	24548.50
996	Chevrolet	F-150	2014	20608	Excellent	26969.70
997	Ford	Altima	2016	109851	Good	20507.55
998	Toyota	Silverado	2010	11704	Good	31414.90
999	Nissan	Silverado	2017	128390	Excellent	18580.60

1000 rows × 6 columns

```
In [21]: X = df.drop(columns=['Price'])
y = df['Price']
```

### Out[21]:

		Mala	Madal	V	Mileen	0	Duine
	Unnamed: 0	Make	wodel	rear	willeage	Condition	Price
0	0	Ford	Silverado	2022	18107	Excellent	19094.75
1	1	Toyota	Silverado	2014	13578	Excellent	27321.10
2	2	Chevrolet	Civic	2016	46054	Good	23697.30
3	3	Ford	Civic	2022	34981	Excellent	18251.05
4	4	Chevrolet	Civic	2019	63565	Excellent	19821.85
995	995	Nissan	Camry	2010	149032	Excellent	24548.50
996	996	Chevrolet	F <b>-</b> 150	2014	20608	Excellent	26969.70
997	997	Ford	Altima	2016	109851	Good	20507.55
998	998	Toyota	Silverado	2010	11704	Good	31414.90
999	999	Nissan	Silverado	2017	128390	Excellent	18580.60

```
In [10]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42
```

```
In [15]:
         regressor.fit(X_train, y_train)
Out[15]:
                                                   Pipeline
                                                                                StandardScaler
           ())]),
                                                              ['Year', 'Mileage']),
                                                              ('cat',
                                                              Pipeline(steps=[('imputer',
                                                                                SimpleImputer
           (fill_value='missing',
           strategy='constant')),
                                                                               ('onehot',
                                       preprocessor: QolumnTransformer
                                                                SimpleImputer(strategy='media _
            n')),
                                                               ('scaler', StandardScaler
            ())]),
                                              ['Year', 'Mileage']),
                                             ('cat',
                                              Pipeline(steps=[('imputer',
                                                                SimpleImputer(fill_value='mis
            sing',
                                                                              strategy='const ▼
                           num
                                                                      dat
                                            ['Make', 'Model', 'Condition']
            ['Year', 'Mileage']
                      SimpleImputer
                                                                 SimpleImputer
             SimpleImputer(strategy='medi
                                             SimpleImputer(fill value='missing', strategy='cons
             an')
                                             tant')
                    ▼ StandardScaler
                                                                 OneHot Encoder
                                                   OneHotEncoder(handle_unknown='ignore')
                    StandardScaler()
                                           ▼ RandomForestRegressor
                                           RandomForestRegressor()
In [16]: y_pred = regressor.predict(X_test)
```

```
rmse = mean_squared_error(y_test, y_pred, squared=False)
r2 = r2_score(y_test, y_pred)
print(f'Root Mean Squared Error: {rmse}')
print(f'R-squared: {r2}')
```

Root Mean Squared Error: 135.27231033182338

R-squared: 0.9990969527286926

```
In [19]: def predict_price(make, model, year, mileage, condition):
    input_data = pd.DataFrame([[year, mileage, make, model, condition]], columns=['Year'
    predicted_price = regressor.predict(input_data)
    return predicted_price[0]

In [20]: predicted_price = predict_price('Toyota', 'Camry', 2018, 50000, 'Good')
    print(f'Predicted price: ${predicted_price:.2f}')

Predicted price: $21480.26
In []:
```