20CP412P 21BCP359

PRACTICAL 9

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Code

```
import pandas as pd
import statsmodels.api as sm
df = pd.read csv('CarPrice Assignment.csv')
X = df['horsepower']
y = df['citympg']
X = sm.add constant(X)
# Create the regression model
model\ citympg = sm.OLS(y, X).fit()
print(model citympg.summary())
X highway = df['horsepower']
y highway = df['highwaympg']
X \text{ highway} = \text{sm.add constant}(X \text{ highway})
# Create the regression model
model highway = sm.OLS(y highway, X highway).fit()
print(model highway.summary())
import matplotlib.pyplot as plt
import seaborn as sns
# Scatterplot for citympg vs. horsepower
plt.figure(figsize=(8, 6))
sns.scatterplot(x=df['horsepower'], y=df['citympg'], color='blue')
plt.plot(df['horsepower'],
                            model citympg.predict(sm.add constant(df['horsepower'])),
                                                                                             color='red',
label='Regression Line')
plt.title('Scatterplot of City MPG vs Horsepower')
plt.xlabel('Horsepower')
plt.ylabel('City MPG')
plt.legend()
plt.show()
# Scatterplot for highwaympg vs. horsepower
plt.figure(figsize=(8, 6))
sns.scatterplot(x=df['horsepower'], y=df['highwaympg'], color='green')
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Output the model summary

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print(model curbweight.summary())
# Scatterplot for Engine Size vs. Price
plt.figure(figsize=(14, 6))
plt.subplot(1, 2, 1)
sns.scatterplot(x=df['enginesize'], y=df['price'], color='blue')
                            model enginesize.predict(sm.add constant(df['enginesize'])),
plt.plot(df['enginesize'],
                                                                                              color='red'.
label='Regression Line')
plt.title('Scatterplot of Engine Size vs Price')
plt.xlabel('Engine Size (L)')
plt.ylabel('Price ($)')
plt.legend()
# Scatterplot for Curb Weight vs. Price
plt.subplot(1, 2, 2)
sns.scatterplot(x=df['curbweight'], y=df['price'], color='green')
plt.plot(df['curbweight'], model curbweight.predict(sm.add constant(df['curbweight'])), color='orange',
label='Regression Line')
plt.title('Scatterplot of Curb Weight vs Price')
plt.xlabel('Curb Weight (lbs)')
plt.ylabel('Price ($)')
plt.legend()
# Show the plots
plt.tight layout()
plt.show()
import pandas as pd
import statsmodels.api as sm
from statsmodels.stats.outliers influence import variance inflation factor
df = pd.read csv('CarPrice Assignment.csv')
# Select numeric variables except 'citympg' and 'highwaympg'
numeric df = df.drop(['price', 'citympg', 'highwaympg'], axis=1).select dtypes(include=[float, int])
independent vars = sm.add constant(numeric df)
# Variance Inflation Factor (VIF) calculation
vif data = pd.DataFrame()
vif data['Feature'] = independent vars.columns
vif data['VIF']
                          [variance inflation factor(independent vars.values,
                                                                                    i)
                                                                                          for
                                                                                                  i
                                                                                                        in
range(independent vars.shape[1])]
print(vif data)
```

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Output

