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Subject: ML (Machine Learning)

EXPERIMENT 1

Exploratory data analysis in car price prediction dataset

1. Import dataset available at following url:

https://archive.ics.uci.edu/ml/machine-learning-databases/autos/imports-85.data

→ The First 5 rows of the dataframe

	symbolin	g no	ormalized- losses	make	fuel- type	aspiration	num- of- doors	•	drive- wheels	engine- location	wheel- base	•••	engine- size	fuel- system	bore	stroke
0		3	?	alfa- romero	gas	std	two	convertible	rwd	front	88.6		130	mpfi	3.47	2.68
1		1	?	alfa- romero	gas	std	two	hatchback	rwd	front	94.5		152	mpfi	2.68	3.47
2		2	164	audi	gas	std	four	sedan	fwd	front	99.8		109	mpfi	3.19	3.40
3		2	164	audi	gas	std	four	sedan	4wd	front	99.4		136	mpfi	3.19	3.40
4		2	?	audi	gas	std	two	sedan	fwd	front	99.8		136	mpfi	3.19	3.40
_																

 $5 \text{ rows} \times 26 \text{ columns}$

✓ 2. Clean dataset.(hint: Replace missing values):

```
import pandas as pd
path = "https://archive.ics.uci.edu/ml/machine-learning-databases/autos/imports-85.data"
df = pd.read_csv(path, header=None, na_values="?")
headers=["symboling", "normalized-losses", "make", "fuel-type", "aspiration", "num-of-doors", "body-style",
            drive-wheels", "engine-location", "wheel-base", "length", "width", "height", "curb-weight", "engine-type",
           "num-of-cylinders", "engine-size", "fuel-system", "bore", "stroke", "compression-ratio", "horsepower",
            "peak-rpm", "city-mpg", "highway-mpg", "price"]
df.columns = headers
numeric_columns = df.select_dtypes(include=['float64', 'int64']).columns
\label{eq:df_numeric_columns} \texttt{df}[\texttt{numeric\_columns}]. \texttt{interpolate}(\texttt{method='linear',limit\_direction='forward',axix=0}) \\
categorical_columns = df.select_dtypes(include=['object']).columns
for column in categorical_columns:
  df[column].fillna(df[column].mode()[0], inplace=True)
print("\nNumber of missing values after Cleaning:")
print(df.isnull().sum())
print("\nThe first 10 rows of the cleaned dataframe:")
print(df.head(50))
     Number of missing values after Cleaning:
     symboling
                            0
     normalized-losses
                            3
     make
                           0
     fuel-type
                           0
     aspiration
     num-of-doors
                            0
     body-style
                           0
     drive-wheels
                           0
     engine-location
                           0
     wheel-base
                           0
     length
                           a
     width
                           a
     height
                           0
     curb-weight
                            0
     engine-type
     num-of-cylinders
     engine-size
     fuel-system
                            0
     bore
     stroke
     compression-ratio
     horsepower
                            a
     peak-rpm
                            a
     city-mpg
                           0
     highway-mpg
                            0
     price
     dtype: int64
     The first 10 rows of the cleaned dataframe:
                                                 make fuel-type aspiration
         symboling normalized-losses
                                   NaN
                                        alfa-romero
                                                             gas
                                                                        std
     1
                  3
                                    NaN
                                         alfa-romero
                                                             gas
                                                                        std
     2
                  1
                                    NaN
                                         alfa-romero
                                                             gas
                                                                        std
     3
                  2
                            164.000000
                                                 audi
                                                                        std
                                                             gas
     4
                  2
                             164.000000
                                                 audi
                                                             gas
                                                                        std
                  2
                             161.000000
                                                 audi
                                                             gas
                                                                        std
                             158.000000
     6
                  1
                                                 audi
                                                                        std
                                                             gas
                  1
                            158.000000
                                                 audi
                                                                        std
                                                             gas
     8
                  1
                            158.000000
                                                 audi
                                                             gas
                                                                       turbo
     9
                            175.000000
                                                 audi
                                                                      turbo
                  0
                                                             gas
                             192.000000
     10
                                                  bmw
                                                             gas
                                                                        std
     11
                  0
                            192.000000
                                                  bmw
                                                             gas
                                                                        std
     12
                  0
                            188,000000
                                                  bmw
                                                             gas
                                                                        std
     13
                  0
                            188.000000
                                                  hmw
                                                             gas
                                                                        std
     14
                  1
                             174.600000
                                                  bmw
                                                                        std
                                                             gas
     15
                  0
                             161.200000
                                                  bmw
                                                             gas
                                                                         std
     16
                  0
                             147.800000
                                                  bmw
                                                             gas
                                                                        std
     17
                             134.400000
                  0
                                                  bmw
                                                             gas
                                                                        std
                                                                        std
     18
                            121.000000
                                           chevrolet
                                                             gas
     19
                             98.000000
                  1
                                            chevrolet
                                                                        std
                                                             gas
     20
                  0
                             81,000000
                                            chevrolet
                                                             gas
                                                                        std
                             118.000000
     21
                                                dodge
                                                                        std
                  1
                                                             gas
                             118.000000
     22
                  1
                                                dodge
                                                             gas
                                                                        std
                             118.000000
     23
                  1
                                                dodge
                                                             gas
                                                                      turbo
     24
                  1
                             148,000000
                                                dodge
                                                             gas
                                                                        std
                             148,000000
                                                dodge
                                                                        std
```

3. Feature extraction using visualization (hint: matplotlib or seaborn graphs)Only regplots of continuous attributes

```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
path = "https://archive.ics.uci.edu/ml/machine-learning-databases/autos/imports-85.data"
df = pd.read_csv(path, header=None, na_values="?")
"peak-rpm","city-mpg","highway-mpg","price"]
df.columns = headers
numeric columns = df.select dtypes(include=['float64', 'int64']).columns
for column in numeric_columns:
   df[column].fillna(df[column].mean(), inplace=True)
categorical_columns = df.select_dtypes(include=['object']).columns
for column in categorical_columns:
   df[column].fillna(df[column].mode()[0], inplace=True)
continuous_attributes = ["wheel-base", "length", "width", "height", "curb-weight", "engine-size",
                      "bore", "stroke", "compression-ratio", "horsepower", "peak-rpm", "city-mpg", "highway-mpg", "price"]
plt.figure(figsize=(20, 20))
for i, attribute in enumerate(continuous_attributes):
   plt.subplot(7, 2, i + 1)
    sns.regplot(x=attribute, y='price', data=df,color='#FF0000')
   plt.title(f'Regplot of {attribute} vs Price')
plt.tight_layout()
plt.show()
```

