#### Car Price Pridiction

#### Load the Dataset

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
data = pd.read_csv('/content/car data.csv')
print(data.head())
      Car_Name Year Selling_Price Present_Price Driven_kms Fuel_Type \
          ritz 2014
                               3.35
                                              5.59
                                                                  Petrol
           sx4 2013
                               4.75
                                              9.54
                                                                  Diesel
    1
                                                         43000
          ciaz 2017
                               7.25
                                              9.85
                                                          6900
                                                                  Petrol
       wagon r 2011
                               2.85
                                              4.15
                                                          5200
                                                                  Petrol
         swift 2014
                               4.60
                                              6.87
                                                         42450
                                                                  Diesel
      Selling_type Transmission Owner
                         Manual
            Dealer
            Dealer
                         Manual
    2
            Dealer
                         Manual
                                     0
            Dealer
                                     0
    3
                         Manual
            Dealer
                         Manual
                                     0
```

### Display basic information

```
print(data.info())
print(data.describe())
<<class 'pandas.core.frame.DataFrame'>
     RangeIndex: 301 entries, 0 to 300
    Data columns (total 12 columns):
     # Column
                                Non-Null Count Dtype
     0
         Car_Name
                                301 non-null
                                                object
                                301 non-null
                                                int64
     1
         Selling_Price
                                301 non-null
                                                float64
         Present Price
                                301 non-null
                                                float64
     4
         Driven_kms
                                301 non-null
                                                int64
         Fuel_Type
                                301 non-null
                                                object
     6
                                301 non-null
                                                object
         Selling_type
     7
         Transmission
                                301 non-null
                                                object
     8
         Owner
                                301 non-null
                                                int64
                                301 non-null
     9
                                                int64
         Car_Age
     10 Price_per_km
                                301 non-null
                                                float64
     11 Price_Age_Interaction 301 non-null
                                                float64
     dtypes: float64(4), int64(4), object(4)
    memory usage: 28.3+ KB
    None
                  Year Selling_Price Present_Price
                                                         Driven_kms
                                                                         Owner \
```

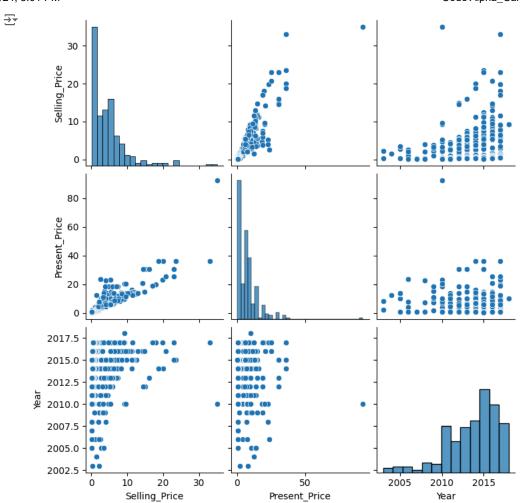
```
301.000000
                       301.000000
                                      301.000000
                                                     301.000000
                                                                 301.000000
count
                         4.322591
                                        7.628472
                                                   36947.205980
                                                                   0.043189
mean
      2013.627907
          2.891554
                         3.867603
                                                   38886.883882
std
                                        8.642584
                                                                   0.247915
       2003.000000
                         0.250000
                                        0.320000
                                                     500.000000
                                                                   0.000000
       2012.000000
                         0.900000
                                        1.200000
                                                   15000.000000
                                                                   0.000000
50%
       2014.000000
                         3.600000
                                        6.400000
                                                   32000.000000
                                                                   0.000000
75%
       2016.000000
                         6.000000
                                        9.900000
                                                   48767.000000
                                                                   0.000000
       2018.000000
                        14.500000
                                       92.600000
                                                  500000.000000
                                                                   3.000000
          Car_Age Price_per_km Price_Age_Interaction
      301.000000
                     301.000000
                                            301.000000
         9.372093
                       0.000355
                                             72.670193
std
         2.891554
                       0.000589
                                             97.689287
         5.000000
                       0.000001
                                              2.560000
min
25%
         7.000000
                       0.000086
                                             11.250000
50%
         9.000000
                       0.000193
                                             55.200000
75%
        11.000000
                       0.000352
                                             95.200000
        20.000000
                       0.006038
                                           1203.800000
/usr/local/lib/python3.10/dist-packages/numpy/lib/function_base.py:4824: UserWarning: Warning: 'partition' will ignore the 'mask' of the MaskedArray.
/usr/local/lib/python3.10/dist-packages/numpy/lib/function_base.py:4824: UserWarning: Warning: 'partition' will ignore the 'mask' of the MaskedArray.
  arr.partition(
/usr/local/lib/python3.10/dist-packages/numpy/lib/function_base.py:4824: UserWarning: Warning: 'partition' will ignore the 'mask' of the MaskedArray.
/usr/local/lib/python3.10/dist-packages/numpy/lib/function_base.py:4824: UserWarning: Warning: 'partition' will ignore the 'mask' of the MaskedArray.
  arr.partition(
```

#### Check for missing values

```
print(data.isnull().sum())
    Car Name
                      0
     Year
                      0
     Selling_Price
                      0
     Present Price
     Driven_kms
     Fuel_Type
                      0
     Selling_type
                      0
     Transmission
                      0
     Owner
     dtype: int64
```

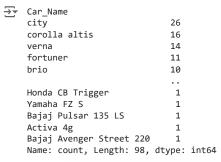
#### Explore relationships between variables

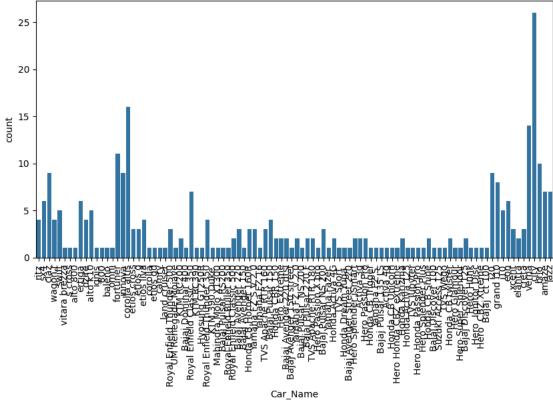
```
sns.pairplot(data[['Selling_Price', 'Present_Price', 'Year']])
plt.show()
```



# Analyze categorical features

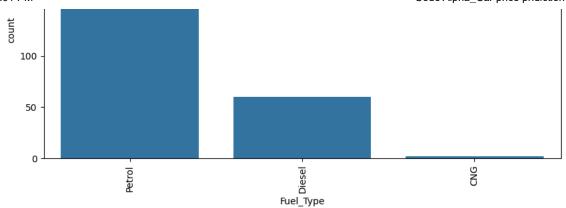
```
for col in ['Car_Name', 'Fuel_Type', 'Selling_type', 'Transmission']:
    print(data[col].value_counts())
    plt.figure(figsize=(10, 5))
    sns.countplot(x=col, data=data)
    plt.xticks(rotation=90)
    plt.show()
```





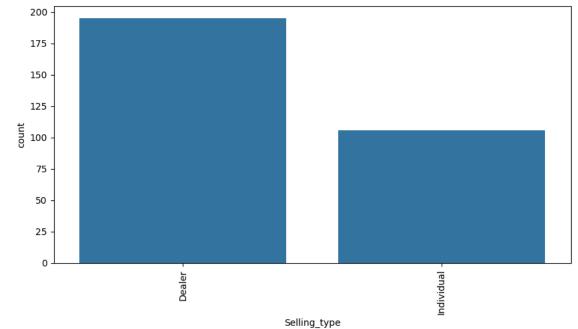
Fuel\_Type
Petrol 239
Diesel 60
CNG 2
Name: count, dtype: int64





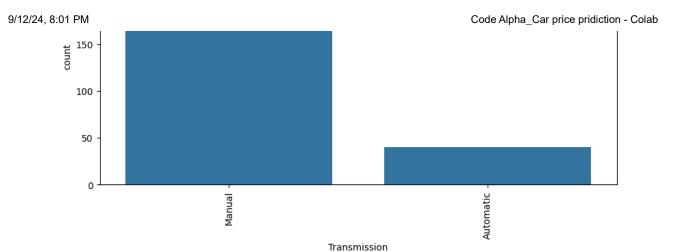
Selling\_type
Dealer 195
Individual 106

Name: count, dtype: int64



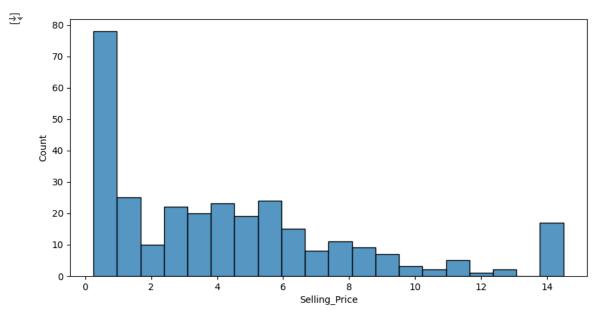
Transmission
Manual 261
Automatic 40
Name: count, dtype: int64

250 -



### Analyze numerical features

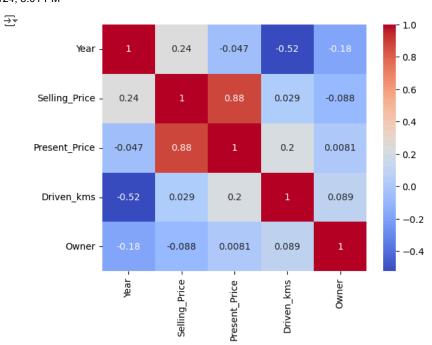
```
plt.figure(figsize=(10, 5))
sns.histplot(data['Selling_Price'], bins=20)
plt.show()
```



#### Correlation matrix

```
import numpy as np

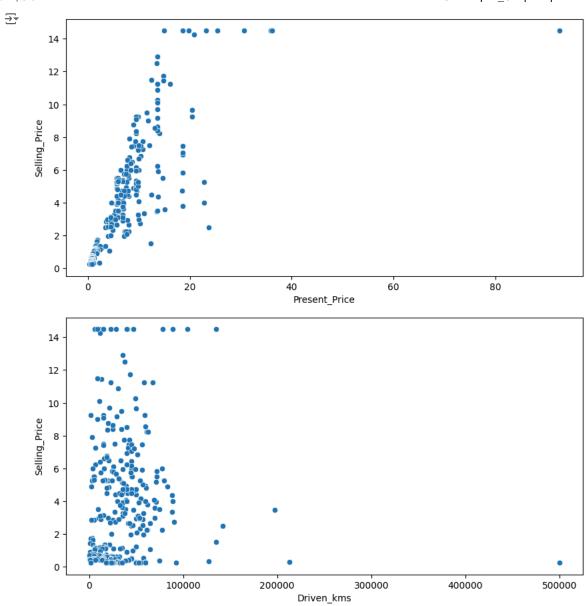
correlation_matrix = data.select_dtypes(include=np.number).corr()
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm')
plt.show()
```



# Analyze the relationship between selling price and other features

```
plt.figure(figsize=(10, 5))
sns.scatterplot(x='Present_Price', y='Selling_Price', data=data)
plt.show()

plt.figure(figsize=(10, 5))
sns.scatterplot(x='Driven_kms', y='Selling_Price', data=data)
plt.show()
```



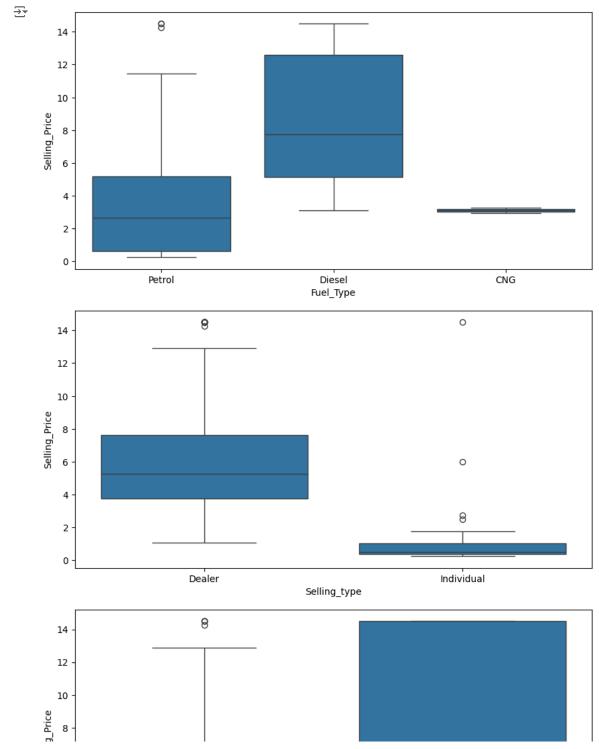
# Analyze the impact of categorical features on selling price

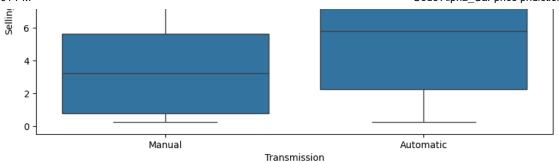
```
plt.figure(figsize=(10, 5))
sns.boxplot(x='Fuel_Type', y='Selling_Price', data=data)
plt.show()

plt.figure(figsize=(10, 5))
https://colab.research.google.com/drive/1JGuQvRUxSPR0Qy2XQXpvDUwkAbSPPGv_#scrollTo=W6P_oKl10Gw1&printMode=true
```

```
plt.show()

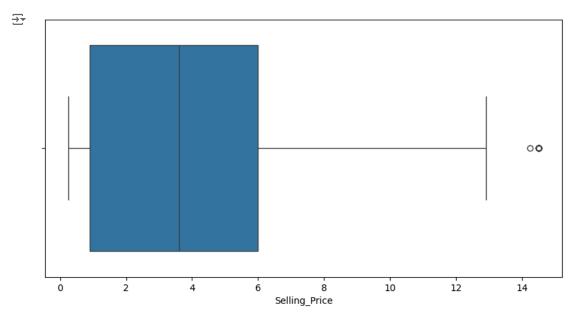
plt.figure(figsize=(10, 5))
sns.boxplot(x='Transmission', y='Selling_Price', data=data)
plt.show()
```





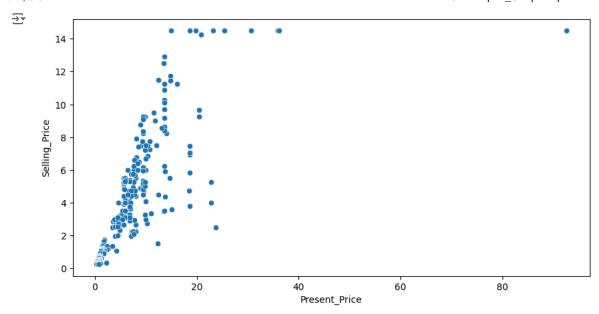
### Identify outliers using box plots

```
plt.figure(figsize=(10, 5))
sns.boxplot(x='Selling_Price', data=data)
plt.show()
```



# Identify outliers using scatter plots

```
plt.figure(figsize=(10, 5))
sns.scatterplot(x='Present_Price', y='Selling_Price', data=data)
plt.show()
```



#### Handle outliers

### Cap outliers

```
# Define upper and lower limits for outliers
upper_limit = data['Selling_Price'].mean() + 3 * data['Selling_Price'].std()
lower_limit = data['Selling_Price'].mean() - 3 * data['Selling_Price'].std()
# Cap outliers
data['Selling_Price'] = data['Selling_Price'].clip(lower=lower_limit, upper=upper_limit)
```

#### Winsorizing

### Feature Engeniering

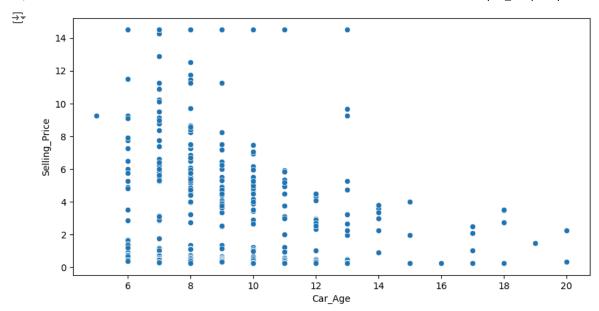
```
# Create a new feature for car age
data['Car_Age'] = 2023 - data['Year']

# Create a feature for price per kilometer driven
data['Price_per_km'] = data['Present_Price'] / data['Driven_kms']

# Create an interaction term between present price and car age
data['Price_Age_Interaction'] = data['Present_Price'] * data['Car_Age']
```

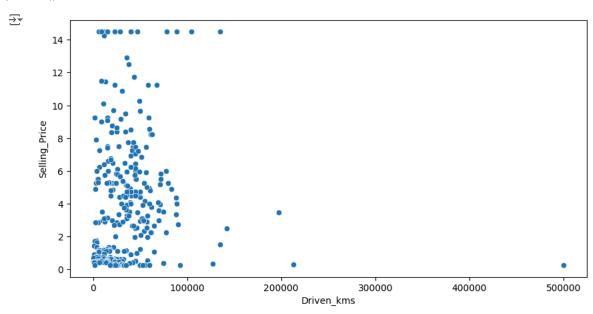
### Analyze the relationship between car age and selling price

```
plt.figure(figsize=(10, 5))
sns.scatterplot(x='Car_Age', y='Selling_Price', data=data)
plt.show()
```



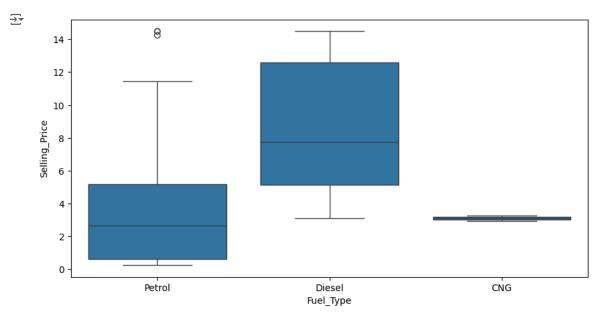
# Analyze the relationship between kilometers driven and selling price

```
plt.figure(figsize=(10, 5))
sns.scatterplot(x='Driven_kms', y='Selling_Price', data=data)
plt.show()
```



# Analyze the impact of fuel type on selling price

```
plt.figure(figsize=(10, 5))
sns.boxplot(x='Fuel_Type', y='Selling_Price', data=data)
plt.show()
```



# Analyze the impact of transmission type on selling price

```
plt.figure(figsize=(10, 5))
sns.boxplot(x='Transmission', y='Selling_Price', data=data)
plt.show()
```