EXPLORATORY DATA ANALYSIS

Step 1: Importing necessary Libraries and the data set and creating dataframe.

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sb
apnadata = pd.read_csv('automobilepricebymileage.csv')
```

Step 2: Obtaining information on dataset.

Step 3: Checking the dimension of the given data set.

```
apnadata.shape
(19, 3)
```

Step 4: Checking size of dataset.

```
apnadata.size
```

Step 5: Understanding the variables in the data set.

```
apnadata.head()
```

	Mileage	Age(yrs)	Sell Price(\$)
0	69000	6	18000
1	35000	3	34000
2	57000	5	26100

Step 6: Checking for null values in dataset.

```
# 40000 4 31500

apnadata.isnull().sum()

Mileage 0
Age(yrs) 0
Sell Price($) 0
dtype: int64
```

Step 7: Dropping duplicate rows and columns if any

```
apnadata.drop_duplicates()
apnadata.shape

(19, 3)
```

Step 8: Slicing-Slicing row indexes from 5 to 10 and column indexes from 0 to 1

```
#apnadata.iloc[row slicing, column slicing]
apnadata.iloc[5:11,0:2]
```

	Mileage	Age(yrs)
5	59000	5
6	52000	5
7	72000	6
8	91000	8
9	67000	6
10	83000	7

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Step 9: Finding number of unique values in dataset

```
apnadata['Age(yrs)'].nunique()
```

Step 10: Finding count of unique values in dataset

```
apnadata['Age(yrs)'].unique()
array([6, 3, 5, 2, 4, 8, 7])
```

Step 11: Using value_counts() method on Mileage and age to identify the count of each category in that column

```
apnadata['Mileage'].value_counts()
     69000
              2
     59000
              2
     67000
              1
     28000
              1
     25400
              1
     82450
              1
              1
     58780
     79000
              1
     83000
              1
     91000
              1
     35000
              1
     72000
              1
     52000
              1
     46000
             1
     22500
              1
     57000
              1
     87600
              1
     Name: Mileage, dtype: int64
apnadata['Age(yrs)'].value_counts()
     5
          5
     6
          3
     7
          3
     3
          2
     2
          2
     4
          2
          2
```

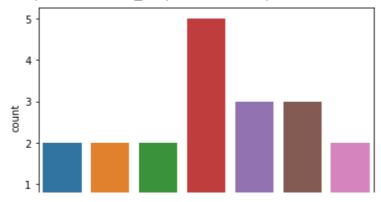
VISUALIZATIONS

Number of automobiles for each corresponding age

Name: Age(yrs), dtype: int64

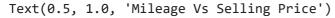
```
sb.countplot(x='Age(yrs)',data=apnadata)
```

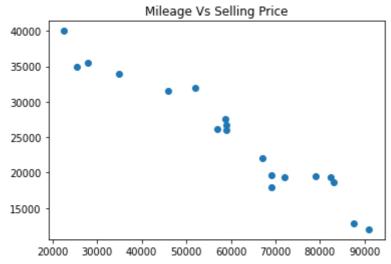
<matplotlib.axes._subplots.AxesSubplot at 0x7fe16486e610>



Mileage of automobile Vs its Selling price

plt.scatter(apnadata['Mileage'],apnadata['Sell Price(\$)'])
plt.title("Mileage Vs Selling Price")





Age of automobile Vs Selling Price

plt.scatter(apnadata['Age(yrs)'],apnadata['Sell Price(\$)'])
plt.title("Age of automobile Vs Selling Price")

```
Text(0.5, 1.0, 'Age of automobile Vs Selling Price')
```

Looking at above two scatter plots, using linear regression model makes sense as we can clearly see a linear relationship between our dependant (i.e. Selling Price) and independant variables (i.e. age and mileage)

The approach we are going to use here is to split available data in two sets

- 1. Training: We will train our model on this dataset
- 2. Testing: We will use this subset to make actual predictions using trained model

APPLYING REGRESSION

Step 1: Splitting the data set into input and output.

```
x = apnadata[['Mileage','Age(yrs)']]
y = apnadata['Sell Price($)']
```

Step 2: Splitting the data set into training and testing data set.

```
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.3)
x_train
```

```
Milazga Agalure)
x_test
          Mileage Age(yrs)
       1
                           3
            35000
      13
            58780
                           4
      0
            69000
                           6
       3
            22500
                           2
      2
            57000
                           5
                           5
      12
            59000
      11
            1 9000
y_train
     8
           12000
     17
           19700
     18
           12800
     10
           18700
     4
           31500
     7
           19300
     5
           26750
     6
           32000
     11
           19500
     16
           35500
     9
           22000
     15
           35000
     14
           19400
     Name: Sell Price($), dtype: int64
y_test
           34000
     1
     13
           27500
     0
           18000
     3
           40000
     2
           26100
     12
           26000
     Name: Sell Price($), dtype: int64
Step 3: Applying a regressor model on training set.
from sklearn.linear_model import LinearRegression
model = LinearRegression()
```

Step 4: To Fit the model(Mapping/Plotting of Inputs with the Outputs)

```
model.fit(x_train, y_train)
```

LinearRegression()

Step 5: Predicting the output.

model.predict(x_test)

array([33900.28281421, 25013.44107011, 21890.08696084, 38361.42248562, 26155.05189498, 25370.35294593])

Step 6: Calculating the accuracy of the model

model.score(x_test,y_test)

0.9145603157635419

Therefore, the accuracy of our model is 91%.

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