kaviyadevi 20106064

In [6]:
```
#to import libraries
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

In [4]:
```
#to import dataset
data=pd.read_csv(r"C:\Users\user\Downloads\14_Iris - 14_Iris.csv")
data
```

Out[4]:

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa
...	...	...	...	...	...	...
145	146	6.7	3.0	5.2	2.3	Iris-virginica
146	147	6.3	2.5	5.0	1.9	Iris-virginica
147	148	6.5	3.0	5.2	2.0	Iris-virginica
148	149	6.2	3.4	5.4	2.3	Iris-virginica
149	150	5.9	3.0	5.1	1.8	Iris-virginica

150 rows × 6 columns

In [7]:
```
#to display top 5 rows
data.head()
```

Out[7]:

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

DATA CLEANING AND PREPROCESSING

In [8]:
```
#
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):
 #   Column         Non-Null Count  Dtype
---  ------         --------------  -----
 0   Id             150 non-null    int64
 1   SepalLengthCm  150 non-null    float64
 2   SepalWidthCm   150 non-null    float64
 3   PetalLengthCm  150 non-null    float64
 4   PetalWidthCm   150 non-null    float64
 5   Species        150 non-null    object
dtypes: float64(4), int64(1), object(1)
memory usage: 7.2+ KB
```

In [9]:
```
#to display summary of statistics(here to know min max value)
data.describe()
```

Out[9]:

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	150.000000	150.000000	150.000000	150.000000	150.000000
mean	75.500000	5.843333	3.054000	3.758667	1.198667
std	43.445368	0.828066	0.433594	1.764420	0.763161
min	1.000000	4.300000	2.000000	1.000000	0.100000
25%	38.250000	5.100000	2.800000	1.600000	0.300000
50%	75.500000	5.800000	3.000000	4.350000	1.300000
75%	112.750000	6.400000	3.300000	5.100000	1.800000
max	150.000000	7.900000	4.400000	6.900000	2.500000

In [10]:
```
#to display the column heading
data.columns
```

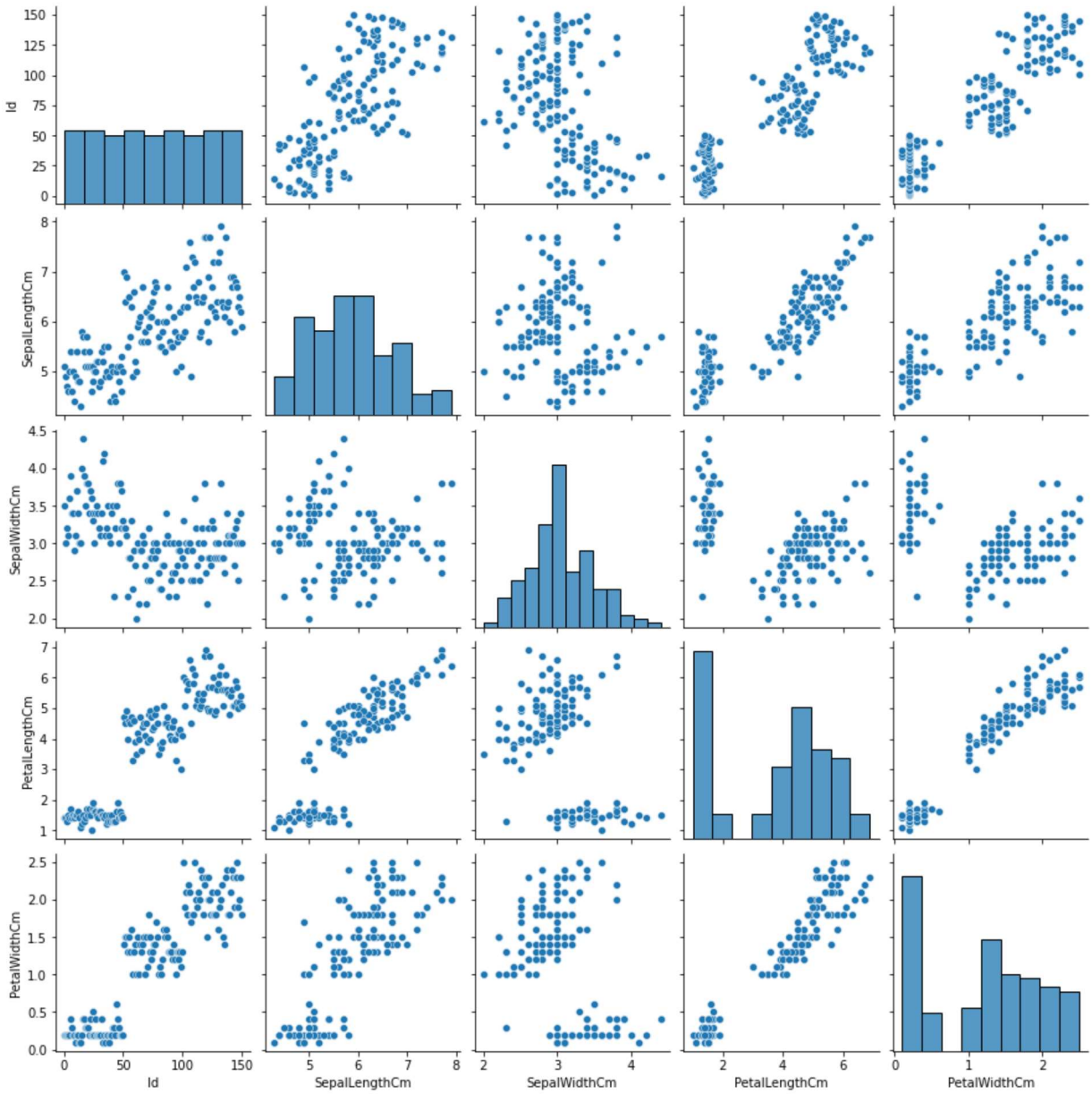
Out[10]:
```
Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',
       'Species'],
      dtype='object')
```

In [11]:
```
#here there is no missing values (identified through info() 5000 data are describ
```

EDA and DATA VISUALIZATION

In [12]:
```
sns.pairplot(data)
```

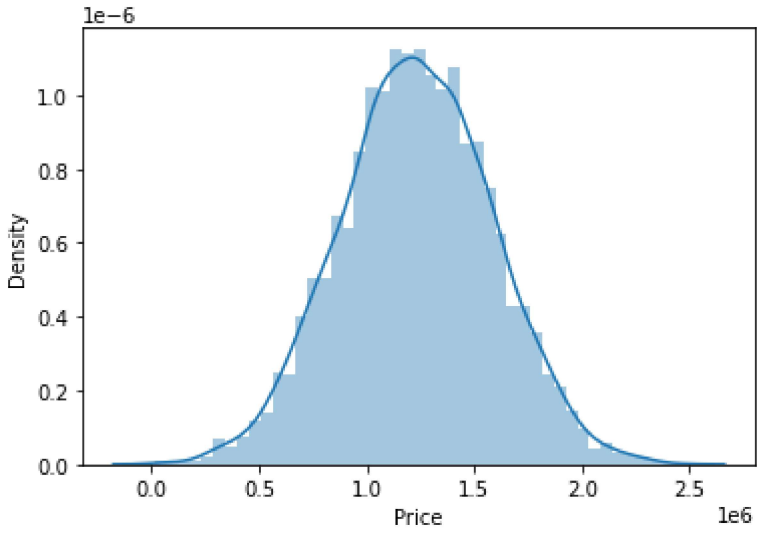
Out[12]:
```
<seaborn.axisgrid.PairGrid at 0x2fbc4f3f2b0>
```



In [41]:
```
sns.distplot(data['PetalWidthCm'])
```

```
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: Futur
eWarning: `distplot` is a deprecated function and will be removed in a future v
ersion. Please adapt your code to use either `displot` (a figure-level function
with similar flexibility) or `histplot` (an axes-level function for histogram
s).
  warnings.warn(msg, FutureWarning)
```

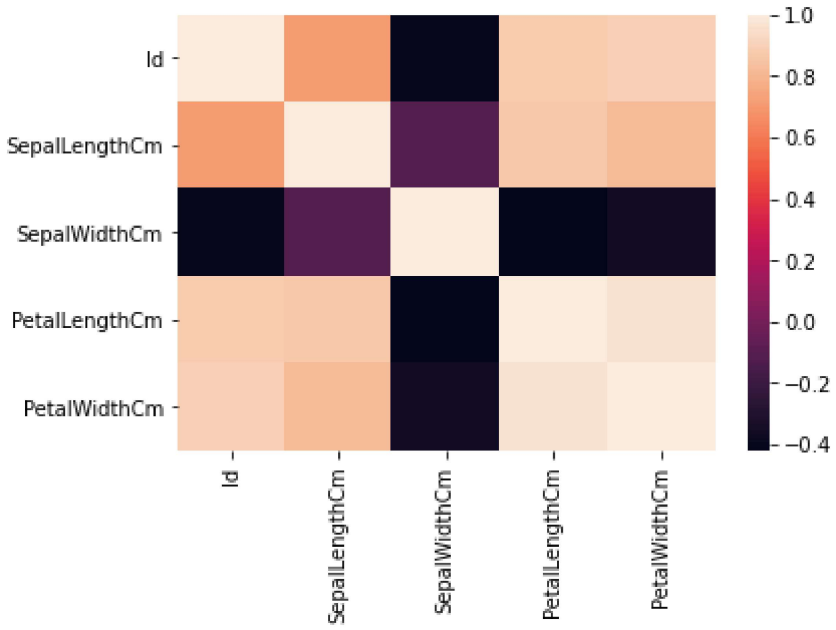
Out[41]: <AxesSubplot:xlabel='Price', ylabel='Density'>



In [13]:
```
df=data[['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',
       'Species']]
```

In [14]:
```
sns.heatmap(df.corr())
```

Out[14]: <AxesSubplot:>



TRAINING MODEL

In [18]:
```
x=df[['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm']]
y=df['PetalWidthCm']
```

In [19]:
```
#to split my dataset into trainning and test

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)
```

In [20]:
```
from sklearn.linear_model import LinearRegression

lr=LinearRegression()
lr.fit(x_train,y_train)
```

Out[20]:
```
LinearRegression()
```

In [21]:
```
#to find intercept
print(lr.intercept_)
```

```
-0.8001697979277445
```

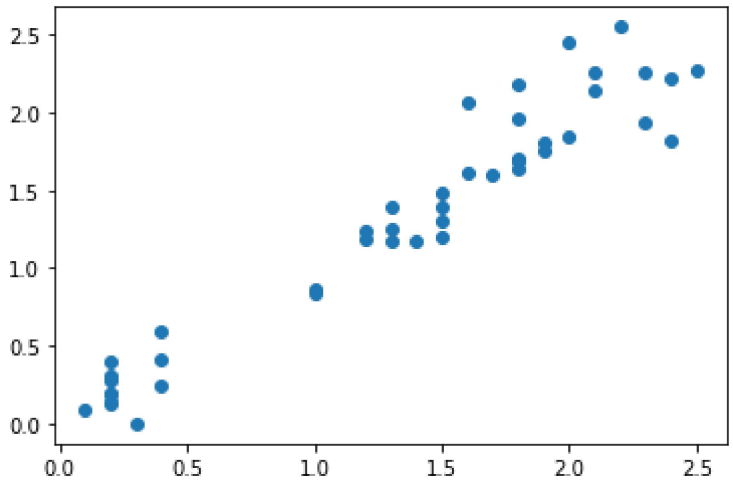
In [22]:
```
coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

Out[22]:

	Co-efficient
Id	0.003633
SepalLengthCm	-0.113643
SepalWidthCm	0.268687
PetalLengthCm	0.414671

In [23]:
```
prediction = lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[23]:
```
<matplotlib.collections.PathCollection at 0x2fbc79cc190>
```



In [24]:
```
print(lr.score(x_test,y_test))
```

```
0.9270500245152584
```

RIDGE AND LASSO REGRESSION

In [25]:
```
from sklearn.linear_model import Ridge,Lasso
```

In [26]:
```
rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
```

Out[26]:
```
Ridge(alpha=10)
```

In [27]:
```
rr.score(x_test,y_test)
```

Out[27]:
```
0.919423436518253
```

In [28]:
```
la=Lasso(alpha=10)
la.fit(x_train,y_train)
```

Out[28]:
```
Lasso(alpha=10)
```

In [29]:
```
la.score(x_test,y_test)
```

Out[29]:
```
0.6607429271533307
```