# IRIS DATASET ANALYSIS

#### **Dataset Information**

The dataset contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linerly seperable from the other two.

Attribute Information: 1.sepal length in cm 2.sepal width in cm 3.petal length in cm 4.petal width in cm 5.classes are:- Iris setosa ;Iris versicolor ;Iris virginica

#### **Import Modules**

```
In [1]:

import pandas as pd
import numpy as np
import os
import seaborn as sns
import matplotlib.pyplot as plt

In [2]:
```

# In [3]:

```
#shape of dataset.
data.shape
```

# Out[3]:

(150, 6)

# In [4]:

#description about stats of dataset.
data.describe()

data = pd.read\_csv("C:/Users/priya/OneDrive/Desktop/Iris.csv")

# Out[4]:

	ld	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 [5]:
```

```
#heads of dataset.
data.head(150)
```

#### Out[5]:

	ld	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 [6]:

```
#tail of dataset.
data.tail()
```

#### Out[6]:

	ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
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

# In [7]:

```
#columns of dataset.
data.columns
```

### Out[7]:

#### In [8]

```
#basic information of dataset.
data.info()
```

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

#### In [9]:

```
Iris_species = data.groupby('Species')
```

# In [10]:

```
#description of stats among the each species class.
Iris_species.describe().T
```

# Out[10]:

	Species	Iris-setosa	Iris-versicolor	Iris-virginica
ld	count	50.000000	50.000000	50.000000
	mean	25.500000	75.500000	125.500000
	std	14.577380	14.577380	14.577380
	min	1.000000	51.000000	101.000000
	25%	13.250000	63.250000	113.250000
	50%	25.500000	75.500000	125.500000
	75%	37.750000	87.750000	137.750000
	max	50.000000	100.000000	150.000000
SepalLengthCm	count	50.000000	50.000000	50.000000
	mean	5.006000	5.936000	6.588000
	std	0.352490	0.516171	0.635880
	min	4.300000	4.900000	4.900000
	25%	4.800000	5.600000	6.225000
	50%	5.000000	5.900000	6.500000
	75%	5.200000	6.300000	6.900000
	max	5.800000	7.000000	7.900000
SepalWidthCm	count	50.000000	50.000000	50.000000
	mean	3.418000	2.770000	2.974000
	std	0.381024	0.313798	0.322497
	min	2.300000	2.000000	2.200000
	25%	3.125000	2.525000	2.800000
	50%	3.400000	2.800000	3.000000
	75%	3.675000	3.000000	3.175000
	max	4.400000	3.400000	3.800000
PetalLengthCm	count	50.000000	50.000000	50.000000
	mean	1.464000	4.260000	5.552000
	std	0.173511	0.469911	0.551895
	min	1.000000	3.000000	4.500000
	25%	1.400000	4.000000	5.100000
	50%	1.500000	4.350000	5.550000
	75%	1.575000	4.600000	5.875000
	max	1.900000	5.100000	6.900000
PetalWidthCm	count	50.000000	50.000000	50.000000
	mean	0.244000	1.326000	2.026000
	std	0.107210	0.197753	0.274650
	min	0.100000	1.000000	1.400000
	25%	0.200000	1.200000	1.800000
	50%	0.200000	1.300000	2.000000
	75%	0.300000	1.500000	2.300000
	max	0.600000	1.800000	2.500000

# In [11]:

```
#check for null values.
data.isnull().sum()
```

# Out[11]:

Id 0
SepalLengthCm 0
SepalWidthCm 0
PetalLengthCm 0
PetalWidthCm 0
Species 0
dtype: int64

# **Exploratory Data Analysis**

Analysis of Dataset using Graph plotting

# List of Graphs plot:-

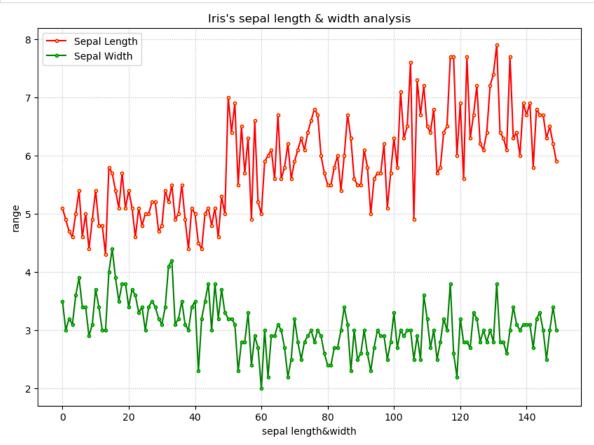
- 1. Line Graph
- 2. Histogram
- 3. Bar Graph
- 4. Scatter Plot
- 5. Pie Chart
- 6. Box Graph

# **LINE Graph**

#### Line Graph of Sepal Length & Width

# In [12]:

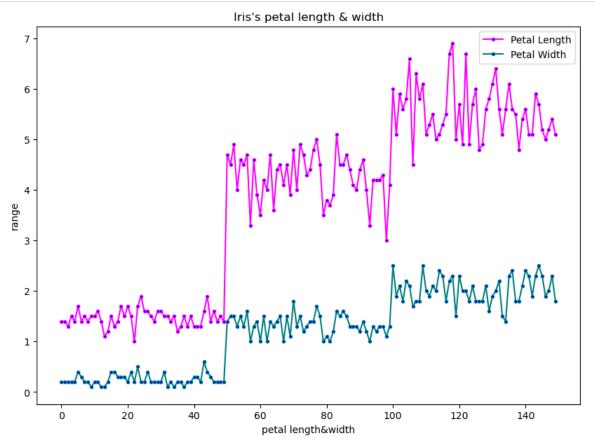
```
from matplotlib import rcParams
rcParams['figure.figsize'] = 10,7
plt.plot(data.SepalLengthCm, color="red",marker='.',markerfacecolor='yellow',label='Sepal Length')
plt.plot(data.SepalWidthCm, color="green",marker='.',markerfacecolor='lime',label='Sepal Width')
plt.title("Iris's sepal length & width analysis")
plt.xlabel("sepal length&width")
plt.ylabel("range")
plt.grid(True, color='silver',linestyle=":")
plt.legend()
plt.show()
```



#### Line Graph of Petal Length & Width

#### In [13]:

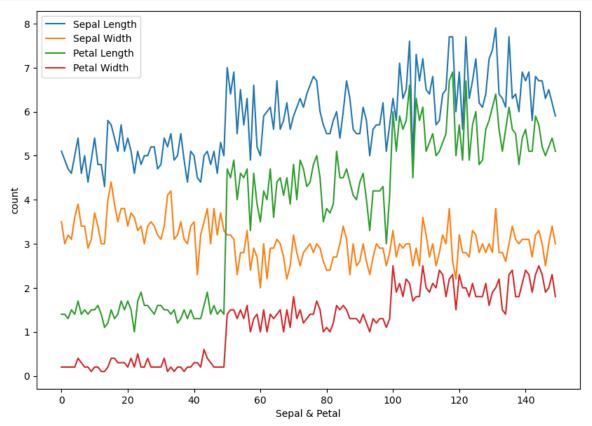
```
from matplotlib import rcParams
rcParams['figure.figsize']=10,7
plt.plot(data.PetalLengthCm,color="magenta",marker=".",markerfacecolor="blue",label='Petal Length')
plt.plot(data.PetalWidthCm,color="teal",marker=".",markerfacecolor="blue",label='Petal Width')
plt.title("Iris's petal length & width")
plt.xlabel('petal length&width')
plt.ylabel('range')
plt.legend()
plt.show()
```



Line Graph of Sepal & Petal

#### In [14]:

```
plt.plot(data.SepalLengthCm,label='Sepal Length')
plt.plot(data.SepalWidthCm,label='Sepal Width')
plt.plot(data.PetalLengthCm,label='Petal Length')
plt.plot(data.PetalWidthCm,label='Petal Width')
plt.xlabel('Sepal & Petal')
plt.ylabel('count')
plt.legend();
```

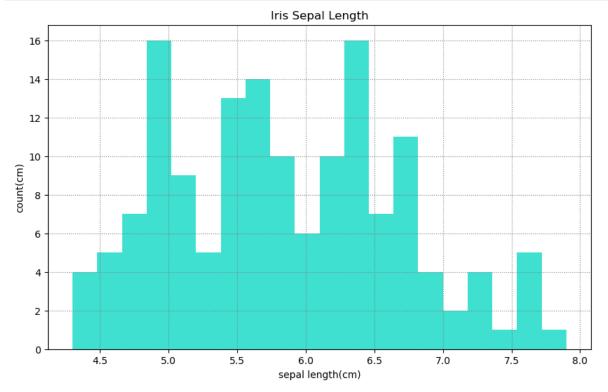


# **HISTOGRAM**

Histogram graph for Sepal Length

# In [15]:

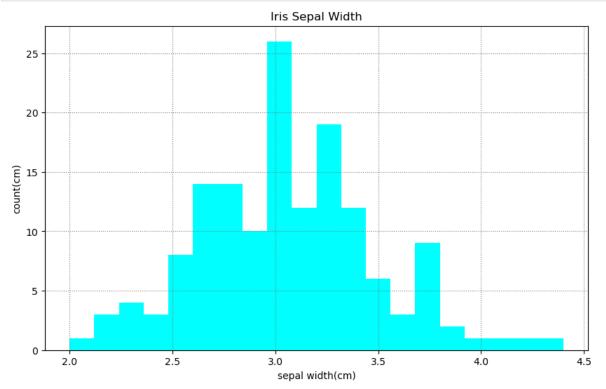
```
plt.figure(figsize=(10,6))
plt.hist(data.SepalLengthCm,color="turquoise",bins=20)
plt.title('Iris Sepal Length')
plt.xlabel('sepal length(cm)')
plt.ylabel('count(cm)')
plt.ylabel('rount(cm)')
plt.grid(True,color='grey',linestyle=':');
```



Histogram Graph for Sepal Width

# In [16]:

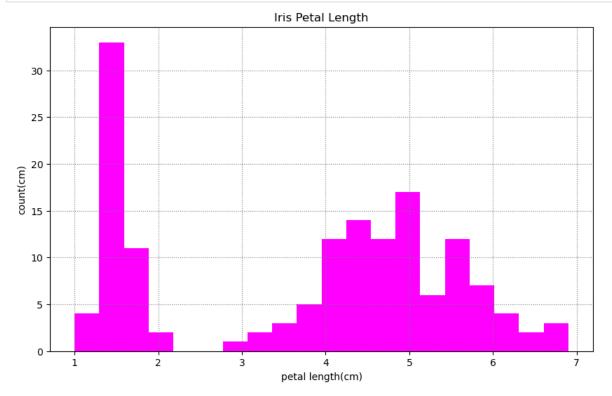
```
plt.figure(figsize=(10,6))
plt.hist(data.SepalWidthCm,color="aqua",bins=20)
plt.title('Iris Sepal Width')
plt.xlabel('sepal width(cm)')
plt.ylabel('count(cm)')
plt.grid(True,color='grey',linestyle=':');
```



Histogram Graph for Petal Length

# In [17]:

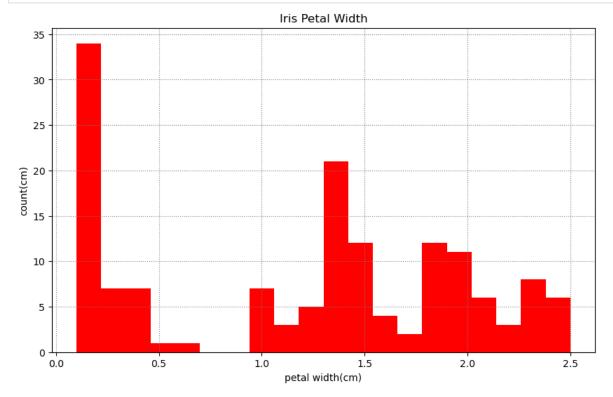
```
plt.figure(figsize=(10,6))
plt.hist(data.PetalLengthCm,color="magenta",bins=20)
plt.title('Iris Petal Length')
plt.xlabel('petal length(cm)')
plt.ylabel('count(cm)')
plt.grid(True,color='grey',linestyle=':');
```



Histogram Graph for Petal Width

# In [18]:

```
plt.figure(figsize=(10,6))
plt.hist(data.PetalWidthCm,color="red",bins=20)
plt.title('Iris Petal Width')
plt.xlabel('petal width(cm)')
plt.ylabel('count(cm)')
plt.grid(True,color='grey',linestyle=':');
```

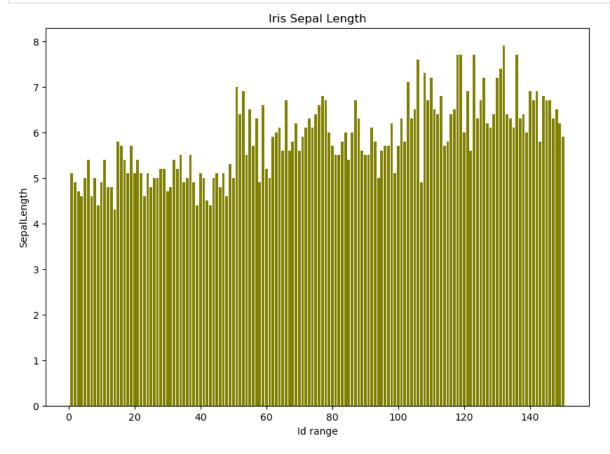


# **BAR GRAPH**

Bar Graph for Sepal Length

```
In [19]:
```

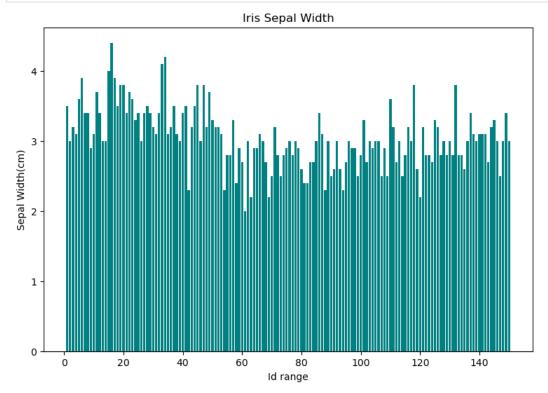
```
x=data['Id']
y=data['SepalLengthCm']
plt.bar(x,y,color='olive')
plt.xlabel("Id range")
plt.ylabel("SepalLength")
plt.title("Iris Sepal Length");
```



Bar Graph for Sepal width

```
In [20]:
```

```
plt.figure(figsize=(9,6))
plt.bar(data.Id,data.SepalWidthCm,color='teal')
plt.title('Iris Sepal Width')
plt.ylabel('Sepal Width(cm)')
plt.xlabel('Id range');
```

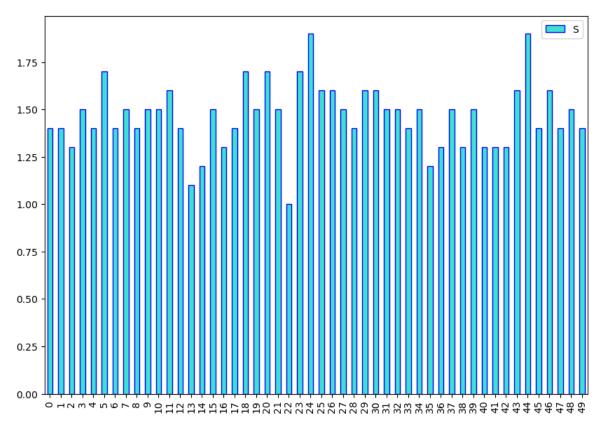


Bar Graph for Petal Length

# In [21]:

```
m=list(data['PetalLengthCm'])
n=m[0:50]
n=pd.DataFrame(n)
print(type(n))
n.plot(kind='bar',color='turquoise',edgecolor='blue')
plt.legend("Species")
plt.show();
```

<class 'pandas.core.frame.DataFrame'>

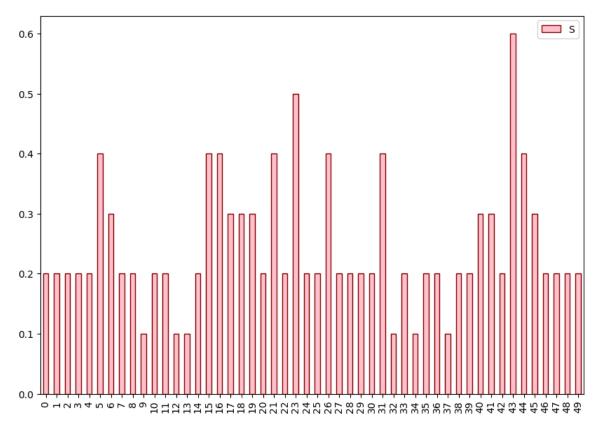


Bar Graph for Petal Width

```
In [22]:
```

```
m=list(data['PetalWidthCm'])
n=m[0:50]
n=pd.DataFrame(n)
print(type(n))
n.plot(kind='bar',color='pink',edgecolor='maroon')
plt.legend("Species")
plt.show();
```

<class 'pandas.core.frame.DataFrame'>

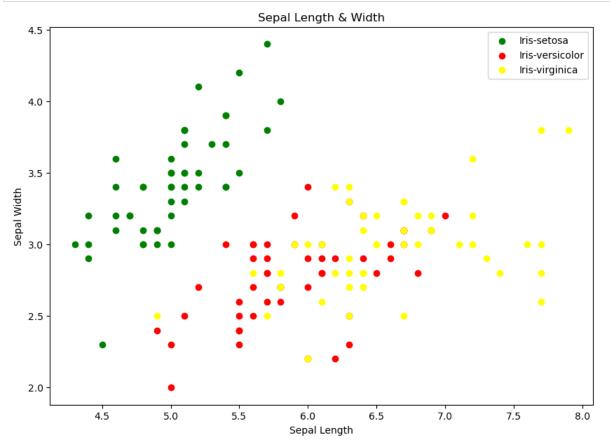


# **SCATTER PLOT**

ScatterPlot for Sepal Length & Width

#### In [23]:

```
colors = ['green', 'red', 'yellow']
species = ['Iris-setosa', 'Iris-versicolor', 'Iris-virginica']
for i in range(3):
    x = data[data['Species'] == species[i]]
    plt.scatter(x['SepalLengthCm'], x['SepalWidthCm'], c= colors[i], label = species[i])
    plt.title('Sepal Length & Width')
    plt.xlabel("Sepal Length")
    plt.ylabel("Sepal Width")
    plt.legend();
```

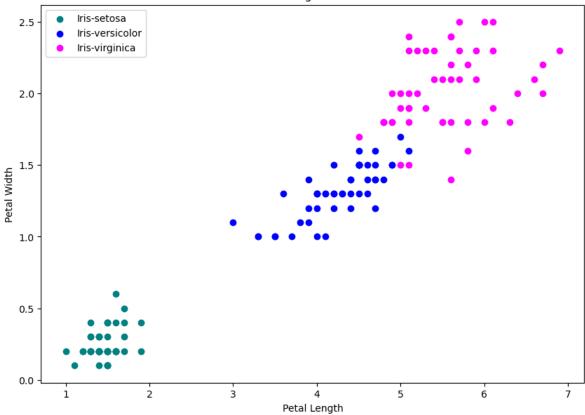


Scatter Plot for Petal Length & Width

#### In [24]:

```
colors = ['teal', 'blue', 'magenta']
species = ['Iris-setosa', 'Iris-versicolor', 'Iris-virginica']
for i in range(3):
    x = data[data['Species'] == species[i]]
    plt.scatter(x['PetalLengthCm'], x['PetalWidthCm'], c= colors[i], label = species[i])
    plt.title('Petal Length & Width')
    plt.xlabel("Petal Length")
    plt.ylabel("Petal Width")
    plt.legend();
```





# **PIE CHART**

# Pie Chart for Iris Species

```
In [25]:
```

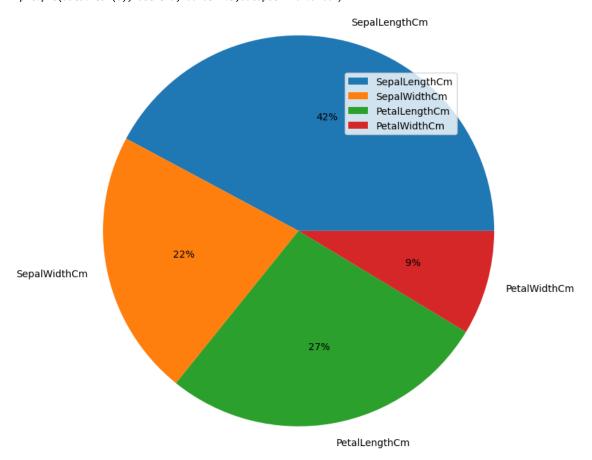
```
#deleting a column from dataset
data = data.drop(columns=['Id'])
```

#### In [26]:

```
a = list(data.columns)
a = a[0:4]
plt.figure(figsize= (6,8))
plt.pie(data.mean(0),labels=a,radius=1.5,autopct ="%2.0f%%")
plt.legend();
```

C:\Users\priya\AppData\Local\Temp\ipykernel\_4468\2211255378.py:4: FutureWarning: Dropping of nuisance columns in Data Frame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.

plt.pie(data.mean(0),labels=a,radius=1.5,autopct ="%2.0f%%")

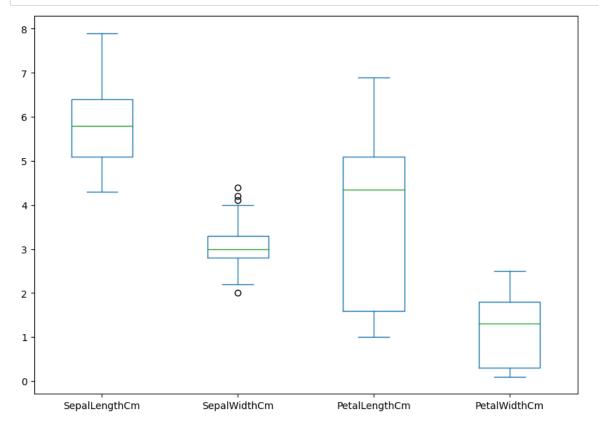


# **BOX PLOT**

Box Plot for Iris Species

In [27]:

```
data = pd.DataFrame(data)
data.plot(kind='box', vert=True);
```



# **Correlation Matrix**

The correlation matrix display correlation coefficients for different variables. The value is in range of -1 to 1.

```
In [28]:
```

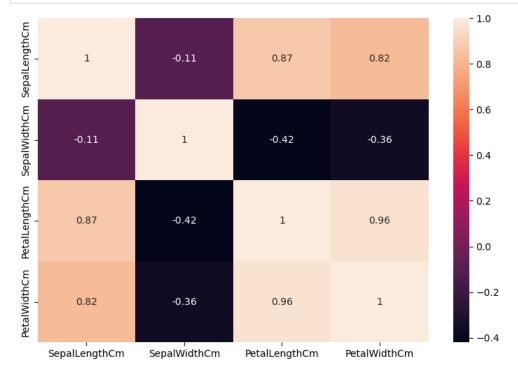
data.corr()

Out[28]:

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
SepalLengthCm	1.000000	-0.109369	0.871754	0.817954
SepalWidthCm	-0.109369	1.000000	-0.420516	-0.356544
PetalLengthCm	0.871754	-0.420516	1.000000	0.962757
PetalWidthCm	0.817954	-0.356544	0.962757	1.000000

# In [29]:

```
corr=data.corr()
plt.subplots(figsize=(9,6))
sns.heatmap(corr,annot=True);
```



# THANKS:)