

Iris Flower Classification

```
In [1]:
```

```
1 import numpy as np
2 import pandas as pd
3 import seaborn as sns
4 import matplotlib.pyplot as plt
5 from sklearn import svm, datasets
```

```
In [64]:
```

```
1 pwd
```

```
Out[64]: 'C:\\\\Users\\\\Shivv'
```

```
In [2]:
```

```
1 df = pd.read_csv('C:\\\\Users\\\\Shivv\\\\iris.data')
```

```
In [4]:
```

```
1 df.set_axis(['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm', 'Species'],
2             axis = 'columns', inplace = True)
3
```

```
In [67]:
```

```
1 df.head()
```

```
Out[67]:
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	4.9	3.0	1.4	0.2	Iris-setosa
1	4.7	3.2	1.3	0.2	Iris-setosa
2	4.6	3.1	1.5	0.2	Iris-setosa
3	5.0	3.6	1.4	0.2	Iris-setosa
4	5.4	3.9	1.7	0.4	Iris-setosa

```
In [68]: 1 df.tail()
```

```
Out[68]:
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
144	6.7	3.0	5.2	2.3	Iris-virginica
145	6.3	2.5	5.0	1.9	Iris-virginica
146	6.5	3.0	5.2	2.0	Iris-virginica
147	6.2	3.4	5.4	2.3	Iris-virginica
148	5.9	3.0	5.1	1.8	Iris-virginica

```
In [69]: 1 df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 149 entries, 0 to 148
Data columns (total 5 columns):
 #   Column      Non-Null Count  Dtype  
--- 
 0   SepalLengthCm  149 non-null   float64
 1   SepalWidthCm   149 non-null   float64
 2   PetalLengthCm  149 non-null   float64
 3   PetalWidthCm   149 non-null   float64
 4   Species        149 non-null   object  
dtypes: float64(4), object(1)
memory usage: 5.9+ KB
```

```
In [70]: 1 df.nunique()
```

```
Out[70]:
```

SepalLengthCm	35
SepalWidthCm	23
PetalLengthCm	43
PetalWidthCm	22
Species	3
	dtype: int64

```
In [71]: 1 df.describe()
```

```
Out[71]:
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	149.000000	149.000000	149.000000	149.000000
mean	5.848322	3.051007	3.774497	1.205369
std	0.828594	0.433499	1.759651	0.761292
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.400000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

```
In [72]: 1 df.keys
```

```
Out[72]:
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	4.9	3.0	1.4	0.2	Iris-setosa
1	4.7	3.2	1.3	0.2	Iris-setosa
2	4.6	3.1	1.5	0.2	Iris-setosa
3	5.0	3.6	1.4	0.2	Iris-setosa
4	5.4	3.9	1.7	0.4	Iris-setosa
..
144	6.7	3.0	5.2	2.3	Iris-virginica
145	6.3	2.5	5.0	1.9	Iris-virginica
146	6.5	3.0	5.2	2.0	Iris-virginica
147	6.2	3.4	5.4	2.3	Iris-virginica
148	5.9	3.0	5.1	1.8	Iris-virginica

[149 rows x 5 columns]>

```
In [73]: 1 df.values
```

```
Out[73]: array([[4.9, 3.0, 1.4, 0.2, 'Iris-setosa'],
       [4.7, 3.2, 1.3, 0.2, 'Iris-setosa'],
       [4.6, 3.1, 1.5, 0.2, 'Iris-setosa'],
       [5.0, 3.6, 1.4, 0.2, 'Iris-setosa'],
       [5.4, 3.9, 1.7, 0.4, 'Iris-setosa'],
       [4.6, 3.4, 1.4, 0.3, 'Iris-setosa'],
       [5.0, 3.4, 1.5, 0.2, 'Iris-setosa'],
       [4.4, 2.9, 1.4, 0.2, 'Iris-setosa'],
       [4.9, 3.1, 1.5, 0.1, 'Iris-setosa'],
       [5.4, 3.7, 1.5, 0.2, 'Iris-setosa'],
       [4.8, 3.4, 1.6, 0.2, 'Iris-setosa'],
       [4.8, 3.0, 1.4, 0.1, 'Iris-setosa'],
       [4.3, 3.0, 1.1, 0.1, 'Iris-setosa'],
       [5.8, 4.0, 1.2, 0.2, 'Iris-setosa'],
       [5.7, 4.4, 1.5, 0.4, 'Iris-setosa'],
       [5.4, 3.9, 1.3, 0.4, 'Iris-setosa'],
       [5.1, 3.5, 1.4, 0.3, 'Iris-setosa'],
       [5.7, 3.8, 1.7, 0.3, 'Iris-setosa'],
       [5.1, 3.8, 1.5, 0.3, 'Iris-setosa'],
       ...])
```

```
In [74]: 1 df.shape
```

```
Out[74]: (149, 5)
```

```
In [75]: 1 df.isna()
```

```
Out[75]:
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False
...
144	False	False	False	False	False
145	False	False	False	False	False
146	False	False	False	False	False
147	False	False	False	False	False
148	False	False	False	False	False

149 rows × 5 columns

```
In [76]: 1 df.isna().sum()
```

```
Out[76]:
```

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

```
In [77]: 1 df['Species'].value_counts()
```

```
Out[77]:
```

Iris-versicolor	50
Iris-virginica	50
Iris-setosa	49
Name: Species, dtype: int64	

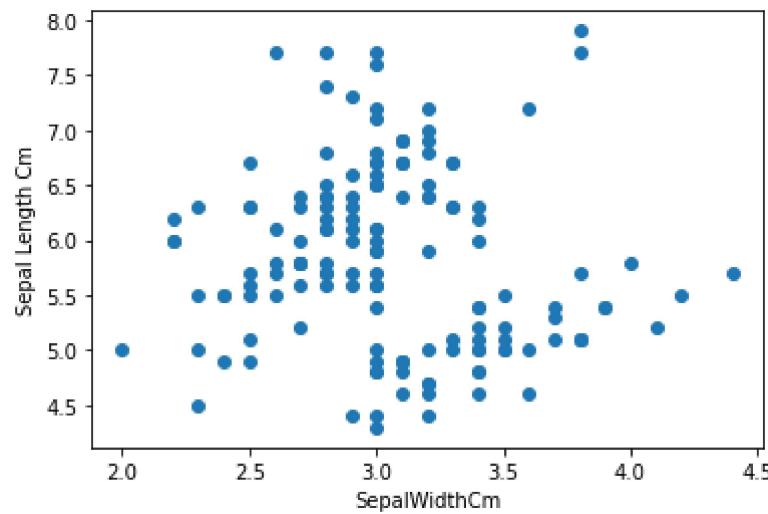
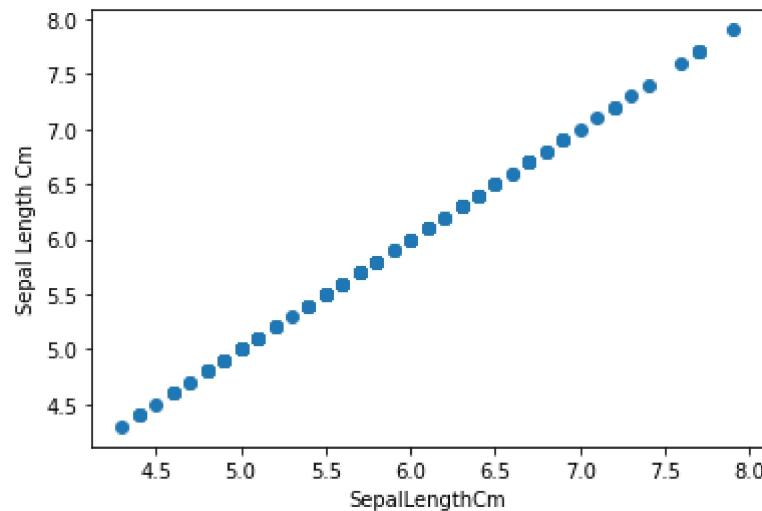
```
In [78]: 1 df.columns
```

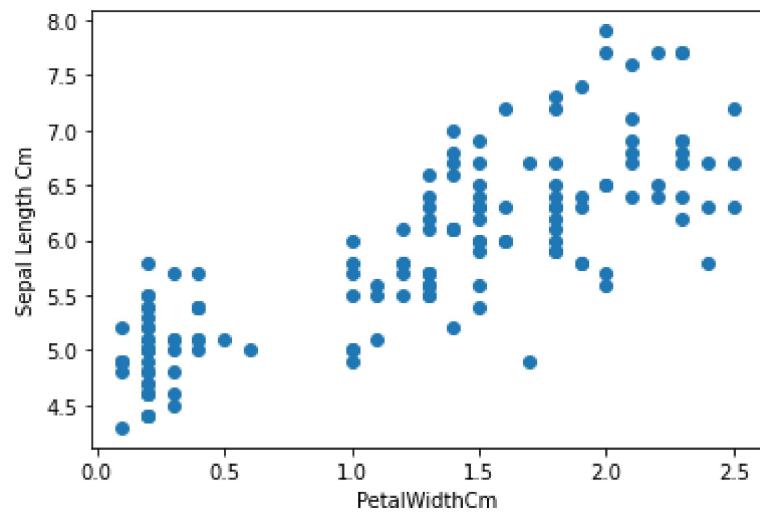
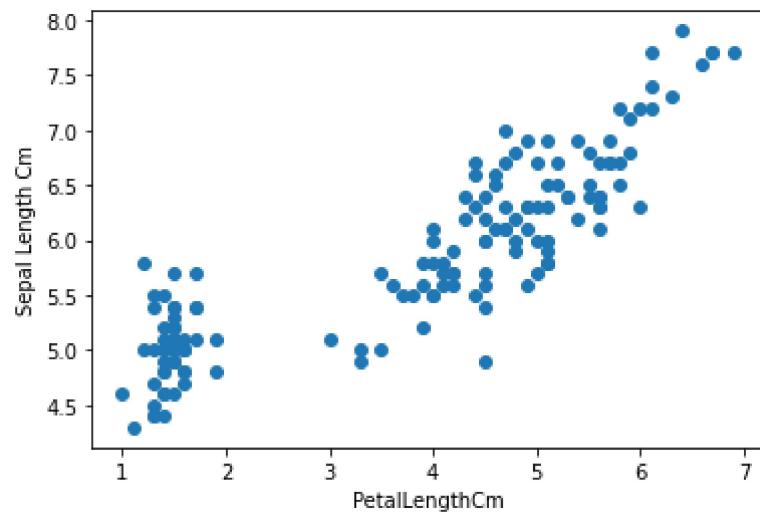
```
Out[78]: Index(['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',
   'Species'],
   dtype='object')
```

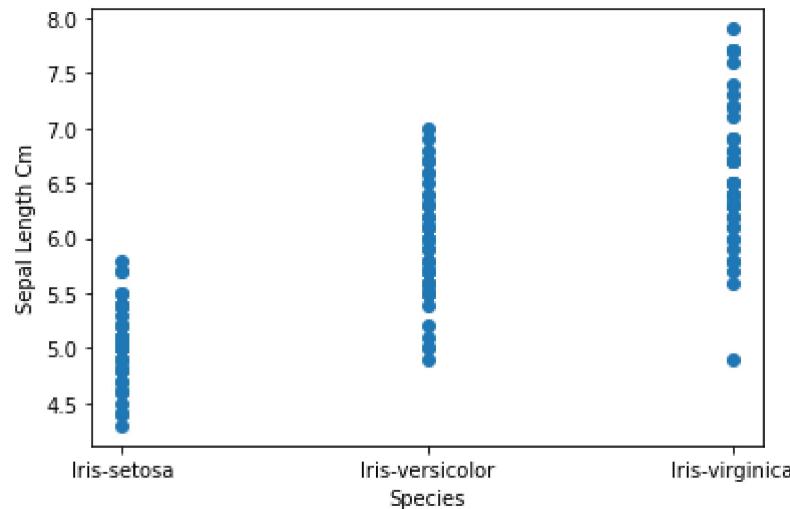
Scatter Plot

In [60]:

```
1 for i in df.columns:  
2     plt.scatter(df[str(i)],df[ 'SepalLengthCm' ])  
3     plt.xlabel(i)  
4     plt.ylabel("Sepal Length Cm")  
5     plt.show()
```



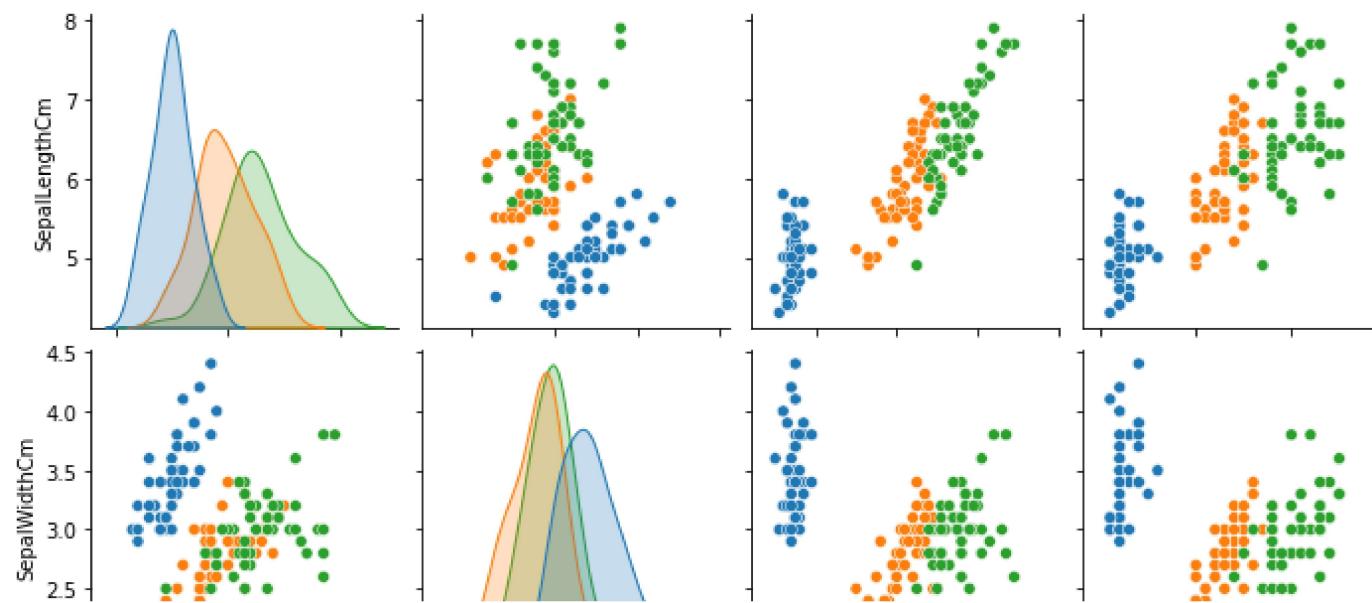




Pair Plot

In [5]: 1 sns.pairplot(df, hue = 'Species')

Out[5]: <seaborn.axisgrid.PairGrid at 0x12988d38070>



Type Markdown and LaTeX: α^2

In [25]: 1 df.corr()

Out[25]:

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
SepalLengthCm	1.000000	-0.103784	0.871283	0.816971
SepalWidthCm	-0.103784	1.000000	-0.415218	-0.350733
PetalLengthCm	0.871283	-0.415218	1.000000	0.962314
PetalWidthCm	0.816971	-0.350733	0.962314	1.000000

In [26]: 1 df.columns

Out[26]: Index(['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm', 'Species'],
dtype='object')

In [59]:

```
1 # col = ['SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm']
2 # plt.figure(figsize= (35,5))
3 # i = 1
4 # for j in col:
5 #     plt.subplot(1,10,i)
6 #     sns.distplot(df[j])
7 #     i = i+1
8 #     plt.tight_layout()
9 #     plt.show()
10
```

In [44]:

```
1 # plt.figure(figsize=(10,10))
2 # sns.heatmap(df.corr(), cmap = 'Blue', annot = True)
```

```
In [32]: 1 df.describe()
```

```
Out[32]:
```

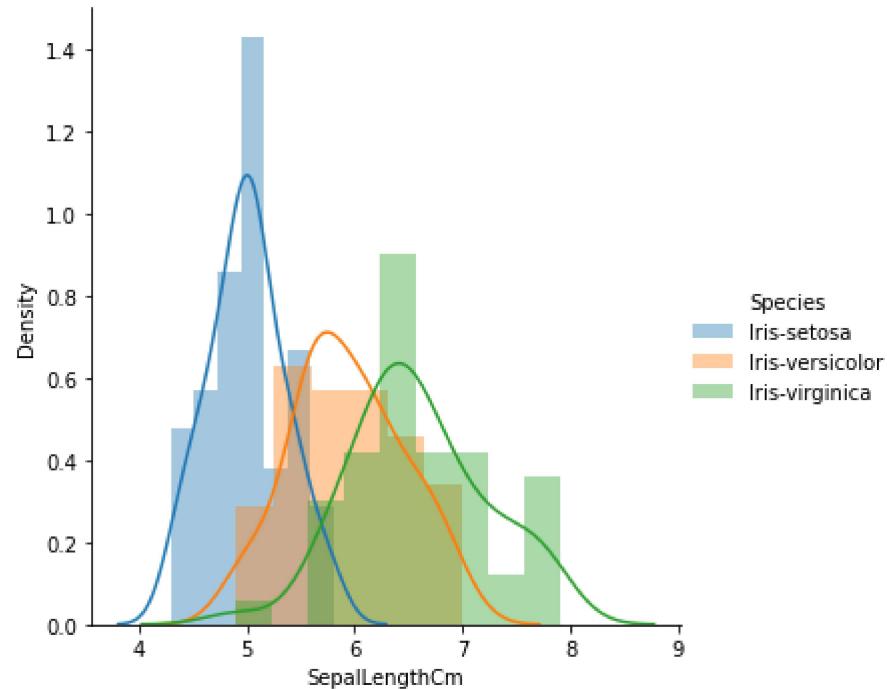
	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	149.000000	149.000000	149.000000	149.000000
mean	5.848322	3.051007	3.774497	1.205369
std	0.828594	0.433499	1.759651	0.761292
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.400000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

FacetGrit

```
In [35]: 1 sns.FacetGrid(df, hue = "Species", height = 5).map(sns.distplot,"SepalLengthCm").add_legend()
```

```
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)  
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)  
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
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figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)
```

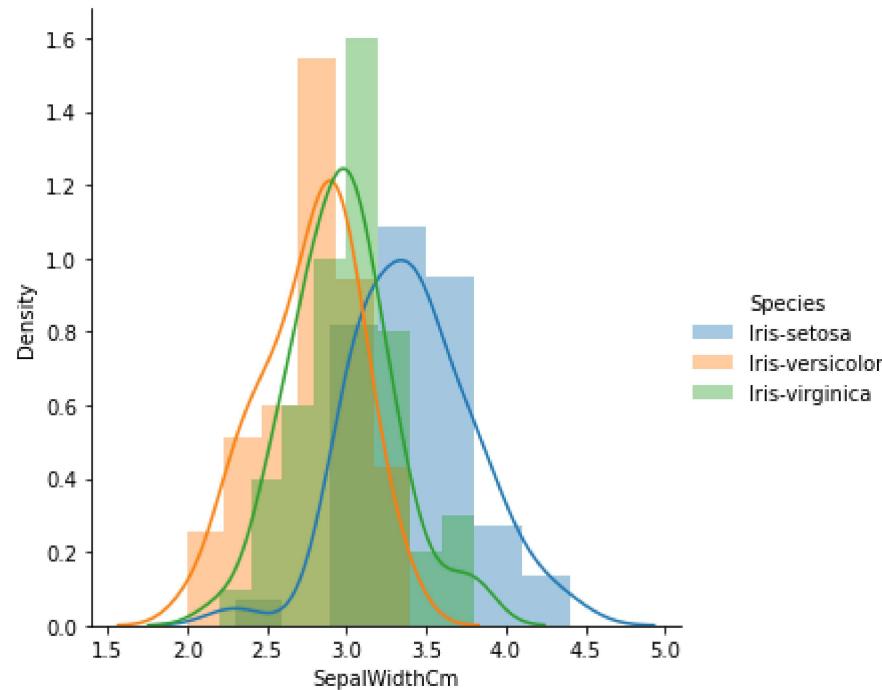
```
Out[35]: <seaborn.axisgrid.FacetGrid at 0x25035bf8d30>
```



```
In [36]: 1 sns.FacetGrid(df, hue = "Species", height = 5).map(sns.distplot,"SepalWidthCm").add_legend()
```

```
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)  
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)  
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)
```

```
Out[36]: <seaborn.axisgrid.FacetGrid at 0x25038f04d00>
```



```
In [37]: 1 sns.FacetGrid(df, hue = "Species", height = 5).map(sns.distplot,"PetalLengthCm").add_legend()
```

C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

```
    warnings.warn(msg, FutureWarning)
```

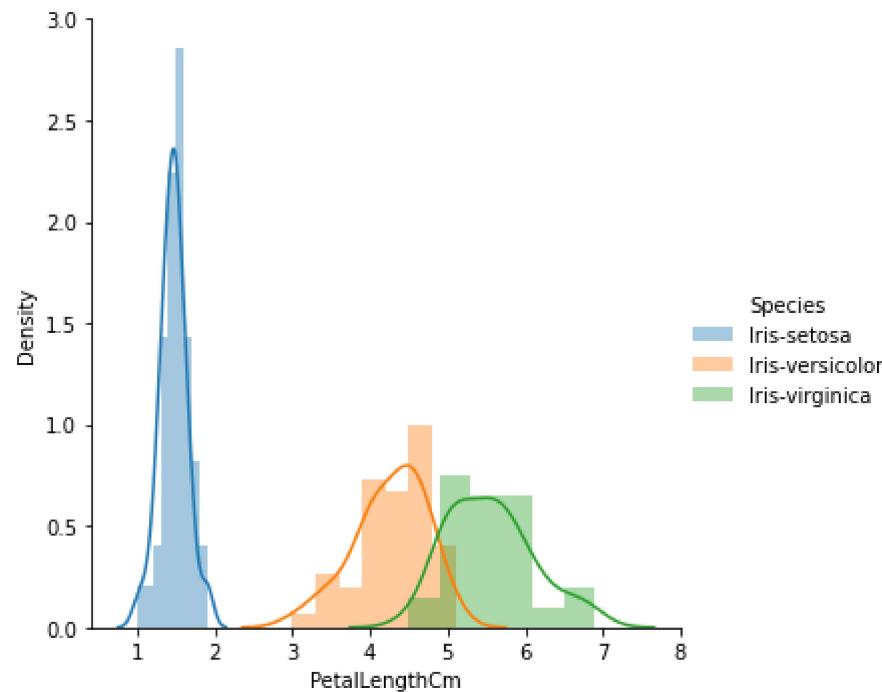
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

```
    warnings.warn(msg, FutureWarning)
```

C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

```
    warnings.warn(msg, FutureWarning)
```

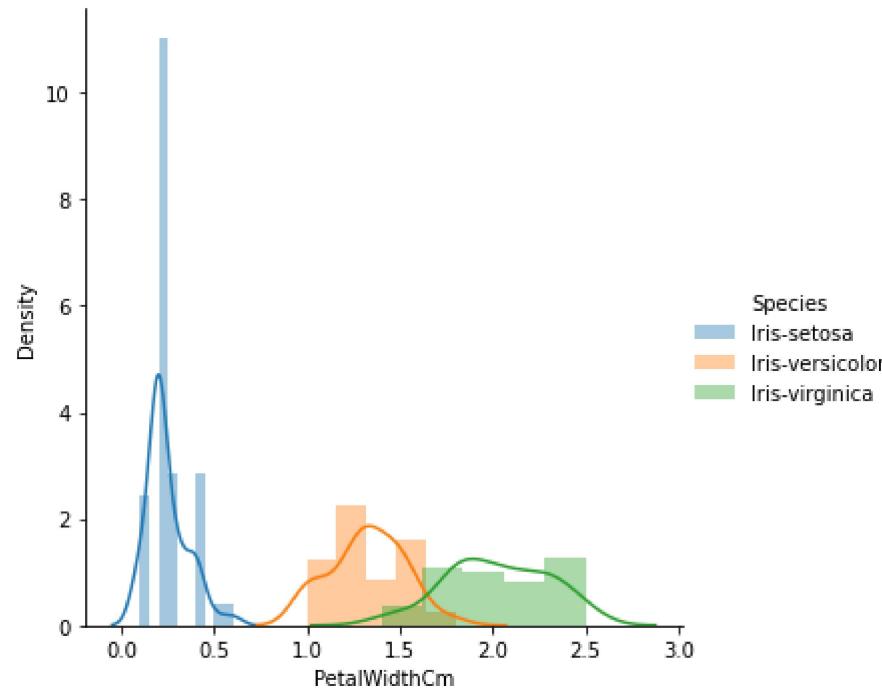
```
Out[37]: <seaborn.axisgrid.FacetGrid at 0x25038fa7730>
```



```
In [38]: 1 sns.FacetGrid(df, hue = "Species", height = 5).map(sns.distplot,"PetalWidthCm").add_legend()
```

```
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)  
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)  
C:\Users\Shivv\anaconda3\lib\site-packages\seaborn\distributions.py:2619: FutureWarning: `distplot` is a dep  
recated function and will be removed in a future version. Please adapt your code to use either `displot` (a  
figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).  
    warnings.warn(msg, FutureWarning)
```

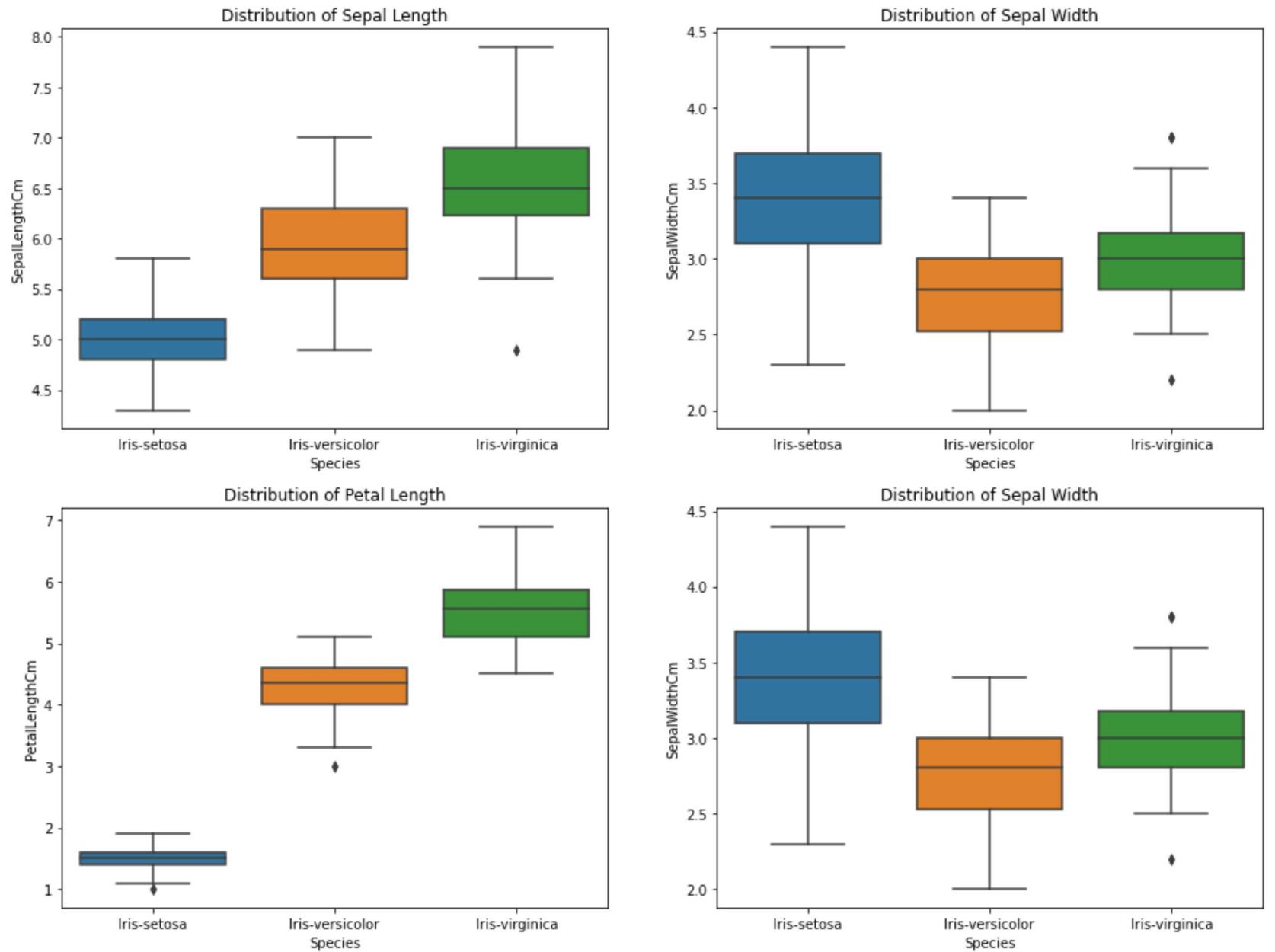
```
Out[38]: <seaborn.axisgrid.FacetGrid at 0x250390404f0>
```



SubPlots

In [42]:

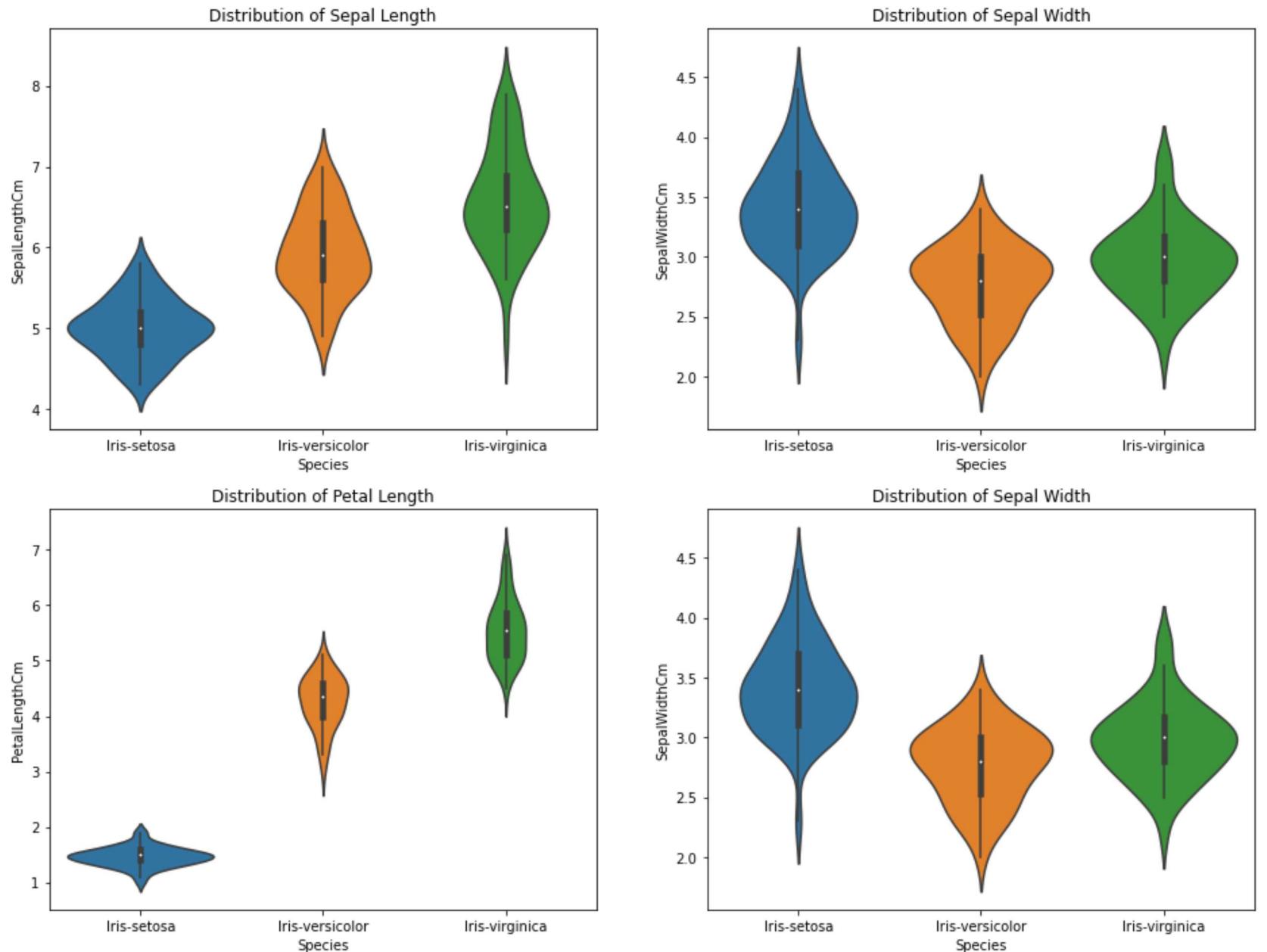
```
1 fig, axes = plt.subplots(2,2, figsize=(16,12))
2 axes[0,0].set_title("Distribution of Sepal Length")
3 sns.boxplot(y = "SepalLengthCm", x ="Species", data= df , orient = 'v', ax = axes[0,0])
4 axes[0,1].set_title("Distribution of Sepal Width")
5 sns.boxplot(y = "SepalWidthCm", x ="Species", data= df , orient = 'v', ax = axes[0,1])
6 axes[1,0].set_title("Distribution of Petal Length")
7 sns.boxplot(y = "PetalLengthCm", x ="Species", data= df , orient = 'v', ax = axes[1,0])
8 axes[1,1].set_title("Distribution of Sepal Width")
9 sns.boxplot(y = "SepalWidthCm", x ="Species", data= df , orient = 'v', ax = axes[1,1])
10 plt.show()
```



ViolinPlot

In [44]:

```
1 fig, axes = plt.subplots(2,2, figsize=(16,12))
2 axes[0,0].set_title("Distribution of Sepal Length")
3 sns.violinplot(y = "SepalLengthCm", x ="Species", data= df , orient = 'v', ax = axes[0,0])
4 axes[0,1].set_title("Distribution of Sepal Width")
5 sns.violinplot(y = "SepalWidthCm", x ="Species", data= df , orient = 'v', ax = axes[0,1])
6 axes[1,0].set_title("Distribution of Petal Length")
7 sns.violinplot(y = "PetalLengthCm", x ="Species", data= df , orient = 'v', ax = axes[1,0])
8 axes[1,1].set_title("Distribution of Sepal Width")
9 sns.violinplot(y = "SepalWidthCm", x ="Species", data= df , orient = 'v', ax = axes[1,1])
10 plt.show()
```

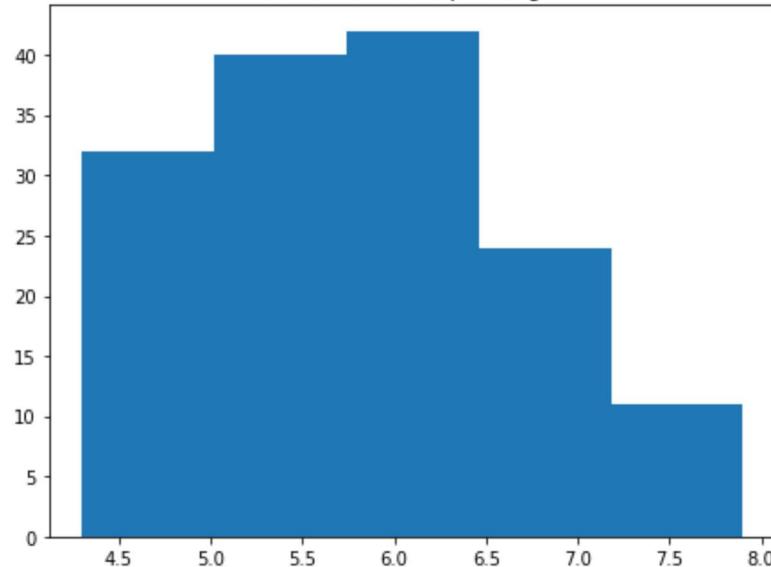


Histogram Plot

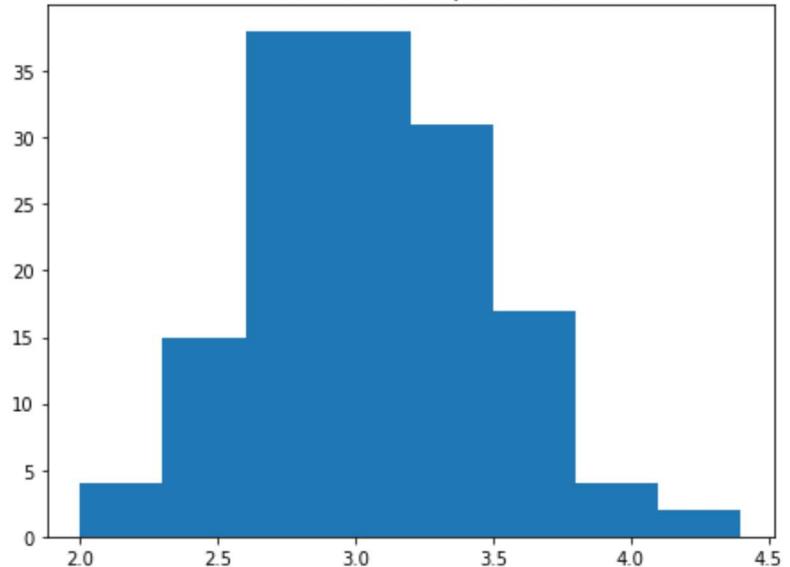
In [13]:

```
1 fig, axes = plt.subplots(2,2, figsize=(16,12))
2 axes[0,0].set_title("Distribution of Sepal Length")
3 axes[0,0].hist(df["SepalLengthCm"], bins = 5);
4 axes[0,1].set_title("Distribution of Sepal Width")
5 axes[0,1].hist(df["SepalWidthCm"], bins = 8);
6 axes[1,0].set_title("Distribution of Petal Length")
7 axes[1,0].hist(df["PetalLengthCm"], bins = 5);
8 axes[1,1].set_title("Distribution of Sepal Width")
9 axes[1,1].hist(df["PetalWidthCm"], bins = 8);
10 plt.show()
```

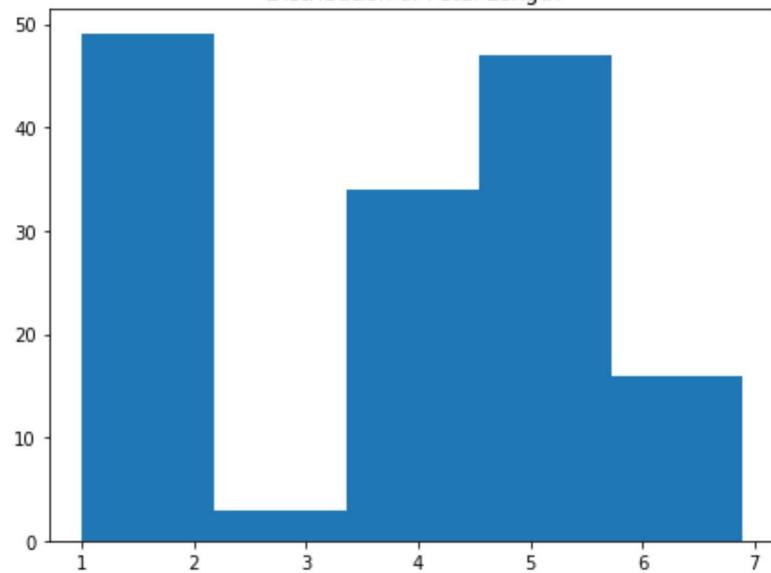
Distribution of Sepal Length



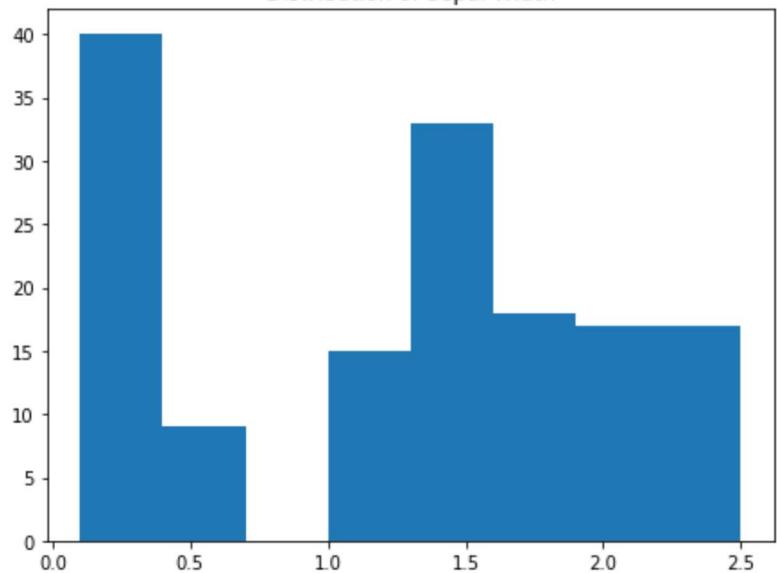
Distribution of Sepal Width



Distribution of Petal Length



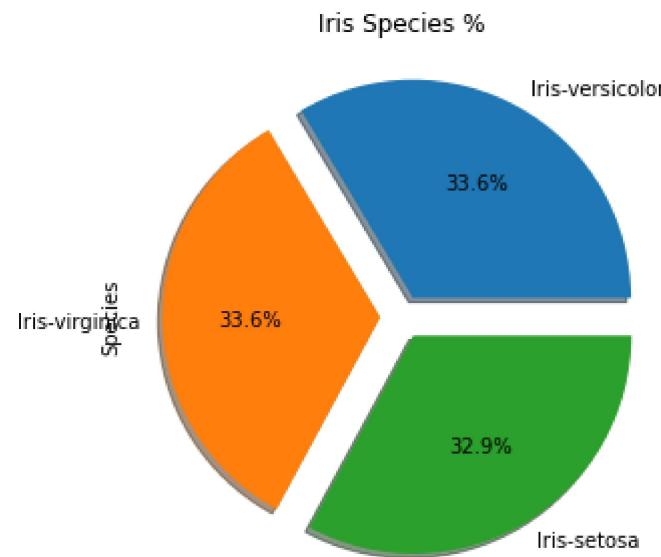
Distribution of Sepal Width



Pie Plot

In [38]:

```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 plt.figure(figsize= (50,15))
4 axes = plt.subplot(111)
5 df['Species'].value_counts().plot.pie(explode=[0.1,0.1,0.1],autopct = '%1.1f%%', shadow = True,figsize= (50,15))
6 plt.title("Iris Species %")
7 plt.show()
```



In [62]:

```
1 df.corr()
```

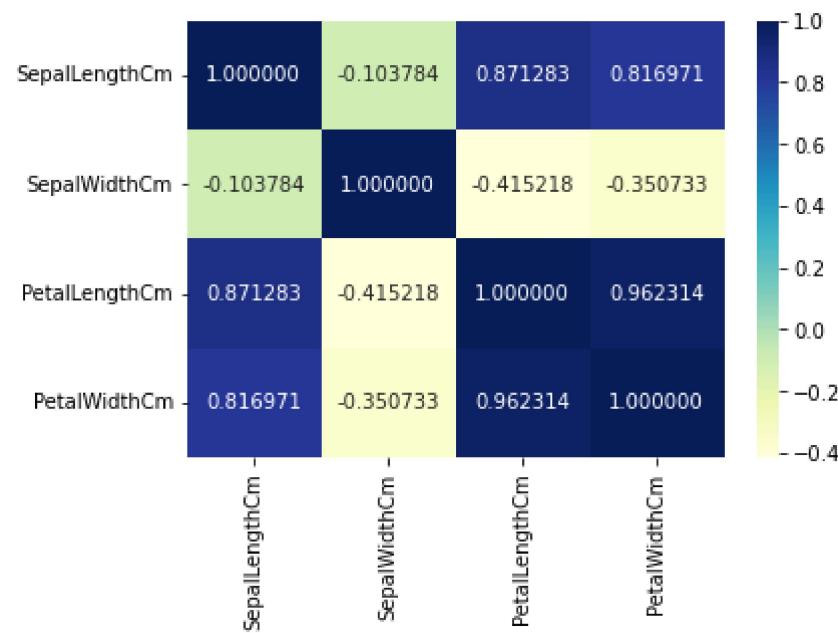
Out[62]:

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
SepalLengthCm	1.000000	-0.103784	0.871283	0.816971
SepalWidthCm	-0.103784	1.000000	-0.415218	-0.350733
PetalLengthCm	0.871283	-0.415218	1.000000	0.962314
PetalWidthCm	0.816971	-0.350733	0.962314	1.000000

Heat Map

In [40]:

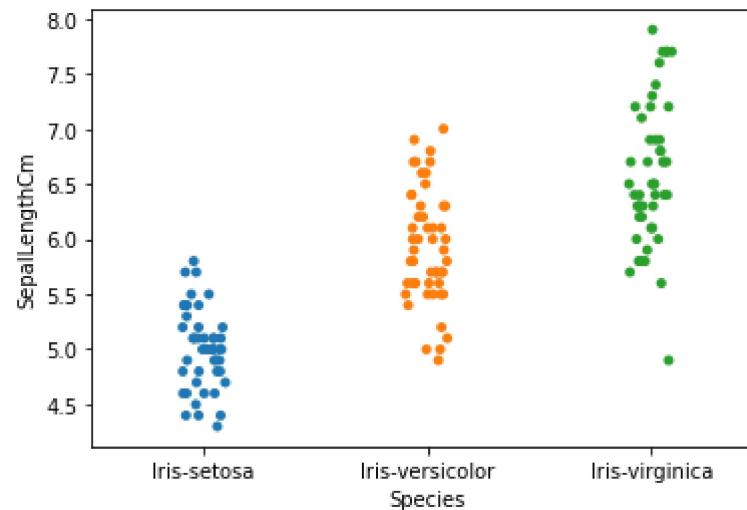
```
1 sns.heatmap(df.corr(), cmap = "YlGnBu", annot = True, fmt = "f")
2 plt.show()
```



Strip Plot

In [41]:

```
1 sns.stripplot( y = 'SepalLengthCm', x = 'Species', data = df)
2 plt.show()
```



In [45]:

```
1 #seperat features and target
2 data = df.values
3 X = data[:,0:4]
4 Y = data[:,4]
```

Train and Test Data

In [47]:

```
1 #split the data to train and test
2 from sklearn.model_selection import train_test_split
3 X_train, X_test, Y_train, Y_test = train_test_split(X,Y, test_size = 0.2)
```

Support Vector Machine

```
In [51]: 1 #support vector machine algorithm
          2 from sklearn.svm import SVC
          3 svn = SVC()
          4 svn.fit(X_train, Y_train)
```

```
Out[51]: SVC()
```

```
In [54]: 1 #predict from the dataset
          2 predictions = svn.predict(X_test)
          3
```

accuracy of data

```
In [55]: 1 from sklearn.metrics import accuracy_score
          2 accuracy_score(Y_test, predictions)
```

```
Out[55]: 0.8666666666666667
```

Classification Report

```
In [56]: 1 #distributed the classification report
          2 from sklearn.metrics import classification_report
          3 classification_report(Y_test, predictions)
```

```
Out[56]: 'precision    recall    f1-score    support\nIris-setosa      1.00      1.00      1.00
Iris-versicolor   0.77      0.91      0.83      11\nIris-virginica  0.91      0.77      0.83
accuracy         0.87      0.87      0.87      30\nmacro avg       0.89      0.89      0.89      0.
weighted avg     0.88      0.87      0.87      30\n'
```

Test Model

In [58]:

```
1 #Test the model
2 x_new = np.array([[2,3,5,5],[1,2,3.8,0],[5.3,2.5,4.6,1.9]])
3 #prediction of the species from the input vector
4 predictions = svn.predict(x_new)
5 print("Predictions of Species: {}".format(predictions))
```

Predictions of Species: ['Iris-virginica' 'Iris-versicolor' 'Iris-versicolor']

Thank You!

In []:

```
1
```