

```
In [1]: 1 import numpy as np
        2 import pandas as pd
        3 import matplotlib.pyplot as plt
        4 import seaborn as sns
        5 %matplotlib inline
```

```
In [2]: 1 idata=pd.read_csv(r'S:\DOCS\Sep\Sep_7\IRIS DATASET _ ADVANCE VISUALIZATION _ EDA 2\Iris.csv')
```

```
In [3]: 1 idata
```

```
Out[3]:
```

	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 [4]: 1 idata.shape
```

```
Out[4]: (150, 6)
```

```
In [5]: 1 idata.head()
```

```
Out[5]:
```

	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

```
In [6]: 1 idata.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 [7]: 1 idata['Species'].unique()
```

```
Out[7]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
```

```
In [8]: 1 idata.Species.value_counts()
```

```
Out[8]: Iris-setosa      50
Iris-versicolor      50
Iris-virginica       50
Name: Species, dtype: int64
```

```
In [9]: 1 # there are three species of plant
```

```
In [10]: 1 idata.isnull().any().any()
```

```
Out[10]: False
```

```
In [11]: 1 # no null value in dataset
```

```
In [12]: 1 idata.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 [13]: 1 # Species is categorical variable type,so convert from object to categorical datatype
```

```
In [14]: 1 idata['Species'].astype('category')
```

```
Out[14]: 0      Iris-setosa
1      Iris-setosa
2      Iris-setosa
3      Iris-setosa
4      Iris-setosa
...
145    Iris-virginica
146    Iris-virginica
147    Iris-virginica
148    Iris-virginica
149    Iris-virginica
Name: Species, Length: 150, dtype: category
Categories (3, object): ['Iris-setosa', 'Iris-versicolor', 'Iris-virginica']
```

```
In [15]: 1 idata.head()
```

```
Out[15]:
```

	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

```
In [16]: 1 # there is no need of id column as it is not useful in our analysis
```

```
In [17]: 1 idata.drop('Id', axis=1, inplace=True)
```

```
In [18]: 1 idata.head()
```

```
Out[18]:
```

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

```
In [19]: 1 # Species is dependent variable and others are independent variable.
```

```
In [20]: 1 # Univariate analysis
```

```
In [21]: 1  
2 sns.countplot(x=idata['Species'])
```

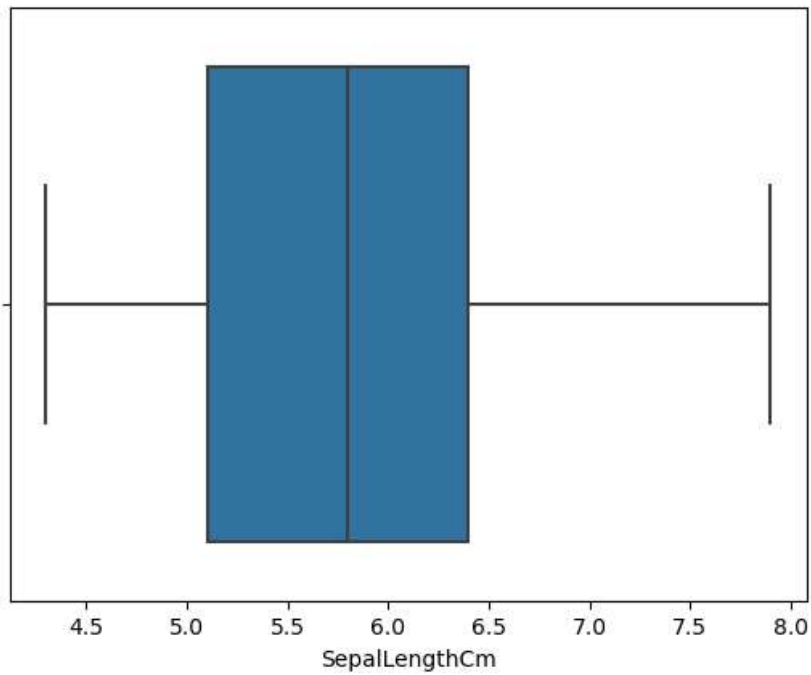
```
Out[21]: <Axes: xlabel='Species', ylabel='count'>
```



```
In [22]: 1 # counts of each species is same
```

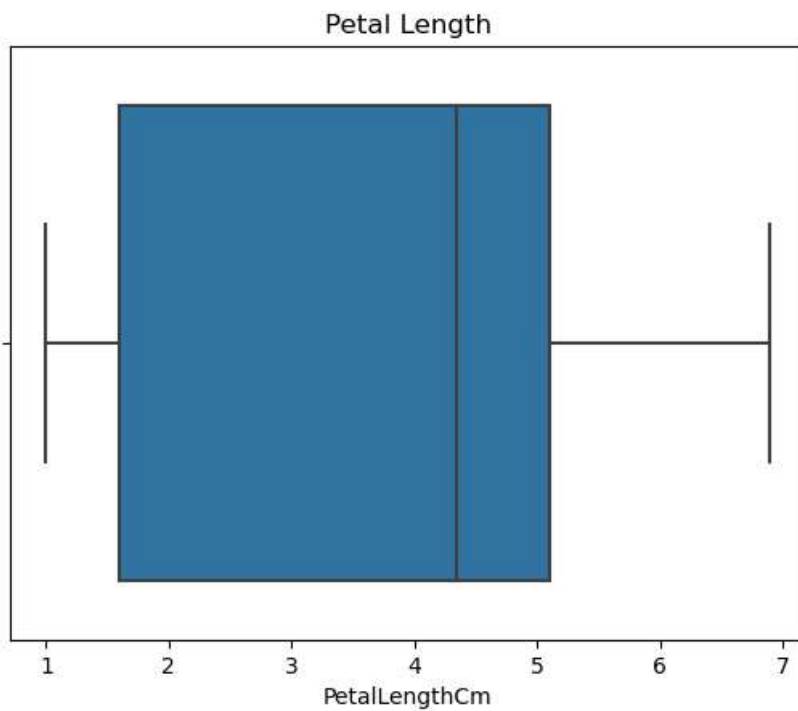
```
In [23]: 1 sns.boxplot(data=idata, x='SepalLengthCm', hue='Species')
```

```
Out[23]: <Axes: xlabel='SepalLengthCm'>
```



```
In [24]: 1 sns.boxplot(data=idata, x='PetalLengthCm')  
2 plt.title('Petal Length')
```

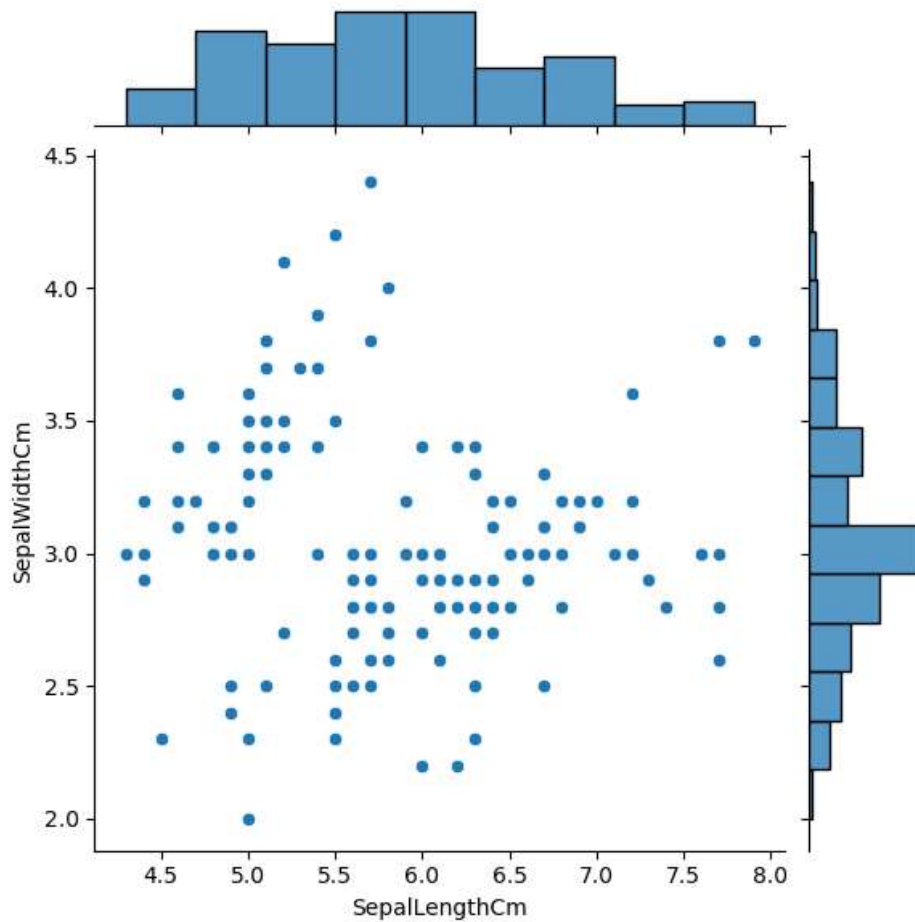
```
Out[24]: Text(0.5, 1.0, 'Petal Length')
```



```
In [25]: 1 # bi variate analysis  
2 # jointplot
```

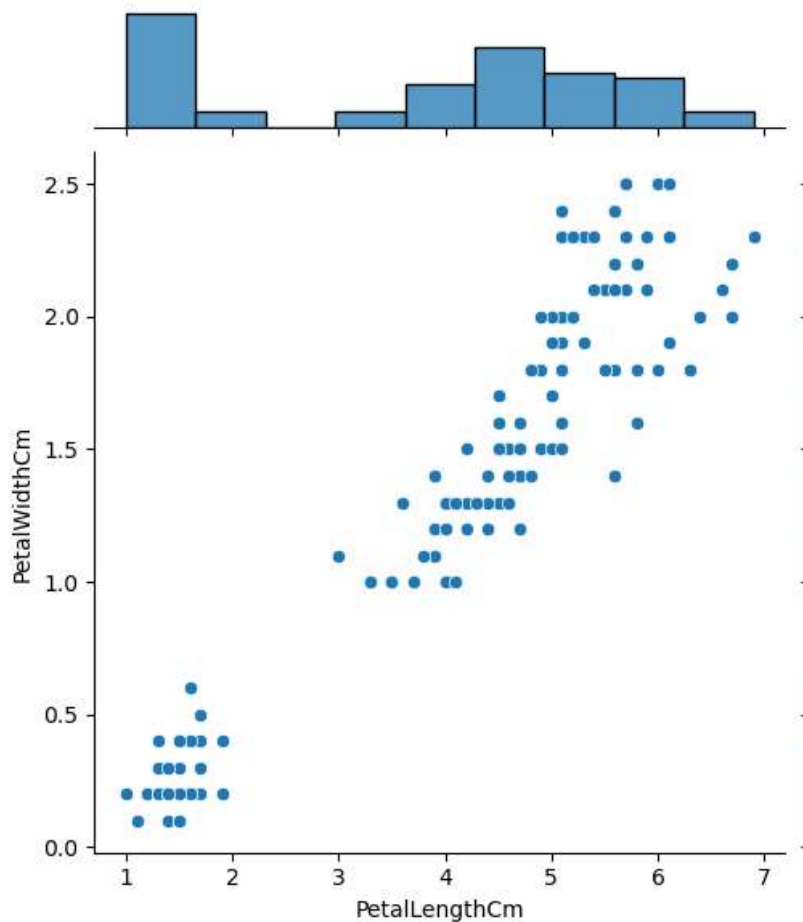
```
In [26]: 1 sns.jointplot(data=idata, x='SepalLengthCm', y='SepalWidthCm')
```

```
Out[26]: <seaborn.axisgrid.JointGrid at 0x20310635610>
```



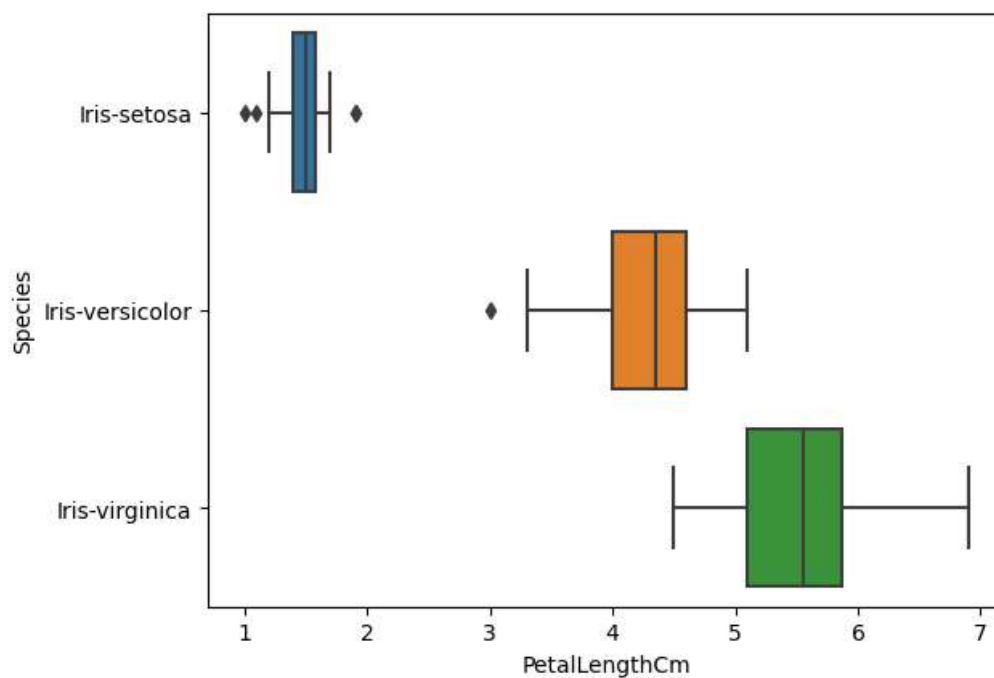
```
In [27]: 1 sns.jointplot(data=idata, x='PetalLengthCm', y='PetalWidthCm')
```

```
Out[27]: <seaborn.axisgrid.JointGrid at 0x203104413d0>
```



```
In [30]: 1 sns.boxplot(data=idata, x='PetalLengthCm', y='Species')
```

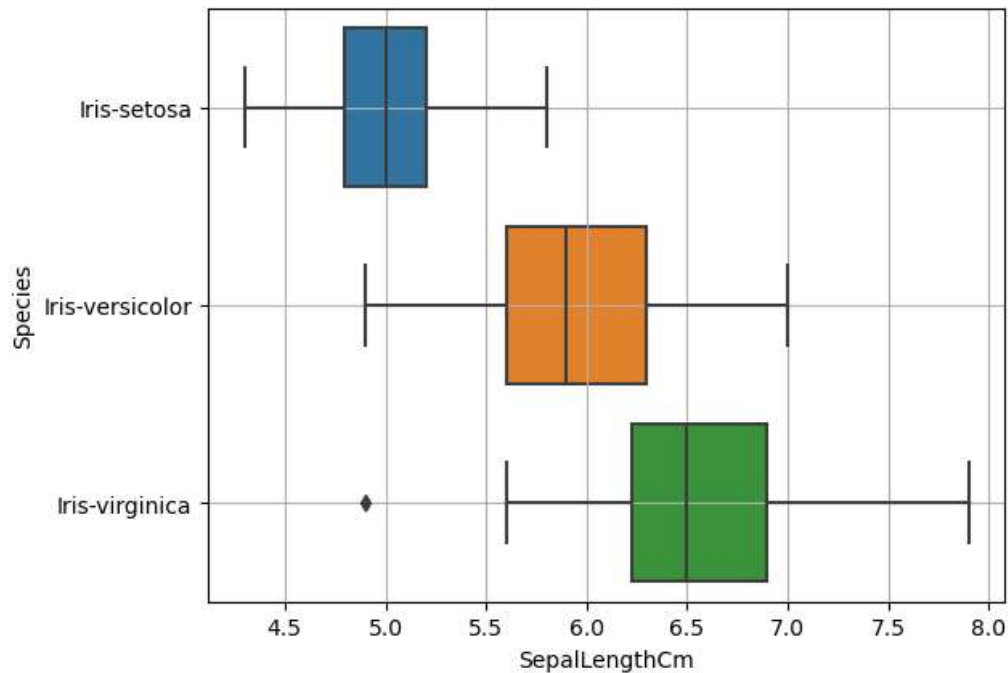
```
Out[30]: <Axes: xlabel='PetalLengthCm', ylabel='Species'>
```



```
In [31]: 1 # Findings
2 '''
3 * petal length of serosa plant is comparatively less around 1 to 2 cm
4 * petwl length of versicolor in between 3 to 5 cm
5 * virginica have largest petal length around 4.5 to 7 cm
6 '''
```

Out[31]: '\n* petal length of serosa plant is comparatively less around 1 to 2 cm \n* petwl length of versicolor in between 3 to 5 cm\n* virginica have largest petal length around 4.5 to 7 cm \n'

```
In [43]: 1 sns.boxplot(data=idata, x='SepalLengthCm', y='Species')
2 plt.grid()
```



```
In [36]: 1 '''
2 iris-setosa is shortest sepal with median length of 5 cm
3 iris-versicolor have slightly longer sepal length with median of 5.9 cm
4 iris-virginica have longer sepal length with median of 6.5 cm
5
6
7 '''
```

Out[36]: '\niris-setosa is shortest sepal with median length of 5 cm\niris-versicolor have slightly longer sepal length with median of 5.9 cm\niris-virginica have longer sepal length with median of 6.5 cm \n\n'

```
In [56]: 1 setosa_data=idata['Species']=='Iris-setosa'
2 idata[setosa_data].describe()
```

```
Out[56]:
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.00000	50.000000	50.000000	50.00000
mean	5.00600	3.418000	1.464000	0.24400
std	0.35249	0.381024	0.173511	0.10721
min	4.30000	2.300000	1.000000	0.10000
25%	4.80000	3.125000	1.400000	0.20000
50%	5.00000	3.400000	1.500000	0.20000
75%	5.20000	3.675000	1.575000	0.30000
max	5.80000	4.400000	1.900000	0.60000

```
In [59]: 1 virginica_data=idata['Species'] == 'Iris-virginica'
        2 idata[virginica_data].describe()
```

```
Out[59]:
```

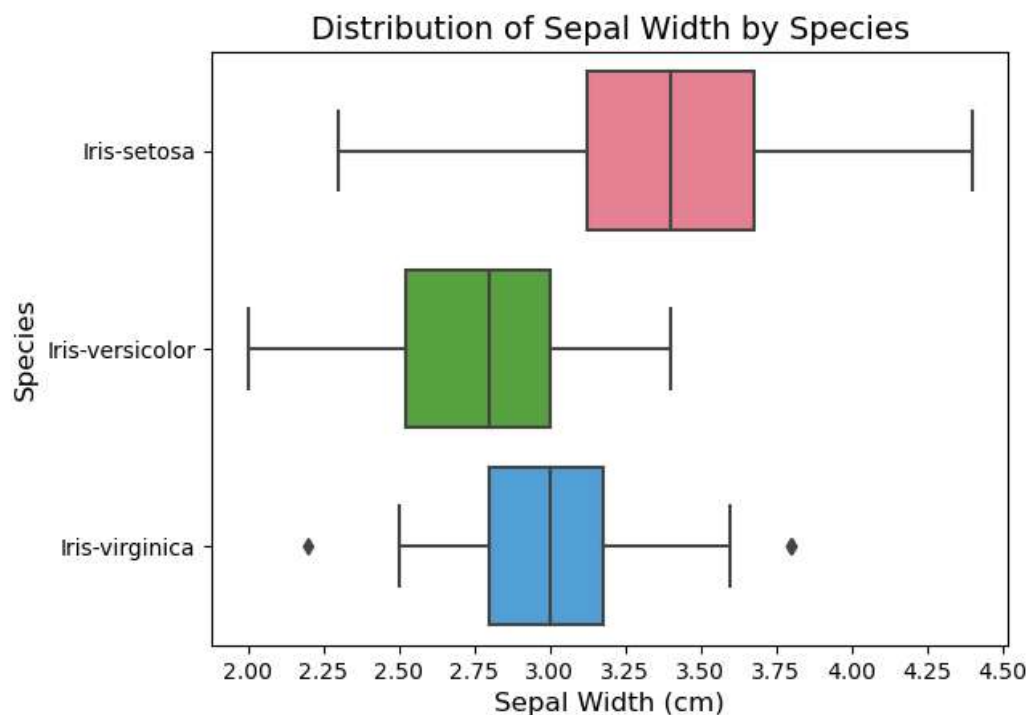
	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.000000	50.000000	50.000000	50.000000
mean	6.588000	2.974000	5.552000	2.026000
std	0.635888	0.322497	0.551895	0.27465
min	4.900000	2.200000	4.500000	1.400000
25%	6.225000	2.800000	5.100000	1.800000
50%	6.500000	3.000000	5.550000	2.000000
75%	6.900000	3.175000	5.875000	2.300000
max	7.900000	3.800000	6.900000	2.500000

```
In [60]: 1 versicolor=idata['Species'] == 'Iris-versicolor'
        2 idata[versicolor].describe()
```

```
Out[60]:
```

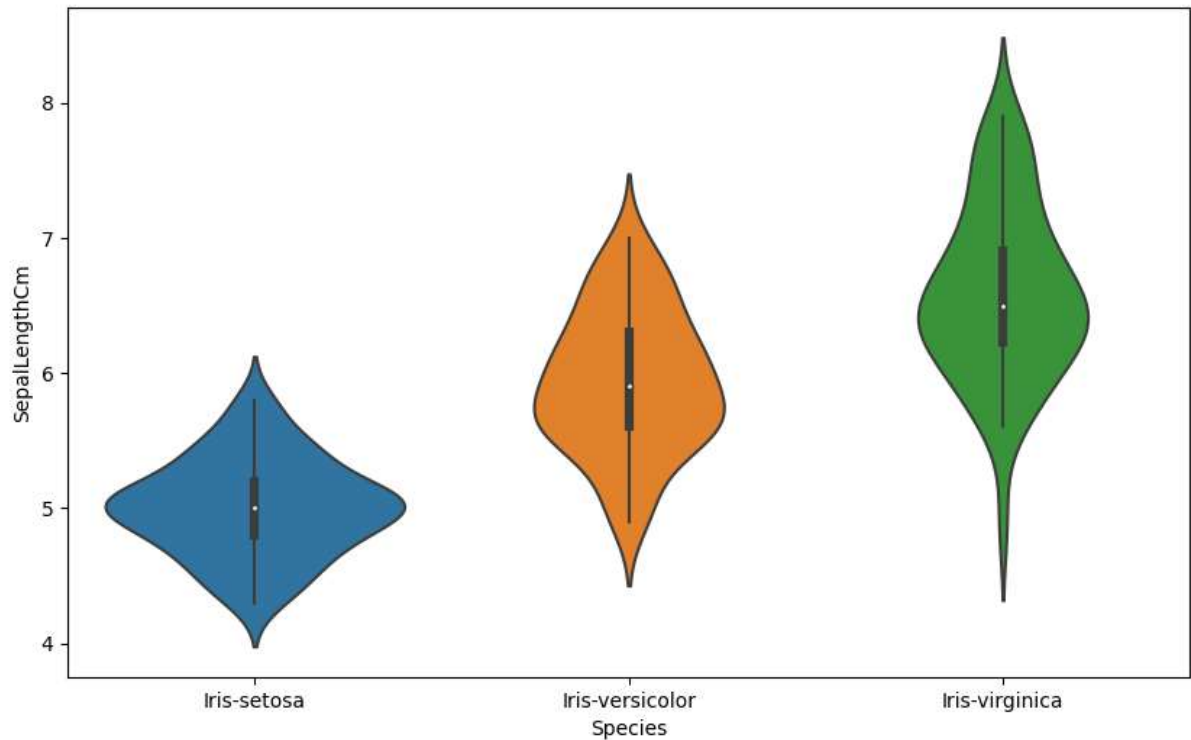
	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
count	50.000000	50.000000	50.000000	50.000000
mean	5.936000	2.770000	4.260000	1.326000
std	0.516171	0.313798	0.469911	0.197753
min	4.900000	2.000000	3.000000	1.000000
25%	5.600000	2.525000	4.000000	1.200000
50%	5.900000	2.800000	4.350000	1.300000
75%	6.300000	3.000000	4.600000	1.500000
max	7.000000	3.400000	5.100000	1.800000

```
In [83]: 1 sns.boxplot(data=idata, x='SepalWidthCm', y='Species', palette="husl")
        2 plt.xlabel('Sepal Width (cm)', fontsize=12)
        3 plt.ylabel('Species', fontsize=12)
        4 plt.title('Distribution of Sepal Width by Species', fontsize=14)
        5 plt.xticks(fontsize=10)
        6 plt.yticks(fontsize=10)
        7 plt.locator_params(axis='x', nbins=11)
        8
```



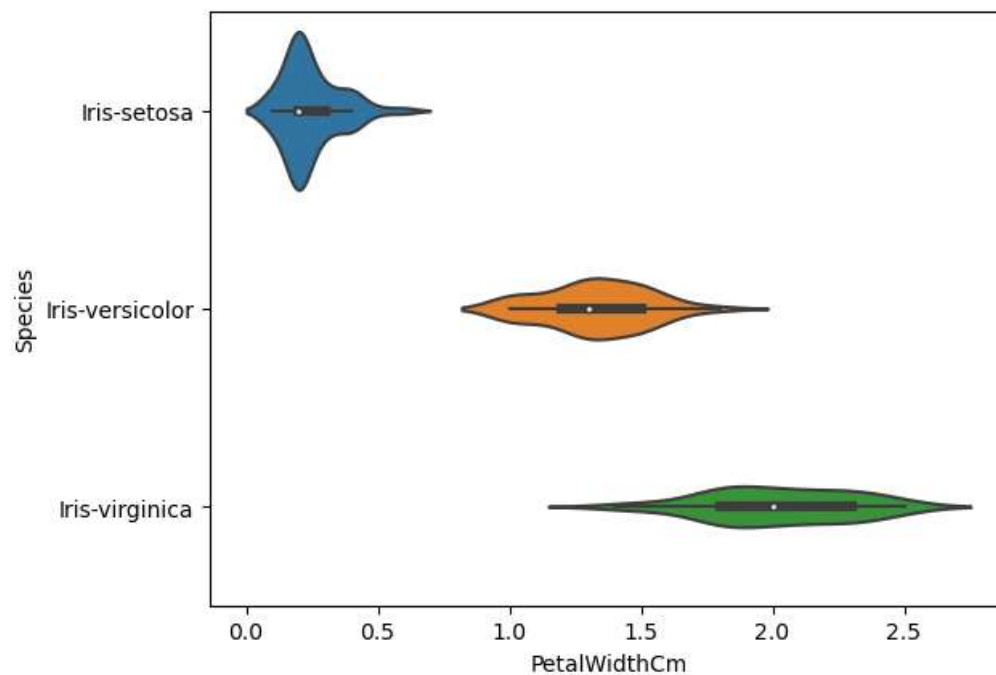

```
In [ ]: 1 '''  
2 iris-setosa have the largest sepal width with median of 3.4 cm  
3 iris-versicolor have smallest sepal width with median of 2.8 cm  
4 iris-virginica sepal width is slightly wider than versicolor  
5 setosa have the widest range between 2.3 cm to 4.4 cm  
6 '''
```

```
In [90]: 1 fig=plt.gcf()  
2 fig.set_size_inches(10,6)  
3 fig=sns.violinplot(data=idata, x='Species', y='SepalLengthCm')
```



```
In [92]: 1 sns.violinplot(data=idata, x='PetalWidthCm', y='Species')
```

Out[92]: <Axes: xlabel='PetalWidthCm', ylabel='Species'>



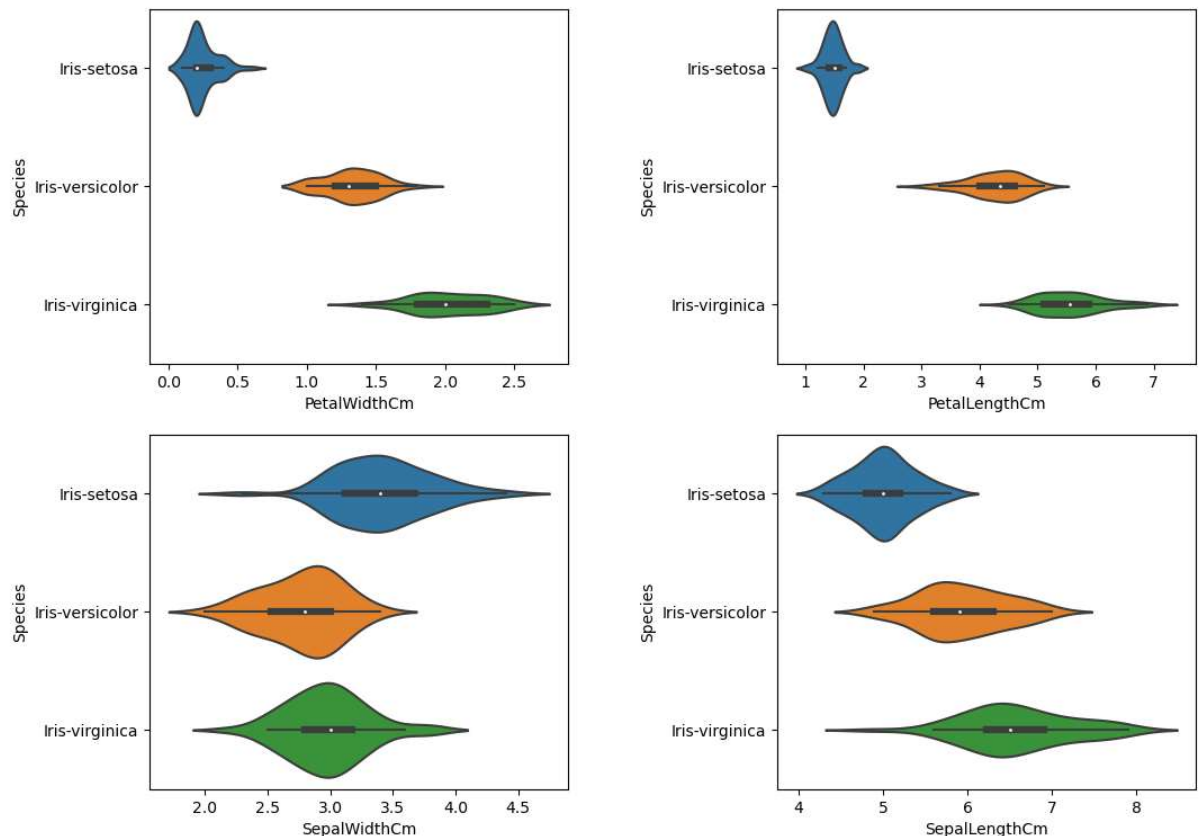
```
In [93]: 1 '''
2 setosa have smallest petal width
3 versicolor have larger petal width
4 virginica have largest petal width
5 petals of all species is skewed to right, means there are more data points at lower range
6 virginica have widest violin shape. means petal width are more spread out
7 '''
```

```
Out[93]: '\nsetosa have smallest petal width \nversicolor have larger petal width\nvirginica have largest petal width\npetals of all species is skewed to right, means there are more data points at lower range\nvirginica have widest violin shape. means petal width are more spread out\n'
```

```
In [94]: 1 # multivariate analysis
```

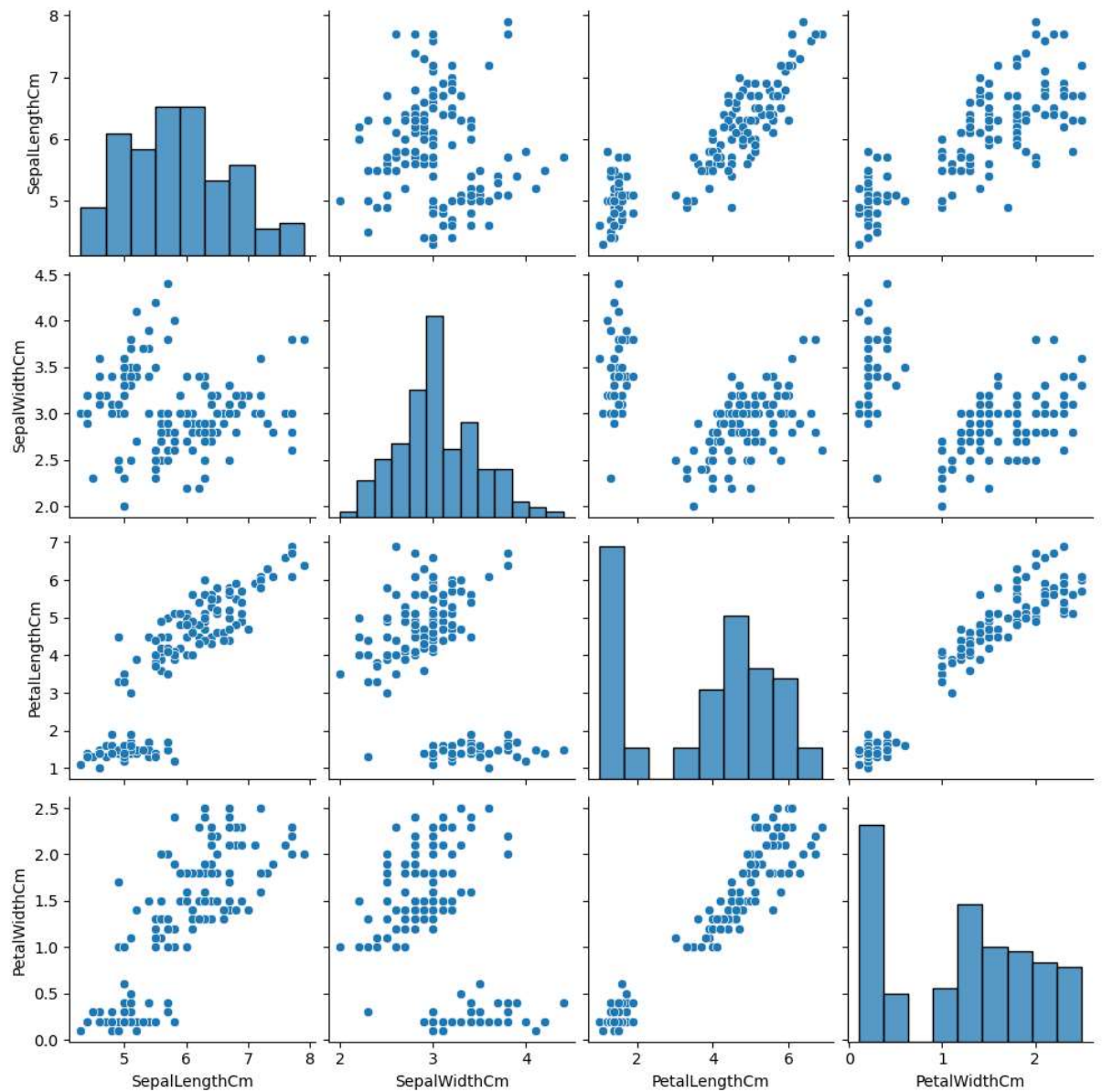
```
In [112]: 1 plt.figure(figsize=(12,9))
2 plt.subplots_adjust(wspace=0.5)
3 plt.subplot(2,2,1)
4 sns.violinplot(data=idata, x='PetalWidthCm', y='Species')
5 plt.subplot(2,2,2)
6 sns.violinplot(data=idata, x='PetalLengthCm', y='Species')
7 plt.subplot(2,2,3)
8 sns.violinplot(data=idata, x='SepalWidthCm', y='Species')
9 plt.subplot(2,2,4)
10 sns.violinplot(data=idata, x='SepalLengthCm', y='Species')
```

```
Out[112]: <Axes: xlabel='SepalLengthCm', ylabel='Species'>
```



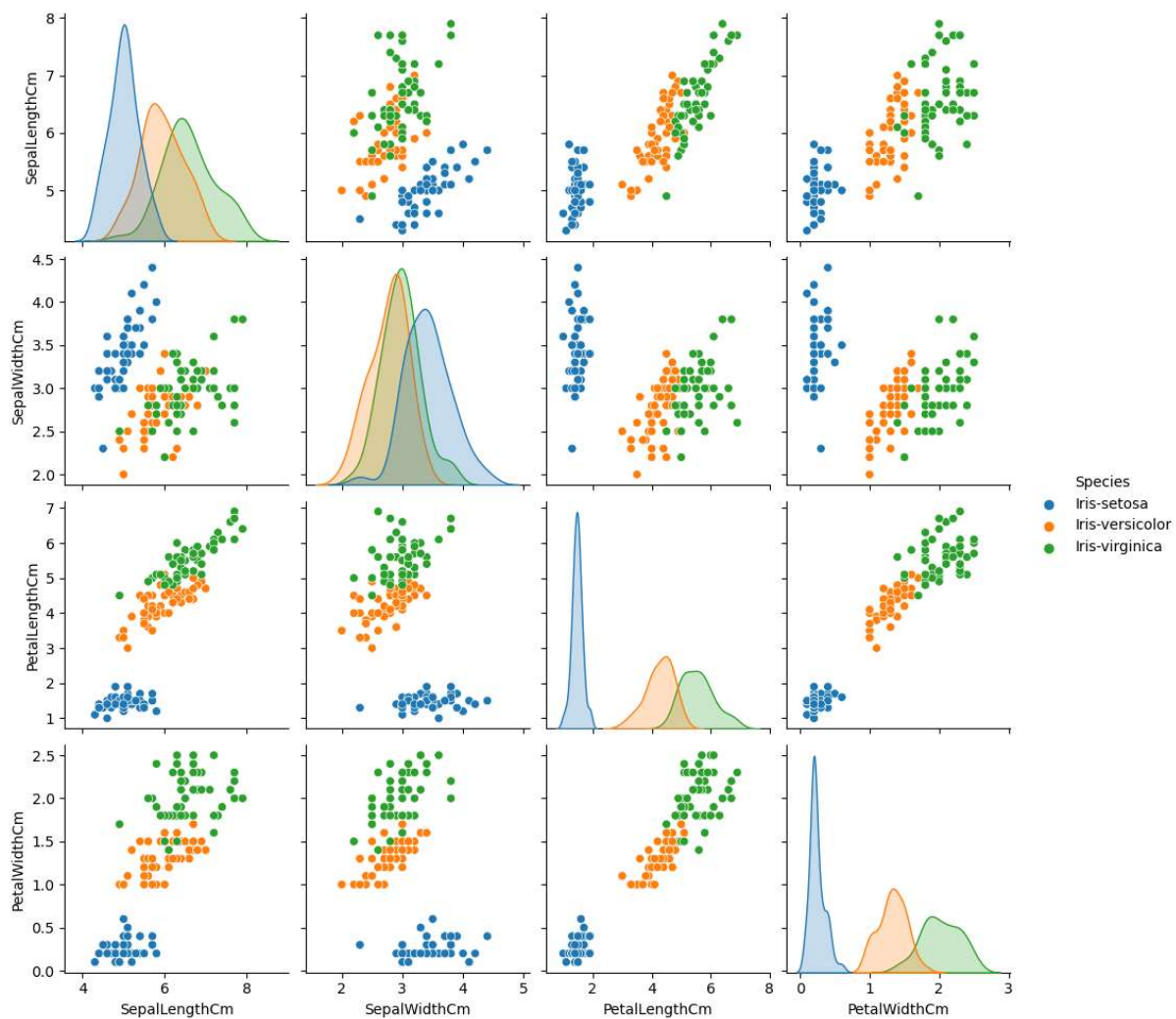
```
In [115]: 1 sns.pairplot(data=idata)
```

```
Out[115]: <seaborn.axisgrid.PairGrid at 0x2031bb612d0>
```



```
In [116]: 1 sns.pairplot(data=idata, hue='Species')
```

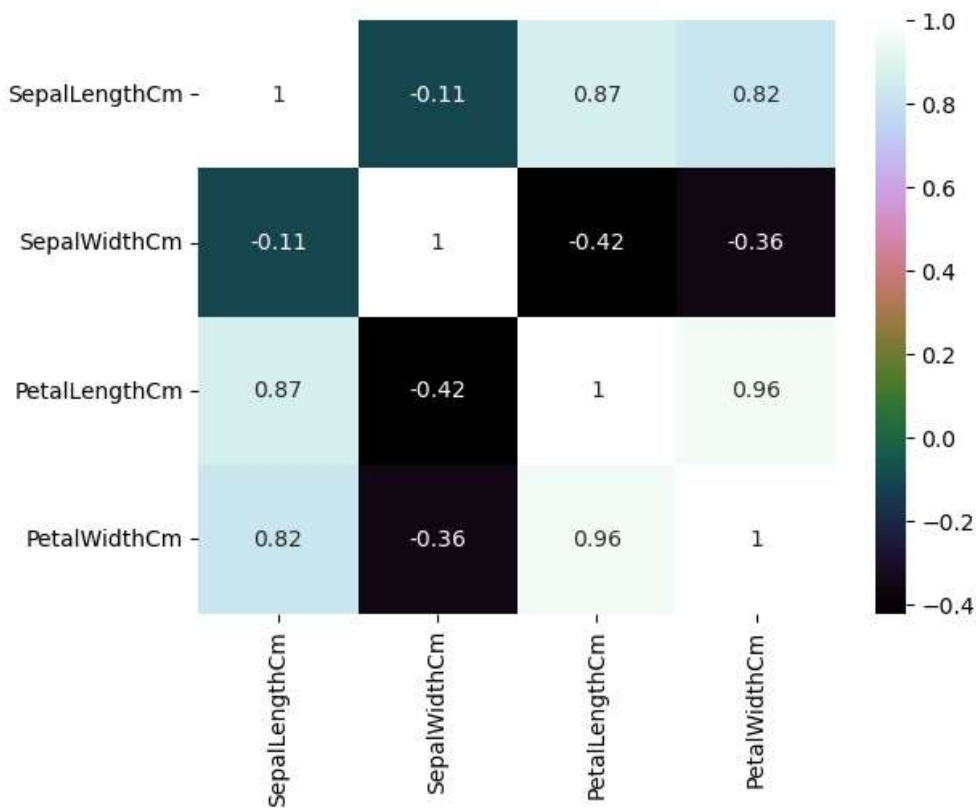
```
Out[116]: <seaborn.axisgrid.PairGrid at 0x2031def92d0>
```



```
In [121]: 1 sns.heatmap(idata.corr(),annot=True,cmap='cubehelix')
```

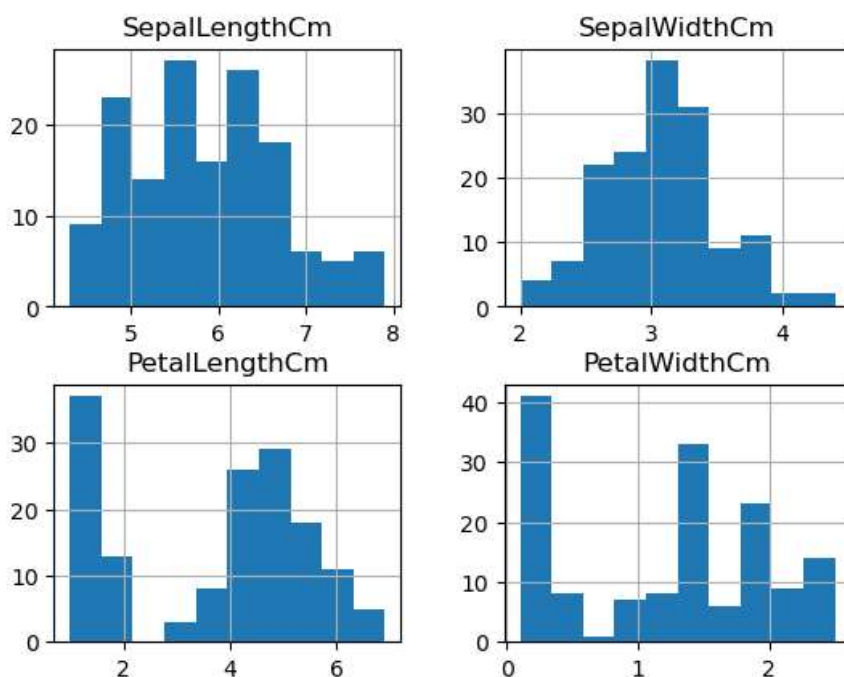
C:\Users\ASUS\AppData\Local\Temp\ipykernel_10876\2409565093.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.
sns.heatmap(idata.corr(),annot=True,cmap='cubehelix')

Out[121]: <Axes: >



```
In [122]: 1 idata.hist() # distribution plot
```

Out[122]: array([[<Axes: title={'center': 'SepalLengthCm'}>,
<Axes: title={'center': 'SepalWidthCm'}>],
[<Axes: title={'center': 'PetalLengthCm'}>,
<Axes: title={'center': 'PetalWidthCm'}>]], dtype=object)

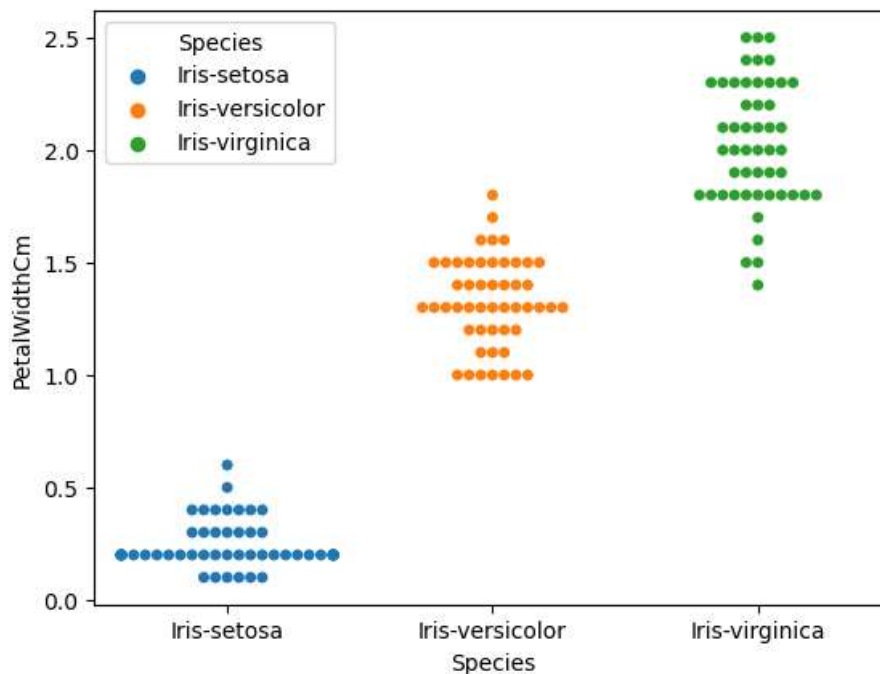


```
In [128]: 1 sns.swarmplot(data=idata,x='Species',y='PetalWidthCm', hue="Species")
```

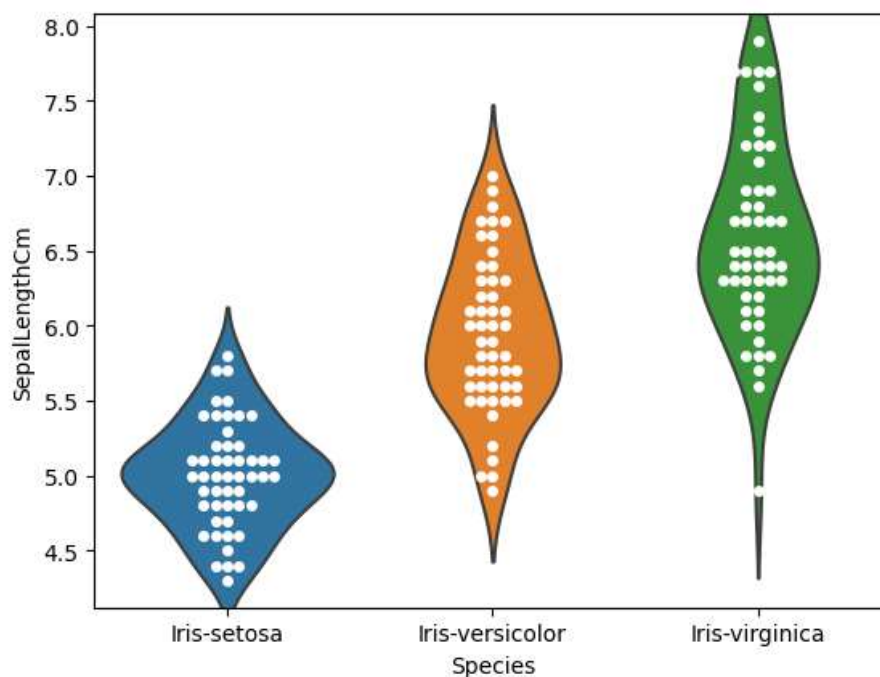
C:\Users\ASUS\anaconda3\Lib\site-packages\seaborn\categorical.py:3544: UserWarning: 6.0% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.
warnings.warn(msg, UserWarning)

```
Out[128]: <Axes: xlabel='Species', ylabel='PetalWidthCm'>
```

C:\Users\ASUS\anaconda3\Lib\site-packages\seaborn\categorical.py:3544: UserWarning: 18.0% of the points cannot be placed; you may want to decrease the size of the markers or use stripplot.
warnings.warn(msg, UserWarning)

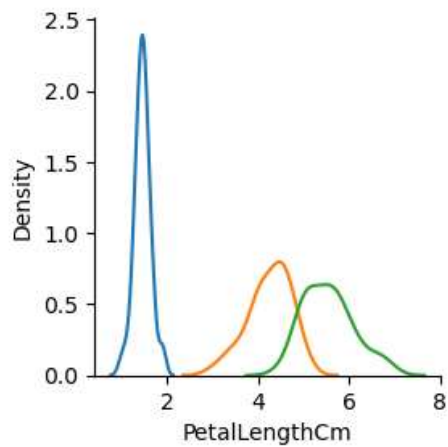


```
In [163]: 1 sns.violinplot(data=idata,x='Species',y='SepalLengthCm', inner=None)
2          2 sns.swarmplot(data=idata,x='Species', y='SepalLengthCm', color='white')
3          3 plt.show()
```

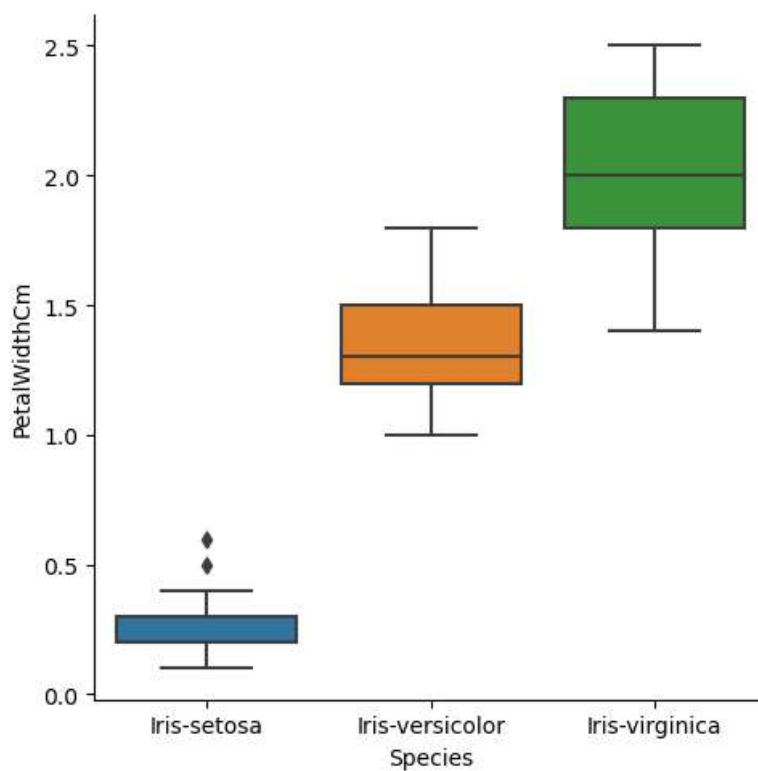


```
In [164]: 1 plt.gcf()
2 g=sns.FacetGrid(idata, hue='Species')
3 g.map(sns.kdeplot, 'PetalLengthCm')
4 plt.ioff()
5 plt.show()
```

<Figure size 640x480 with 0 Axes>

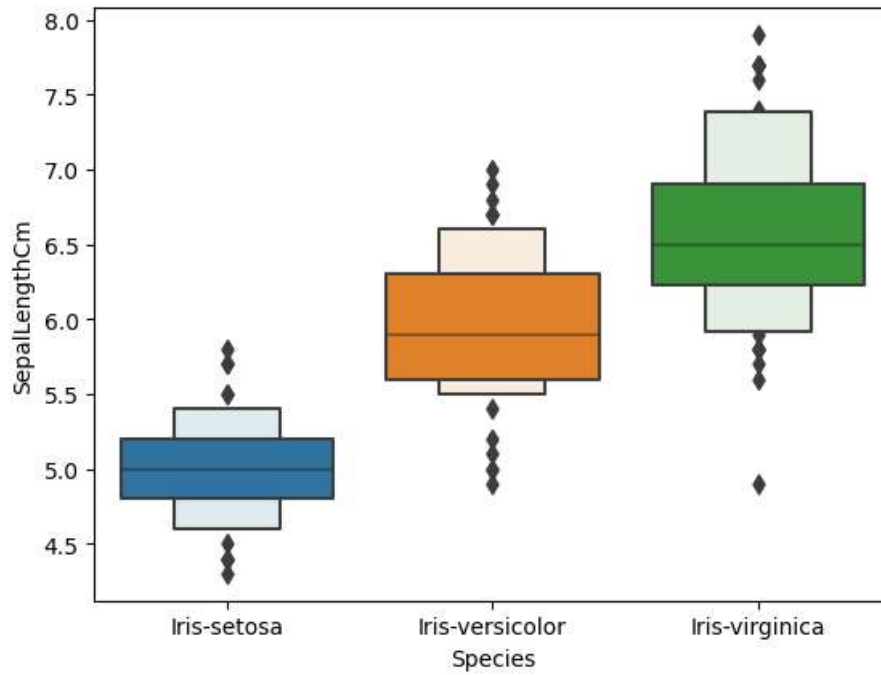


```
In [174]: 1 sns.catplot(data=idata,x='Species',y='PetalWidthCm',kind='box')
2 plt.show()
```



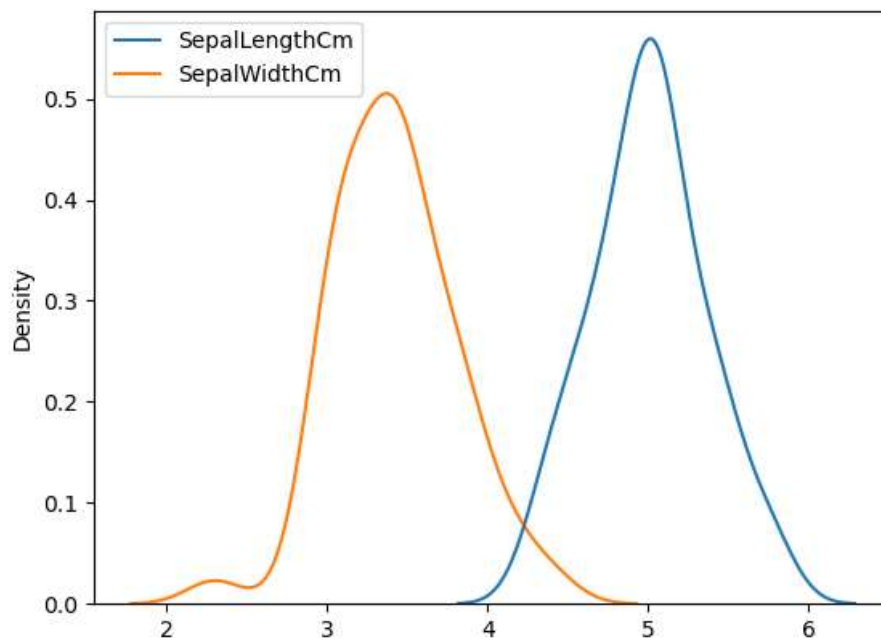

```
In [175]: 1 sns.boxenplot(data=idata,x='Species',y='SepalLengthCm')
```

```
Out[175]: <Axes: xlabel='Species', ylabel='SepalLengthCm'>
```

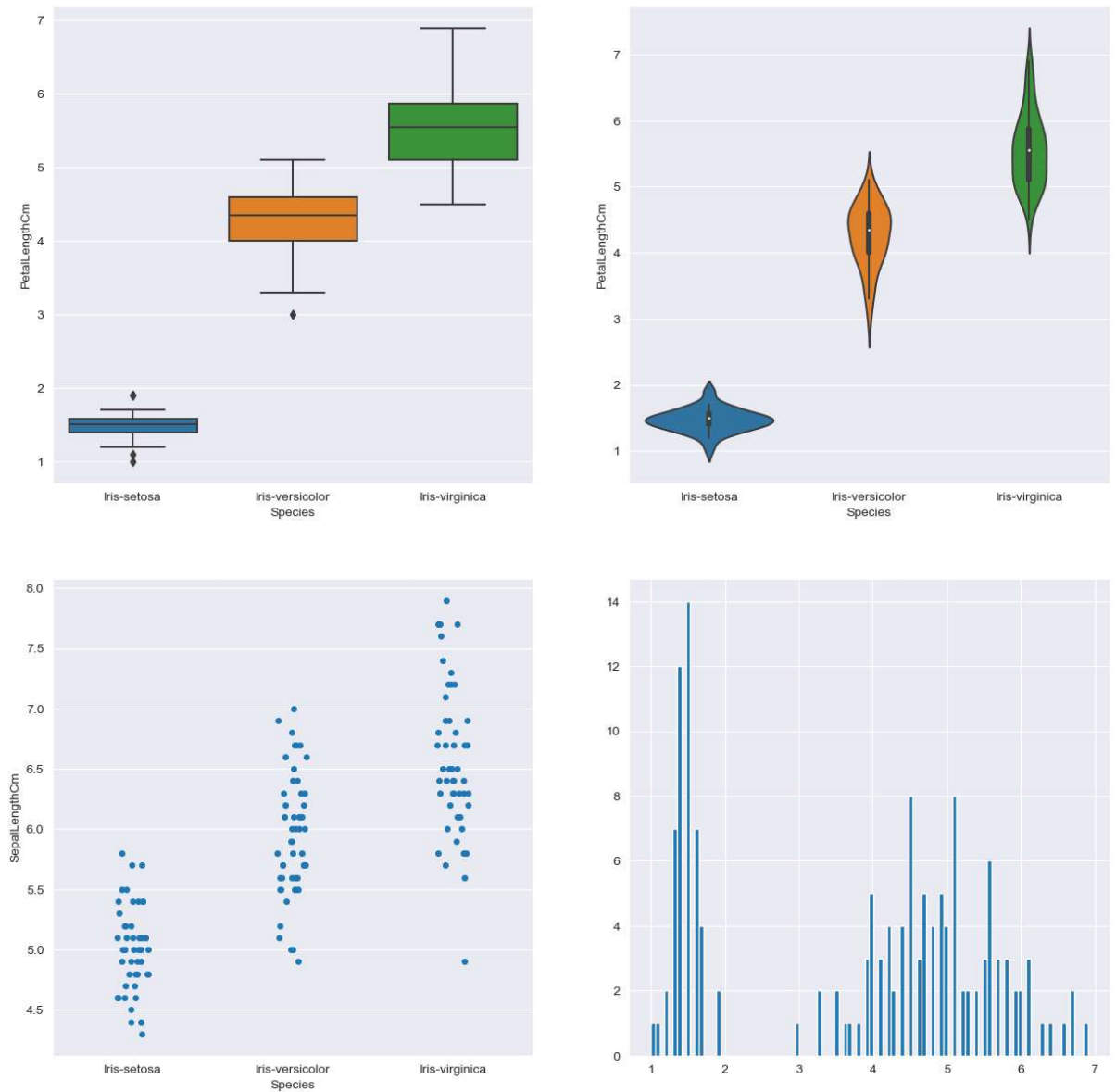


```
In [191]: 1 sub=idata[idata['Species'] == 'Iris-setosa']  
2 sns.kdeplot(data=sub[['SepalLengthCm', 'SepalWidthCm']])
```

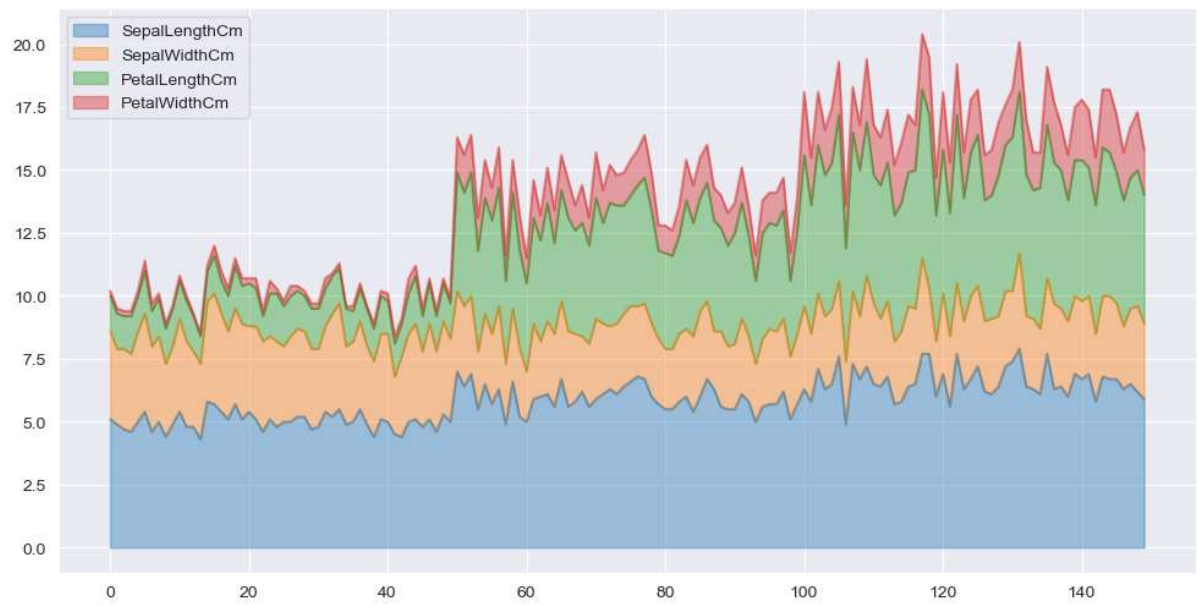
```
Out[191]: <Axes: ylabel='Density'>
```




```
In [195]: 1 sns.set_style('darkgrid')
2 f,axes=plt.subplots(2,2,figsize=(15,15))
3
4 k1=sns.boxplot(x="Species", y="PetalLengthCm", data=idata,ax=axes[0,0])
5 k2=sns.violinplot(x='Species',y='PetalLengthCm',data=idata,ax=axes[0,1])
6 k3=sns.stripplot(x='Species',y='SepalLengthCm',data=idata,jitter=True,ax=axes[1,0])
7 axes[1,1].hist(idata.PetalLengthCm,bins=100)
8 plt.show()
```



```
In [197]: 1 idata.plot.area(y=['SepalLengthCm','SepalWidthCm','PetalLengthCm','PetalWidthCm'],alpha=0.4,fi
```



```
In [ ]:
```

```
1
```