

LETS GROW MORE INTERNSHIP DATA SCEINCE DEC BATCH TASK 1

ANURAG JADHAV

TASK 1 - IRIS FLOWERS CLASSIFICATION ML PROJECT

IMPORT LIBRARIES

```
In [10]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import holoviews as hv
import hvplot
import hvplot.pandas
import plotly.express as px
import scikitplot as skplt
import warnings
warnings.filterwarnings('ignore')
```

```
In [11]: iris = sns.load_dataset('iris')
iris.head()
```

```
Out[11]:
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa

```
In [11]: iris = sns.load_dataset('iris')
iris.head()
```

```
Out[11]:
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

Understanding Dataset

```
In [12]: iris.ndim
```

```
Out[12]: 2
```

```
In [13]: iris.shape
```

```
Out[13]: (150, 5)
```

```
In [14]: iris.columns
```

```
Out[14]: Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
               'species'],
              dtype='object')
```

```
In [15]: iris.isna().sum()
```

```
Out[15]: sepal_length    0
```

In [15]: iris.isna().sum()

Out[15]:

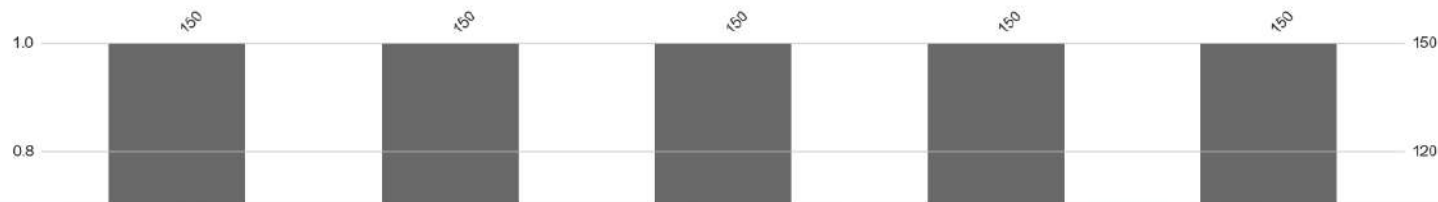
sepal_length	0
sepal_width	0
petal_length	0
petal_width	0
species	0
dtype:	int64

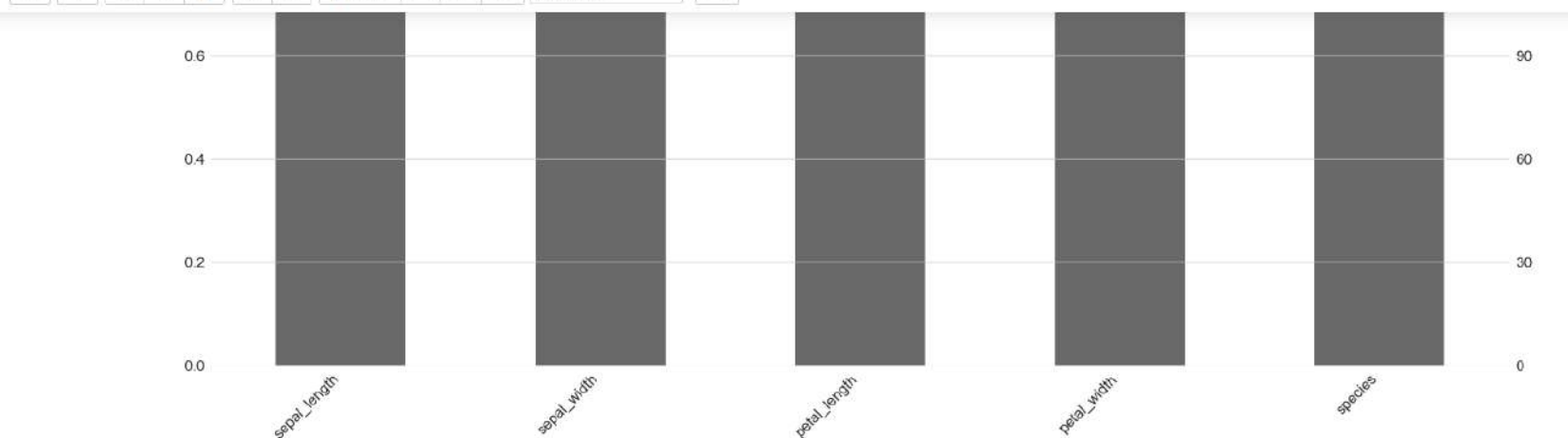
In [16]: iris.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column      Non-Null Count  Dtype
---  -
0   sepal_length  150 non-null    float64
1   sepal_width   150 non-null    float64
2   petal_length  150 non-null    float64
3   petal_width   150 non-null    float64
4   species       150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

In [17]:

```
import missingno as msno
msno.bar(iris)
plt.show()
```





```
In [18]: iris.species.value_counts()
```

```
Out[18]: setosa      50  
versicolor  50  
virginica    50  
Name: species, dtype: int64
```

```
In [19]: iris.species.value_counts(normalize=True)
```

```
Out[19]: setosa      0.333333  
versicolor  0.333333  
virginica    0.333333  
Name: species, dtype: float64
```

```
In [20]: iris.groupby('species').mean()
```

```
Out[20]:
```

	sepal_length	sepal_width	petal_length	petal_width
species				
setosa	5.006	3.428	1.462	0.246
versicolor	5.936	2.770	4.260	1.326
virginica	6.588	2.974	5.552	2.026

In [21]: iris.groupby('species').median()

Out[21]:

	sepal_length	sepal_width	petal_length	petal_width
species				
setosa	5.0	3.4	1.50	0.2
versicolor	5.9	2.8	4.35	1.3
virginica	6.5	3.0	5.55	2.0

In [22]: iris.groupby('species').std()

Out[22]:

	sepal_length	sepal_width	petal_length	petal_width
species				
setosa	0.352490	0.379064	0.173634	0.105386
versicolor	0.516171	0.313798	0.469911	0.197753
virginica	0.635880	0.322497	0.551895	0.274650

In [23]: iris.groupby('species').var()

Out[23]:

	sepal_length	sepal_width	petal_length	petal_width
species				

setosa	0.124249	0.143690	0.030159	0.011106
versicolor	0.266433	0.096469	0.220816	0.039106
virginica	0.404343	0.104004	0.304588	0.075433

In [24]: iris.describe()

Out[24]:

	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.057333	3.758000	1.199333
std	0.828066	0.435866	1.765298	0.762238
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

Visual Data Representation

TABLE

In [25]: iris.hvplot.table()

Out[25]:

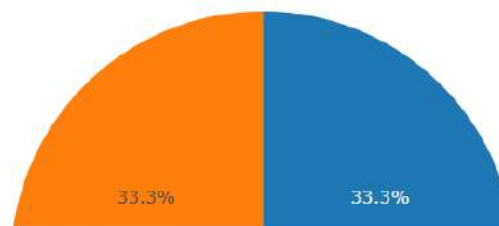
#	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa

2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa
5	5.4	3.9	1.7	0.4	setosa
6	4.6	3.4	1.4	0.3	setosa
7	5.0	3.4	1.5	0.2	setosa
8	4.4	2.9	1.4	0.2	setosa
9	4.9	3.1	1.5	0.1	setosa
10	5.4	3.7	1.5	0.2	setosa

Pie Plot

```
In [26]: fig = px.pie(iris, values = iris.species.value_counts(),  
                  title="Pie Chart of Species count", template="none")  
fig.show()
```

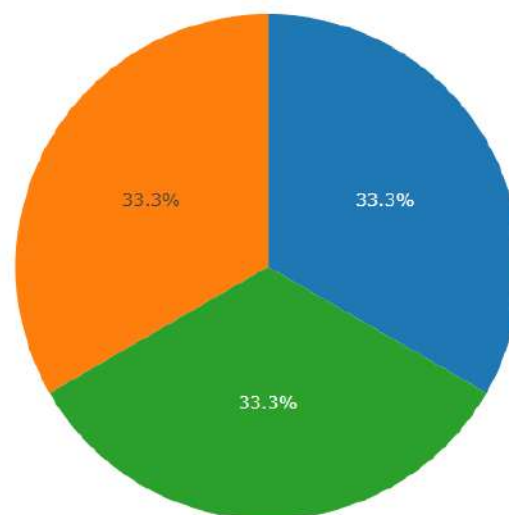
Pie Chart of Species count



Pie Plot

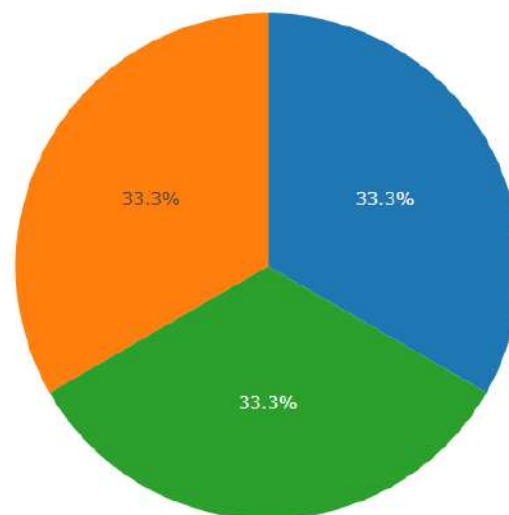
```
In [26]: fig = px.pie(iris, values = iris.species.value_counts(),  
                  title="Pie chart of Species count", template="none")  
fig.show()
```

Pie Chart of Species count



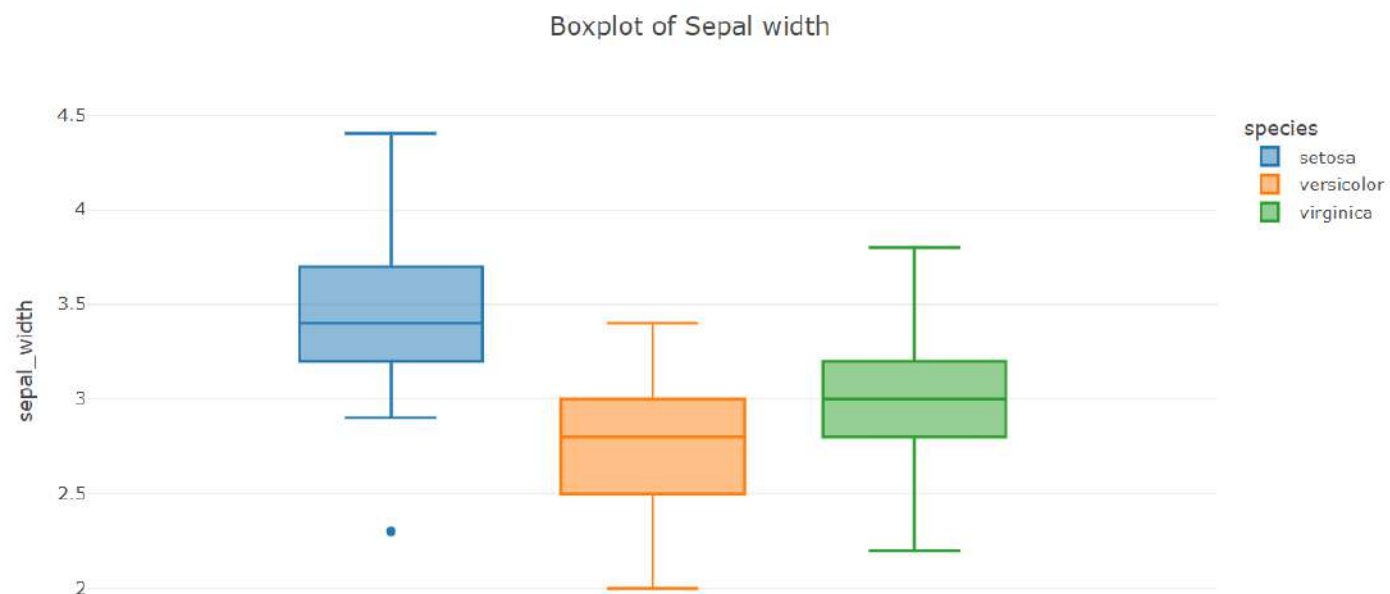

```
In [27]: fig = px.pie(iris, values = iris.species.value_counts(),  
              title="Pie chart of Species count", template="none")  
fig.show()
```

Pie Chart of Species count

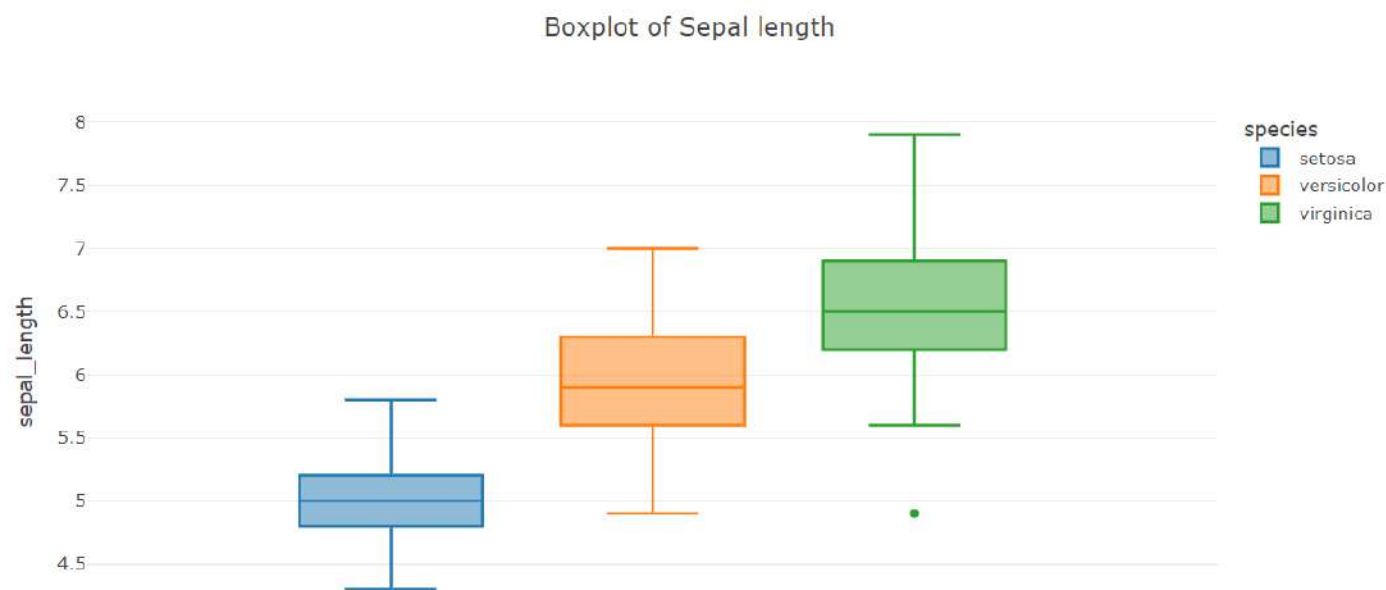


Boxplot

```
In [28]: fig = px.box(iris, y="sepal_width", color="species", title="Boxplot of Sepal width", template="none")
fig.show()
```

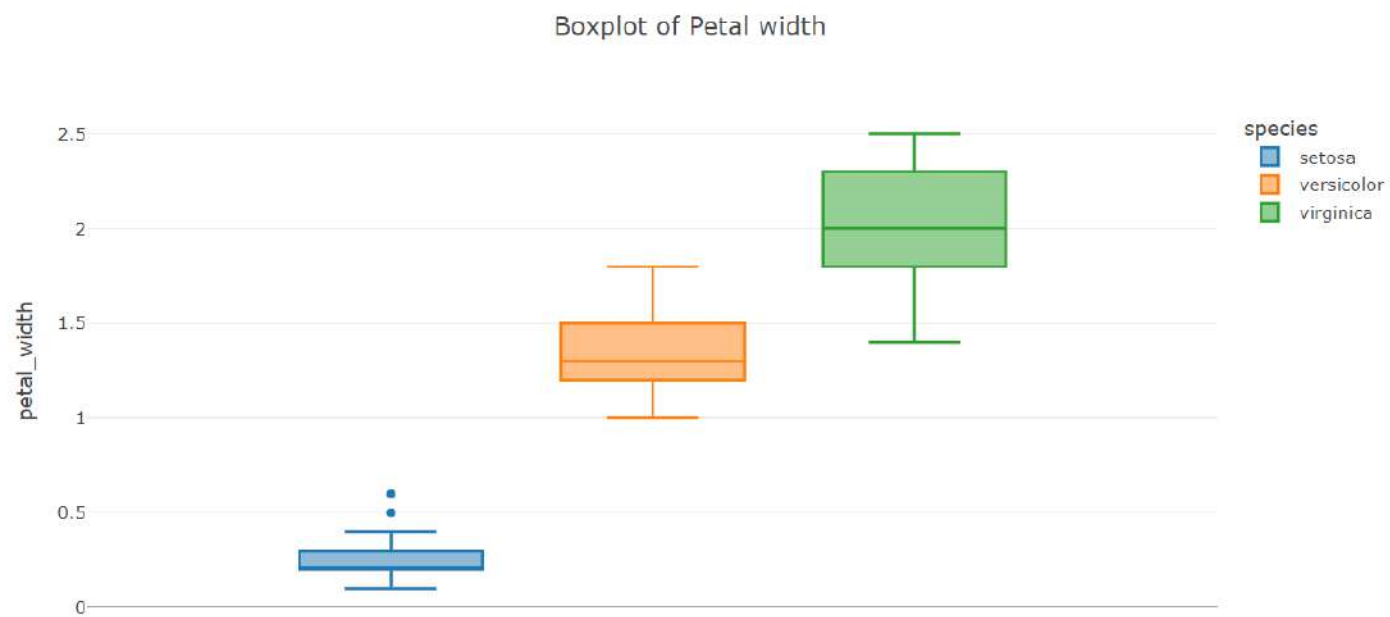


```
In [29]: fig = px.box(iris, y="sepal_length", color="species", title="Boxplot of Sepal length", template="none")
fig.show()
```

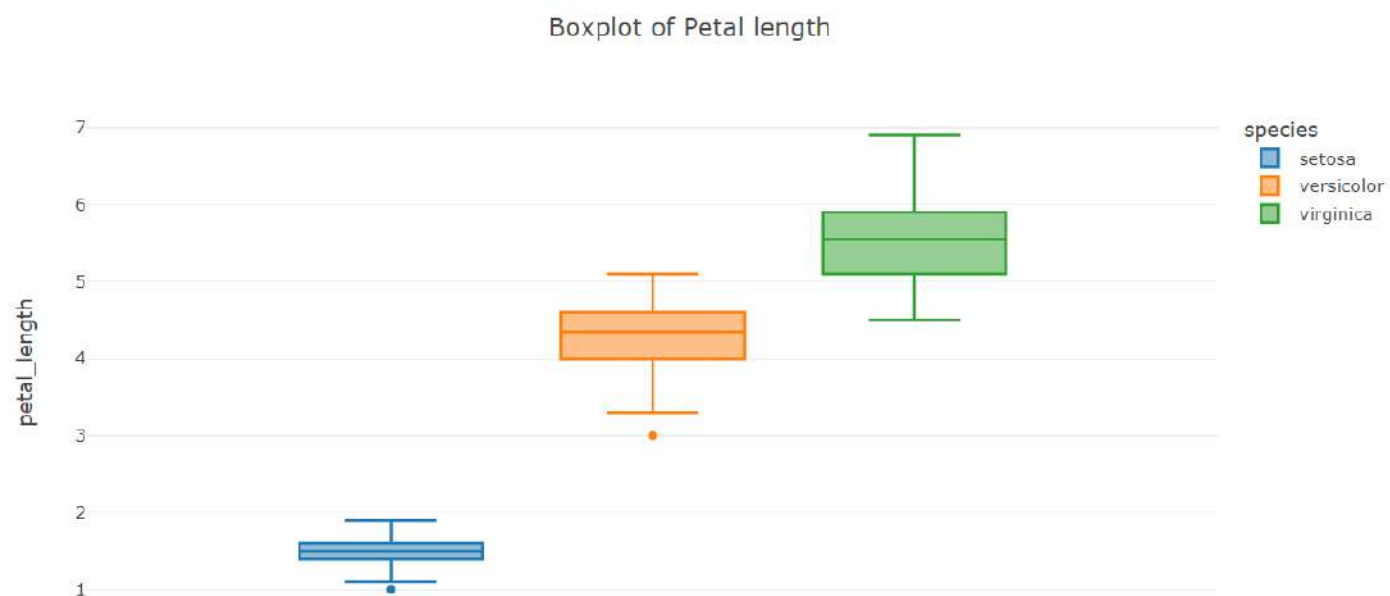


```
In [30]: fig = px.box(iris, y="petal_width", color="species", title="Boxplot of Petal width", template="none")
```

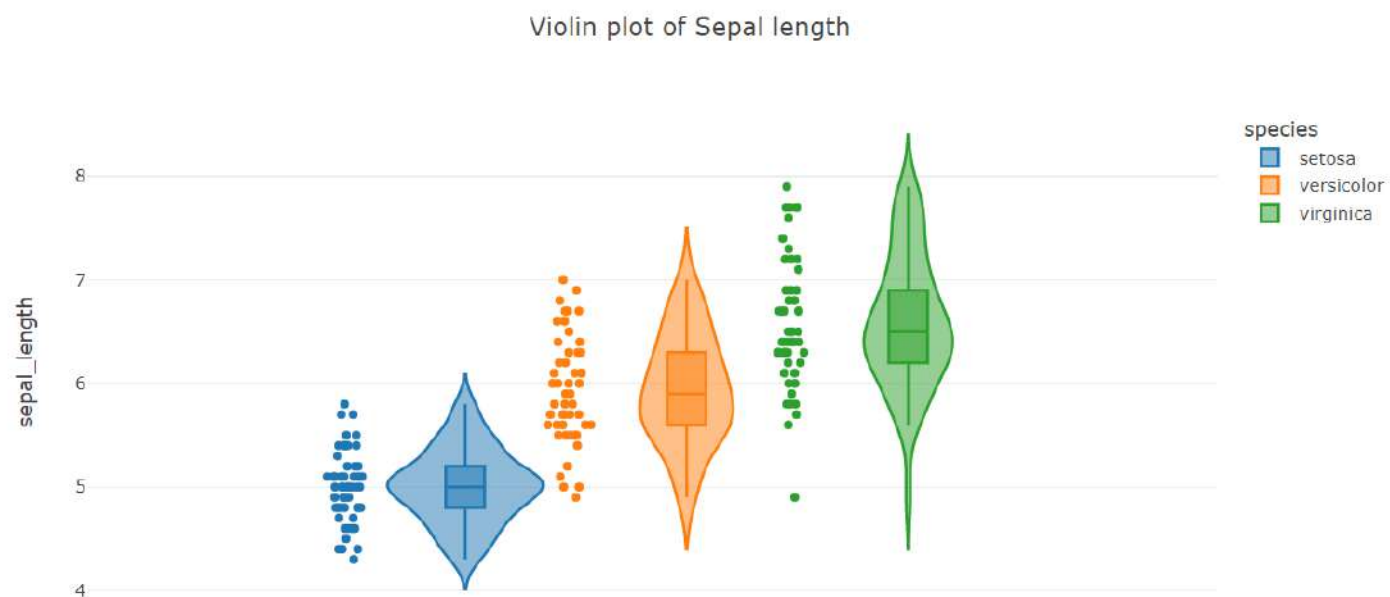
```
In [30]: fig = px.box(iris, y="petal_width", color="species", title="Boxplot of Petal width", template="none")
fig.show()
```



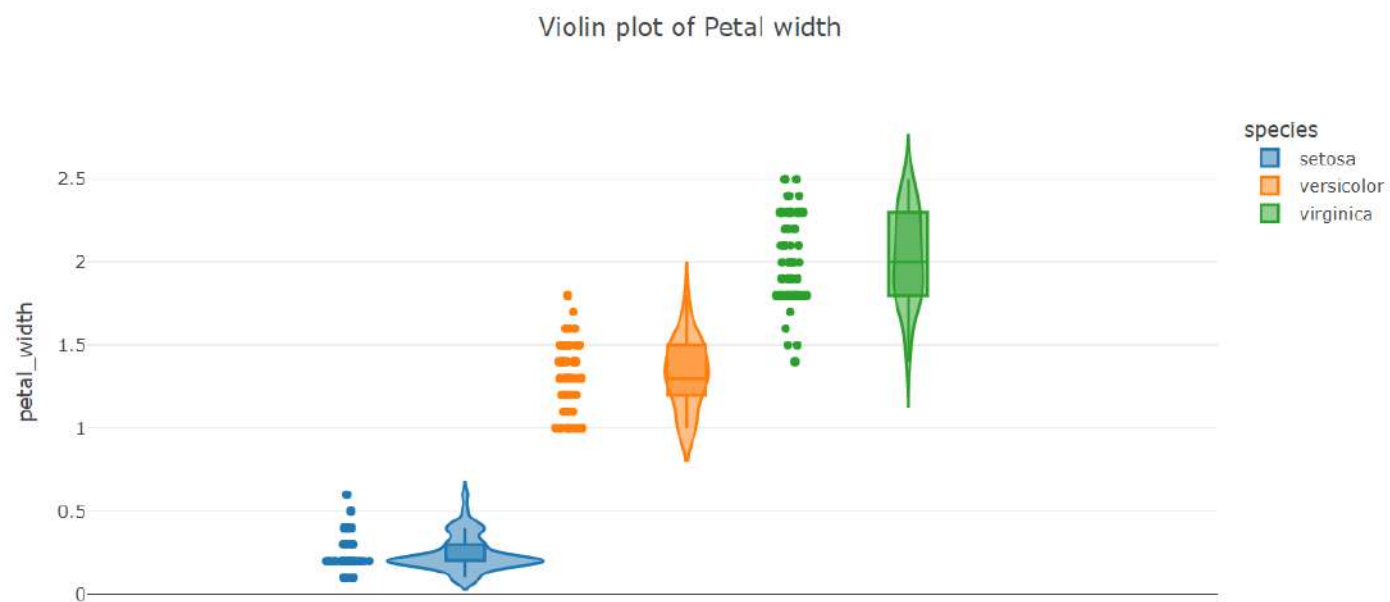
```
In [31]: fig = px.box(iris, y="petal_length", color="species", title="Boxplot of Petal length", template="none")
fig.show()
```



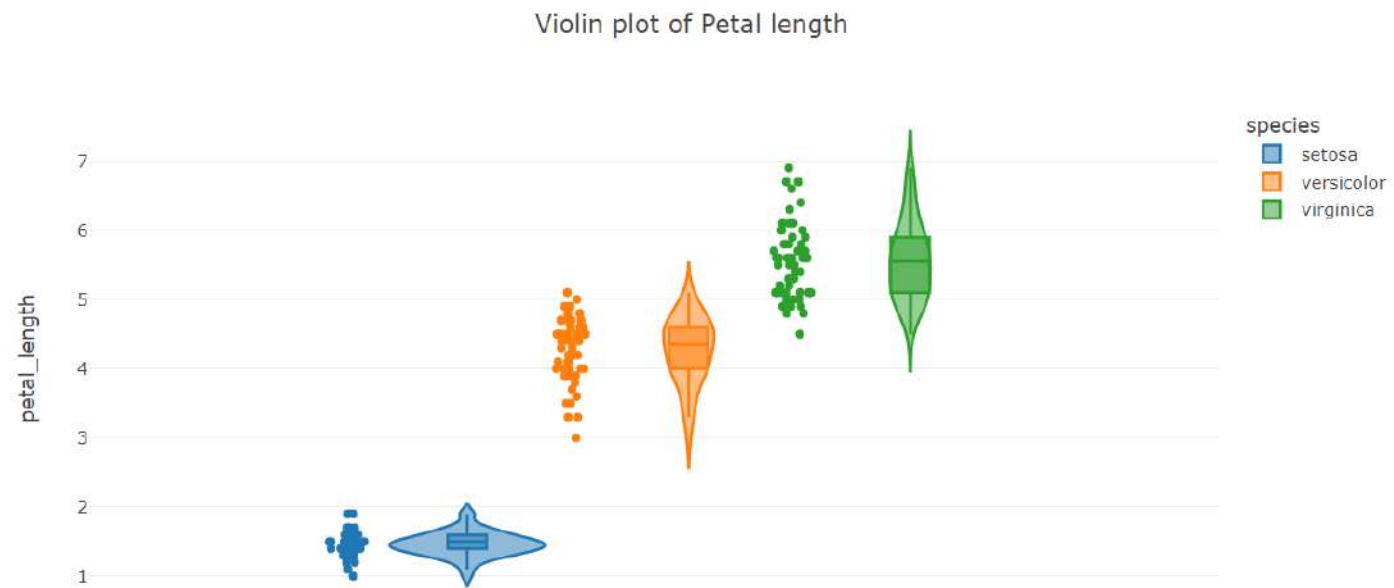
```
In [33]: fig = px.violin(iris, y="sepal_length", box=True, points='all', color="species",
                    ,title="Violin plot of Sepal length",template="none")
fig.show()
```




```
In [34]: fig = px.violin(iris, y="petal_width", box=True, points='all', color="species",
                title="Violin plot of Petal width", template="none")
fig.show()
```



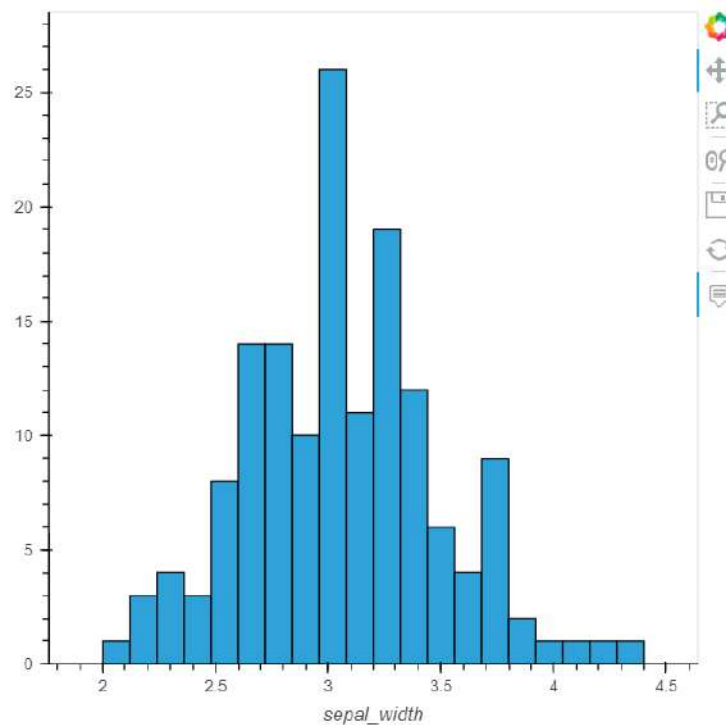
```
In [35]: fig = px.violin(iris, y="petal_length", box=True, points='all', color="species",
                    title="Violin plot of Petal length", template="none")
fig.show()
```



Histograms

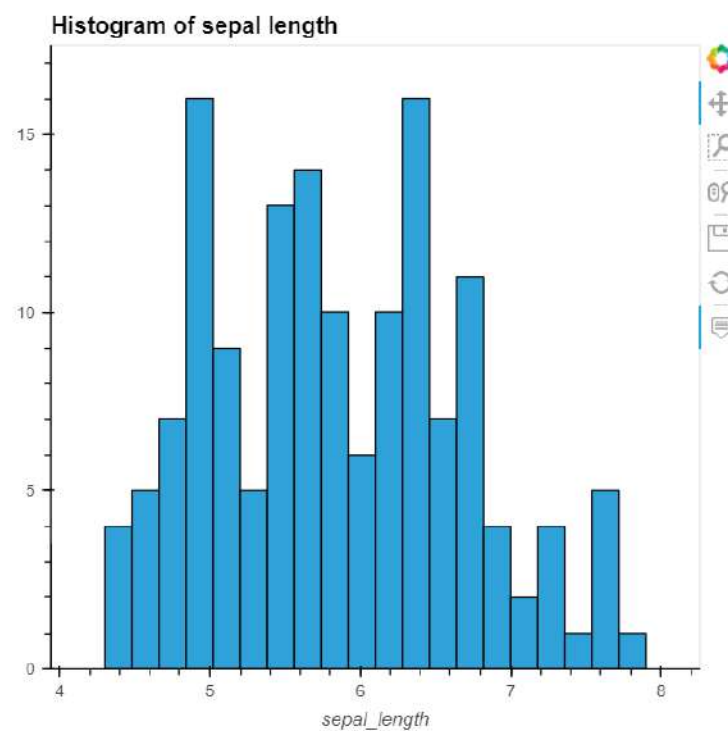
```
In [36]: iris.hvplot.hist("sepal_width",width=500, height=500)
```

Out[36]:



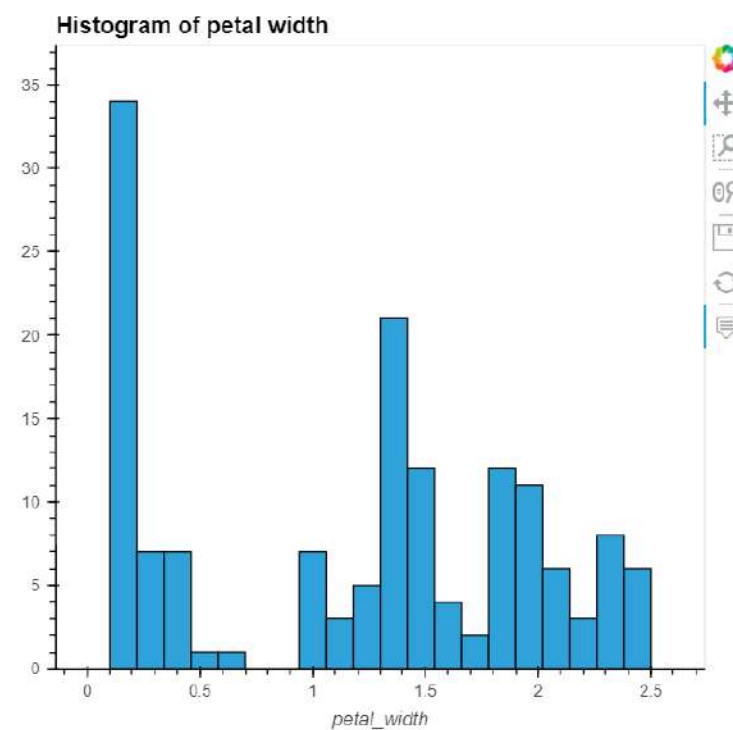
```
In [37]: iris.hvplot.hist("sepal_length",width=500, height=500,title="Histogram of sepal length")
```

Out[37]:



```
In [38]: iris.hvplot.hist("petal_width",width=500, height=500,title="Histogram of petal width")
```

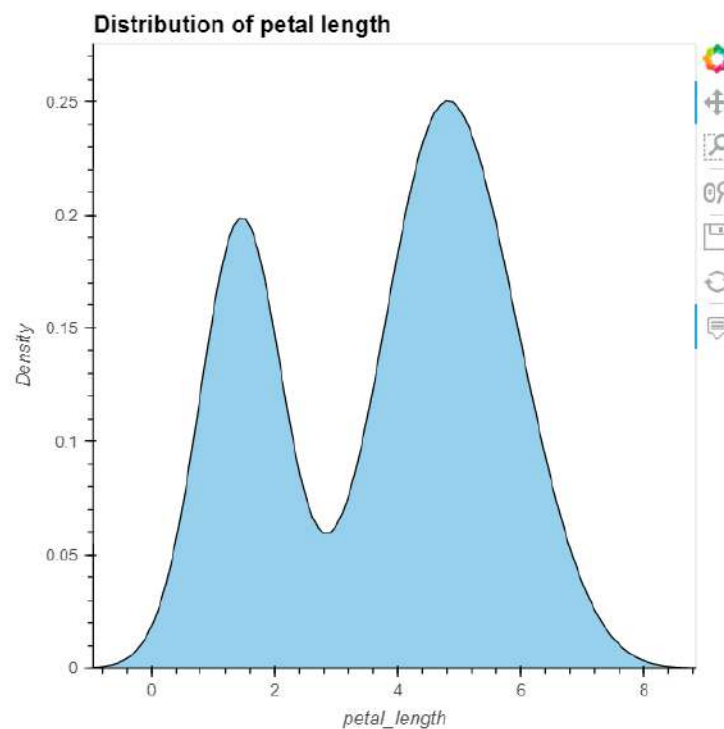
Out[38]:



```
In [39]: iris.hvplot.kde("petal_length",width=500, height=500,title="Distribution of petal length")
```

```
In [39]: iris.hvplot.kde("petal_length",width=500,height=500,title="Distribution of petal length")
```

Out[39]:

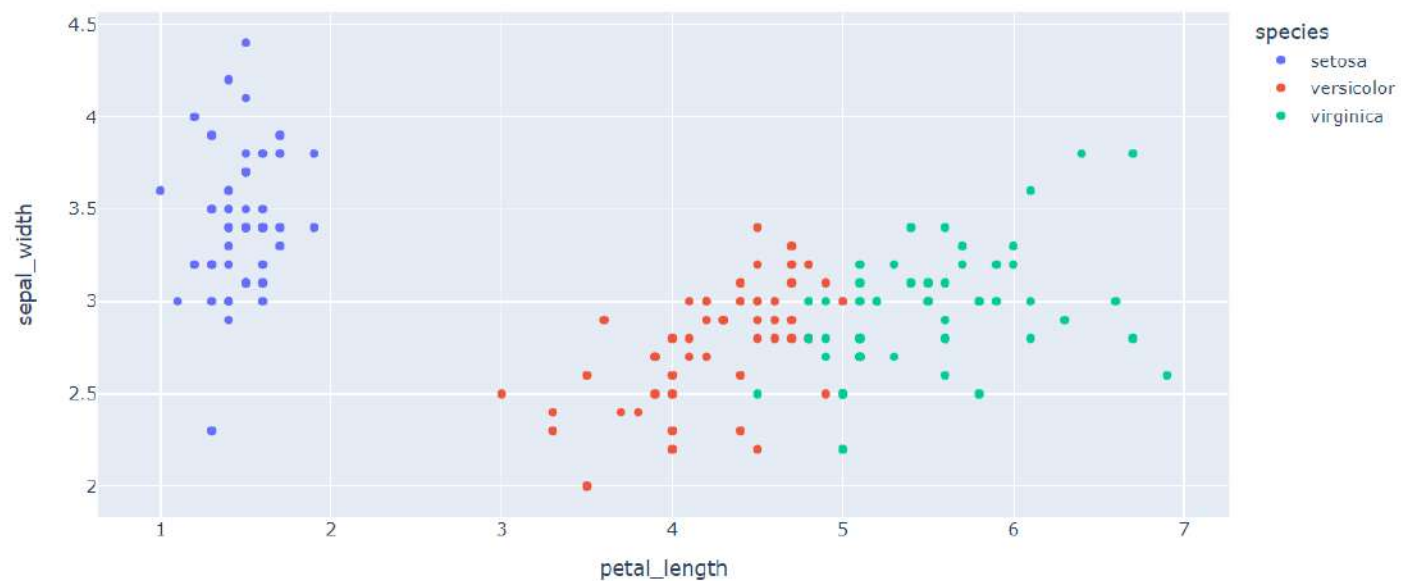


Scatter Plots

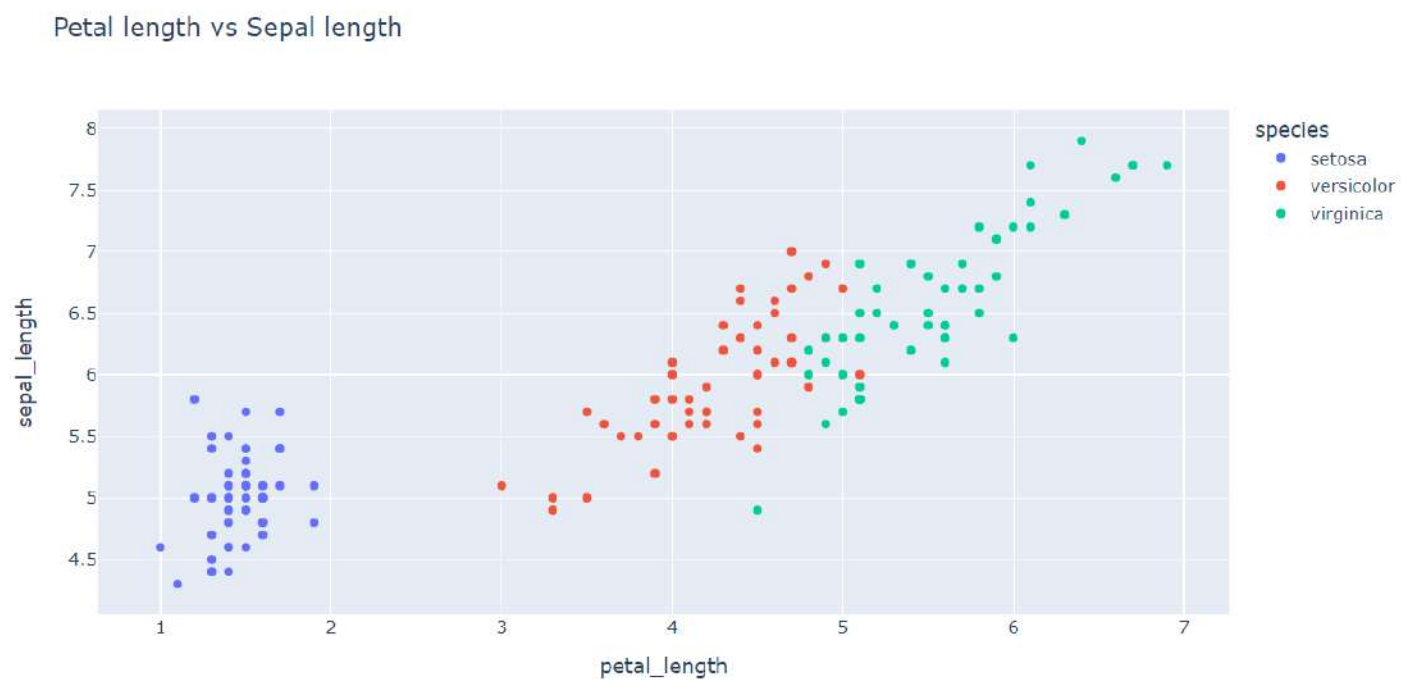
Scatter Plots

```
In [40]: px.scatter(iris,x="petal_length", y="sepal_width", color="species",  
                  title="Petal length vs Sepal Width")
```

Petal length vs Sepal Width

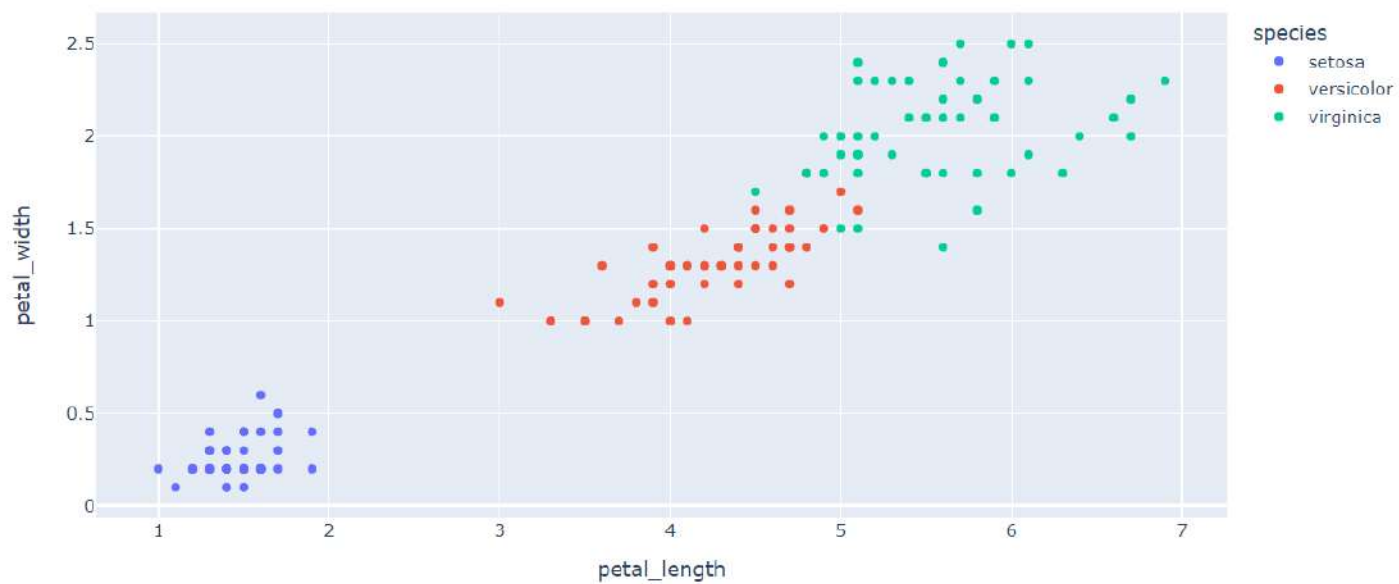


```
In [41]: px.scatter(iris,x="petal_length", y="sepal_length", color="species",  
                title="Petal length vs Sepal length")
```



```
In [42]: px.scatter(iris,x="petal_length", y="petal_width", color="species",  
               title="Petal length vs Petal width")
```

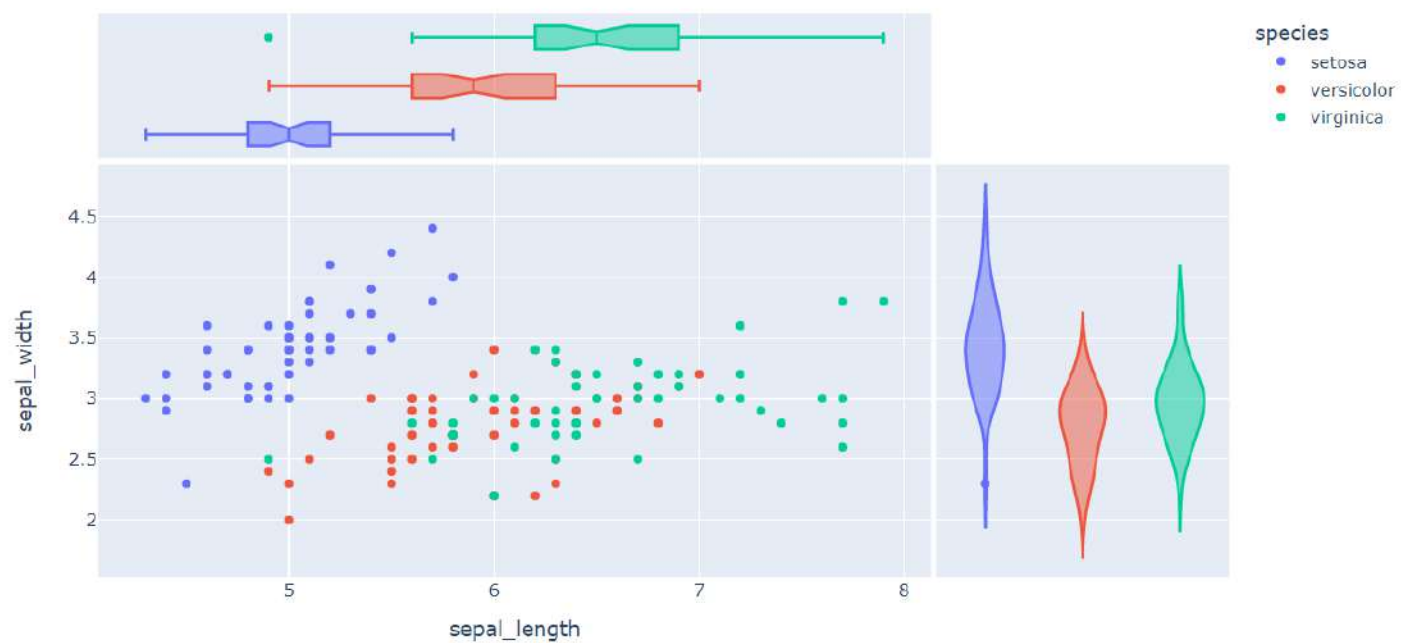
Petal length vs Petal width



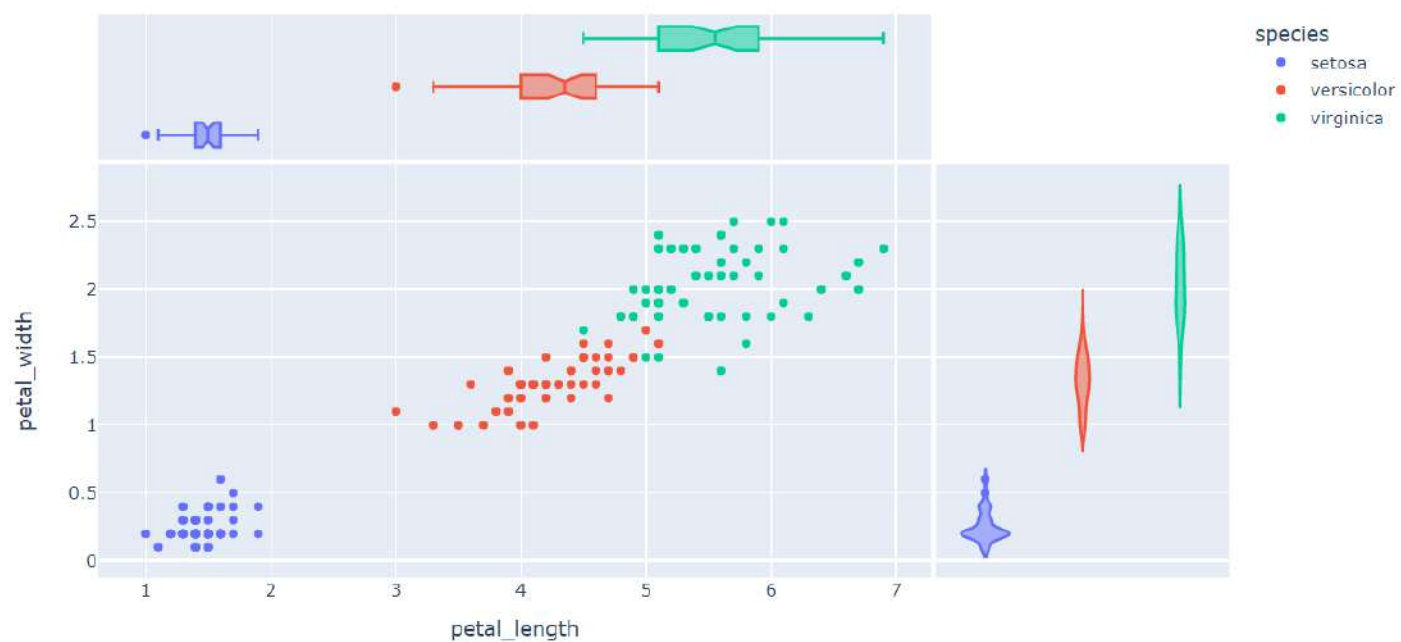
```
In [43]: fig = px.scatter(iris, x="sepal_width", y="sepal_length", color="species",  
                        size='petal_length', hover_data=['petal_width'])  
fig.show()
```



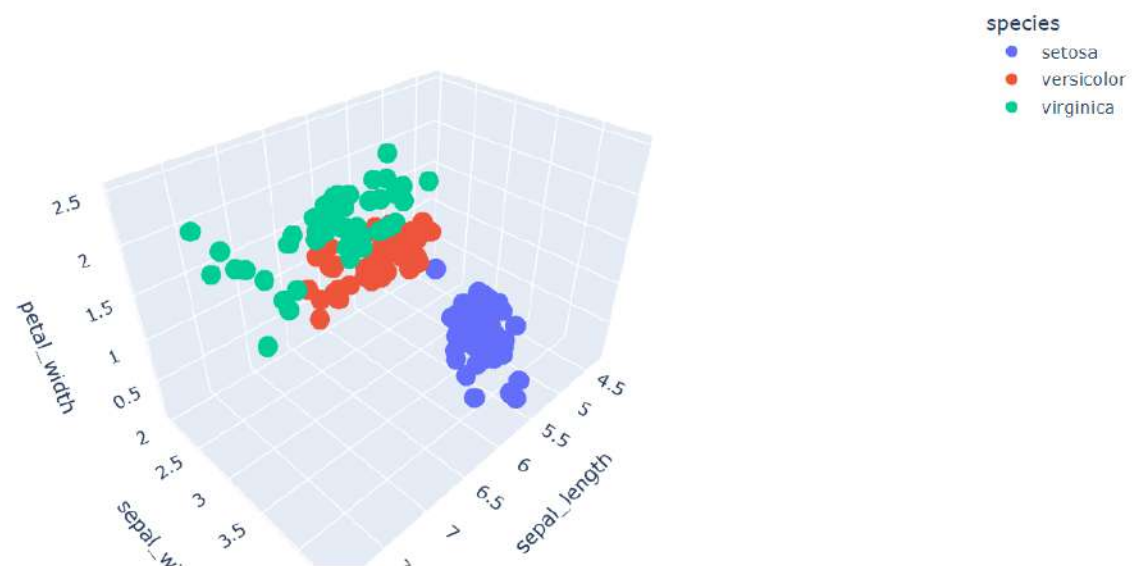
```
In [44]: fig = px.scatter(iris, x="sepal_length", y="sepal_width", color="species",  
                    marginal_x="box", marginal_y="violin")  
fig.show()
```



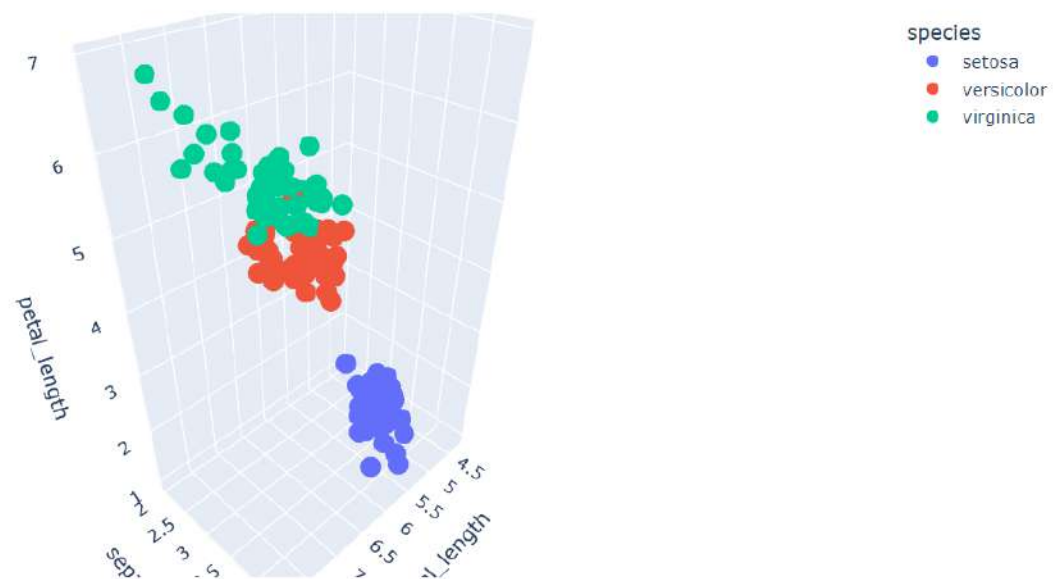
```
In [45]: fig = px.scatter(iris, x="petal_length", y="petal_width", color="species",  
                        marginal_x="box", marginal_y="violin")  
fig.show()
```




```
In [46]: fig = px.scatter_3d(iris, x='sepal_length', y='sepal_width', z='petal_width',  
                             color='species')  
fig.show()
```



```
In [47]: fig = px.scatter_3d(iris, x='sepal_length', y='sepal_width', z='petal_length',  
                             color='species')  
fig.show()
```



Correlation heatmap

In [48]: `iris.corr()`

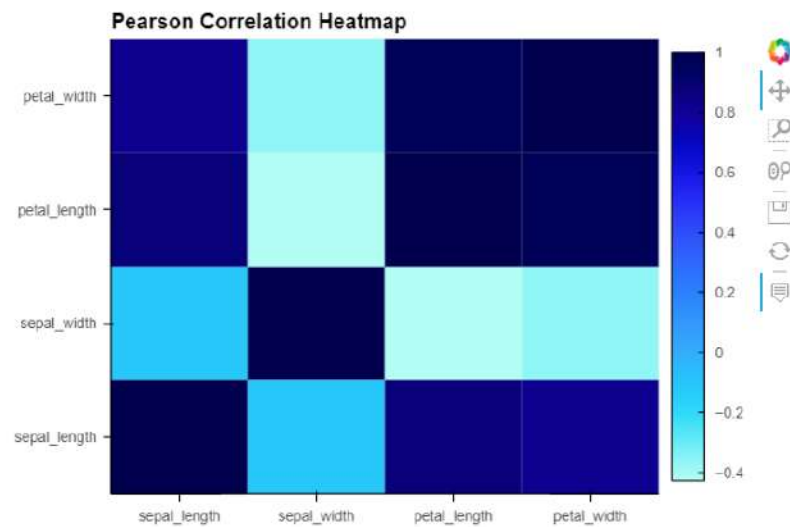
Out[48]:

	sepal_length	sepal_width	petal_length	petal_width
sepal_length	1.000000	-0.117570	0.871754	0.817941
sepal_width	-0.117570	1.000000	-0.426440	-0.366126
petal_length	0.871754	-0.426440	1.000000	0.962865
petal_width	0.817941	-0.366126	0.962865	1.000000

In [49]: `iris.corr().hvplot.heatmap(height=400, width=600,`

`colorbar=True, title = "Pearson Correlation Heatmap")`

Out[49]:



Feature Encoding

```
In [50]: from sklearn.metrics import confusion_matrix, classification_report, accuracy_score
from sklearn.metrics import f1_score, recall_score, precision_score
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder, StandardScaler
from yellowbrick.target import FeatureCorrelation, BalancedBinningReference, ClassBalance
classes = ["Setosa", "Versicolour", "Virginica"]
```

```
In [51]: label_encoder = LabelEncoder()
iris['species'] = label_encoder.fit_transform(iris['species'])
iris['species'].unique()
```

```
Out[51]: array([0, 1, 2])
```

```
In [52]: iris.head()
```

```
Out[52]:
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

```
In [53]: X = iris.iloc[:, :-1].values
y = iris.iloc[:, -1].values
```

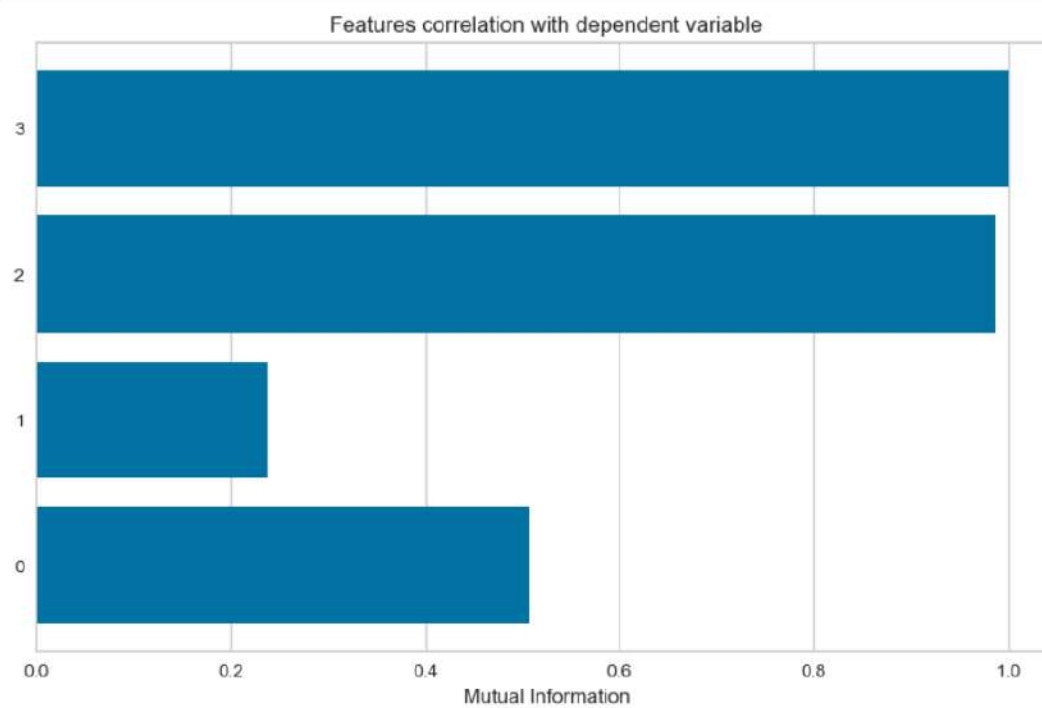
```
In [54]: X[:5]
```

```
Out[54]: array([[5.1, 3.5, 1.4, 0.2],
[4.9, 3. , 1.4, 0.2],
[4.7, 3.2, 1.3, 0.2],
[4.6, 3.1, 1.5, 0.2],
[5. , 3.6, 1.4, 0.2]])
```

```
In [55]: y[:5]
```

Out[55]: array([0, 0, 0, 0, 0])

```
In [56]: plt.figure(figsize=(10,6))
visualizer = FeatureCorrelation(method='mutual_info-classification')
visualizer.fit(X, y)
visualizer.show()
plt.show()
```





```
In [57]: scaler = StandardScaler()
X = scaler.fit_transform(X)
print(X[:5])

[[-0.90068117  1.01900435 -1.34022653 -1.3154443 ]
 [-1.14301691 -0.13197948 -1.34022653 -1.3154443 ]
 [-1.38535265  0.32841405 -1.39706395 -1.3154443 ]
 [-1.50652052  0.09821729 -1.2833891  -1.3154443 ]
 [-1.02184904  1.24920112 -1.34022653 -1.3154443 ]]
```

```
In [58]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)
```

K-Nearest Neighbours

```
In [59]: from sklearn.neighbors import KNeighborsClassifier
knn_clf = KNeighborsClassifier(n_neighbors=3)
knn_clf.fit(X_train, y_train)
```

```
Out[59]: KNeighborsClassifier(n_neighbors=3)
```

```
In [60]: y_pred = knn_clf.predict(X_test)
```

Classification Error:

```
In [61]: print(1 - accuracy_score(y_test, y_pred))

0.033333333333333326
```

Precision Score:

```
In [62]: print(precision_score(y_test, y_pred, average="micro"))

0.9666666666666667
```

Classification Accuracy:


```
In [63]: print(accuracy_score(y_test, y_pred))
```

0.9666666666666667

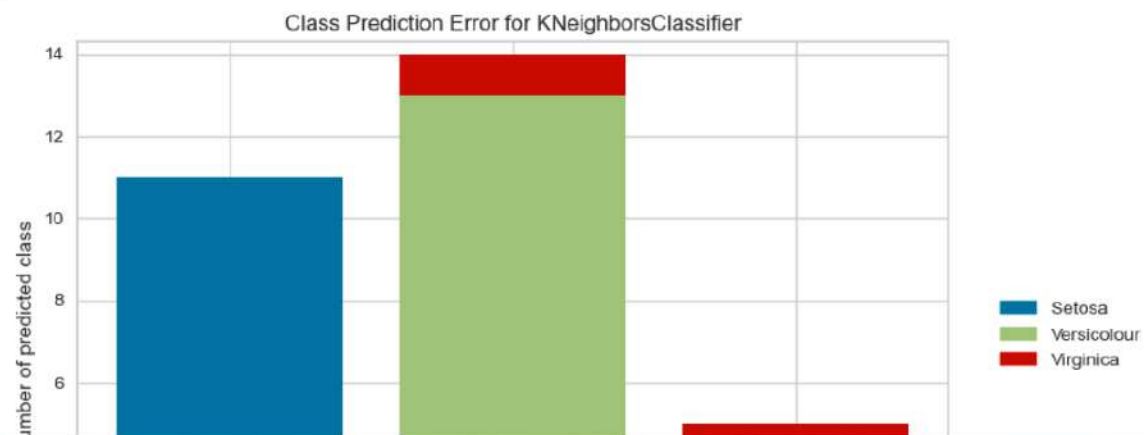
Sensitivity/True Positive Rate/Recall Score

```
In [64]: print(recall_score(y_test, y_pred, average="micro"))
```

0.9666666666666667

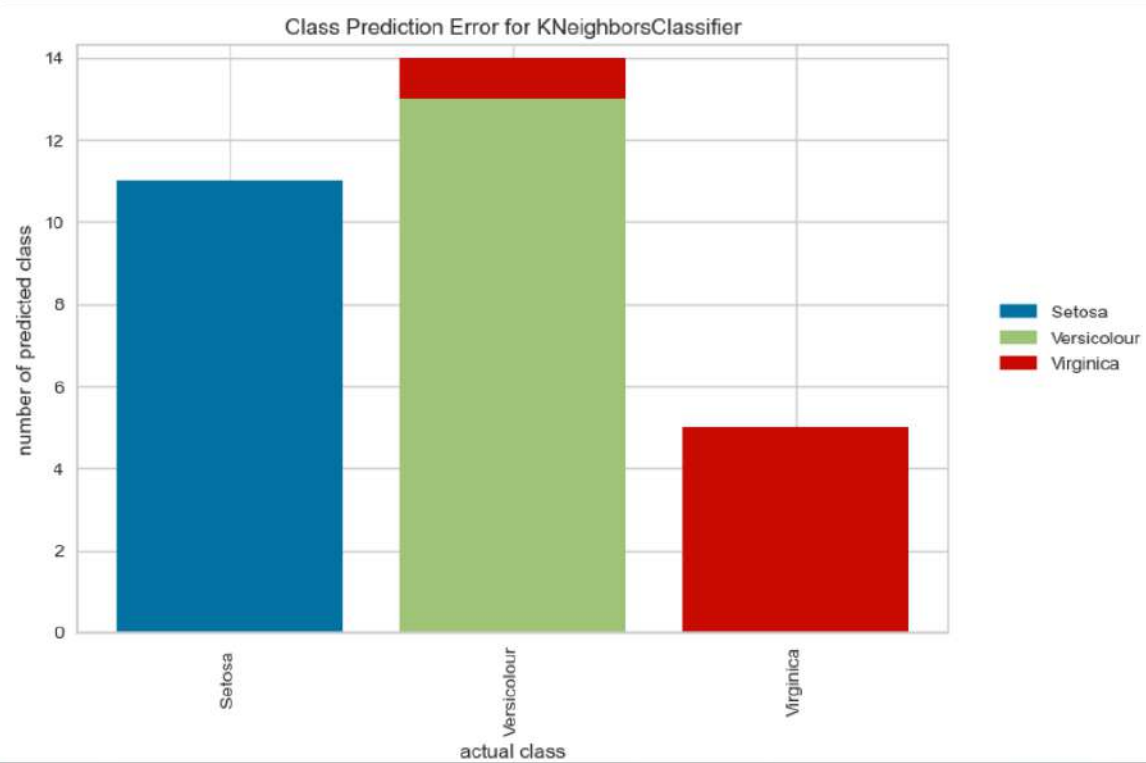
Class Prediction Error

```
In [65]: from yellowbrick.classifier import class_prediction_error
plt.figure(figsize=(10,6))
visualizer = class_prediction_error(knn_clf,X_train,y_train,X_test,y_test,classes=classes)
visualizer.show()
plt.show()
```



Class Prediction Error

```
In [65]: from yellowbrick.classifier import class_prediction_error
plt.figure(figsize=(10,6))
visualizer = class_prediction_error(knn_clf,X_train,y_train,X_test,y_test,classes=classes)
visualizer.show()
plt.show()
```

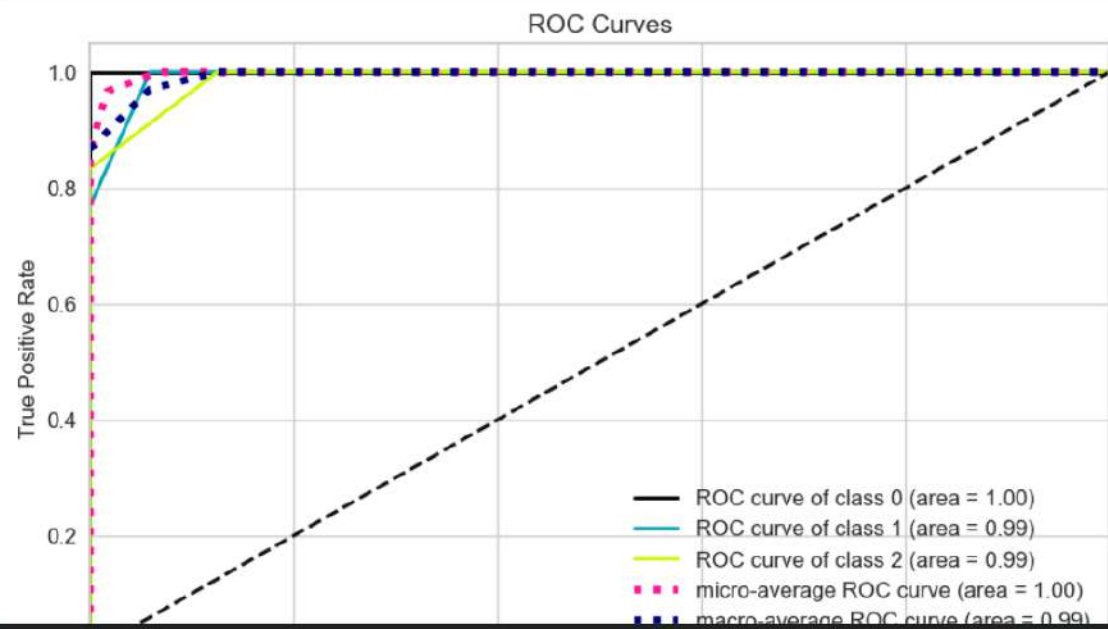


F1 Score

```
In [67]: print( f1_score( y_test, y_pred, average="micro" ))  
0.9666666666666667
```

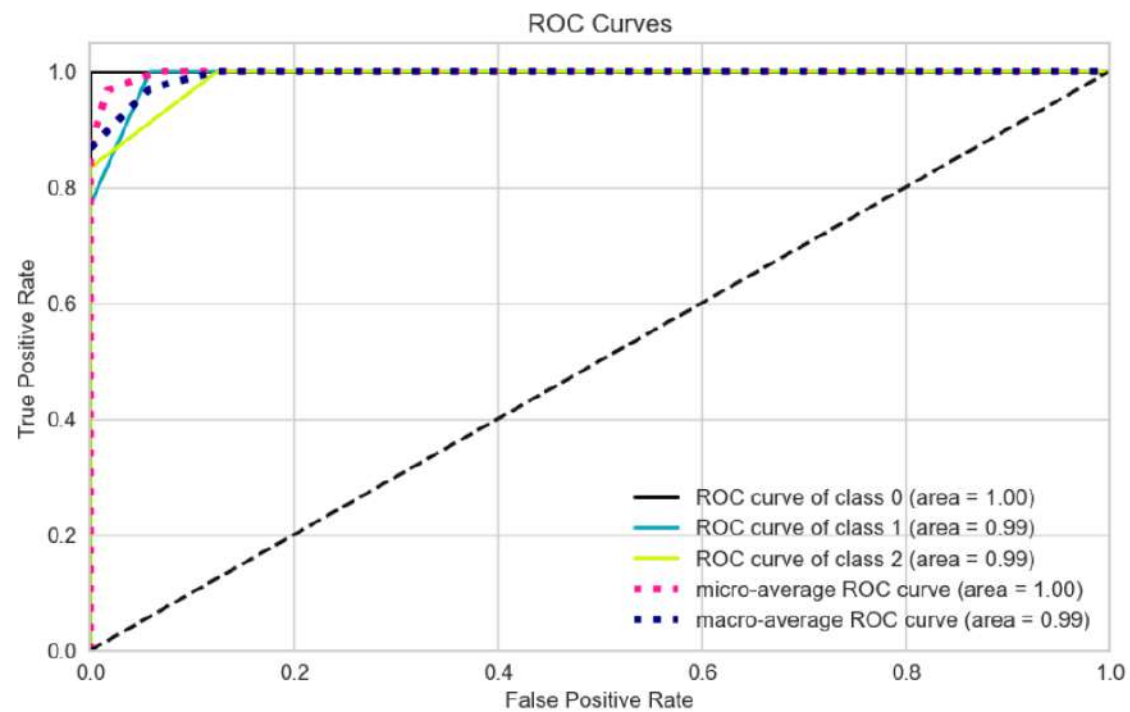
ROC Curve

```
In [68]: y_probas = knn_clf.predict_proba(X_test)  
skplt.metrics.plot_roc(y_test, y_probas, figsize=(10,6), title_fontsize=14, text_fontsize=12)  
plt.show()
```



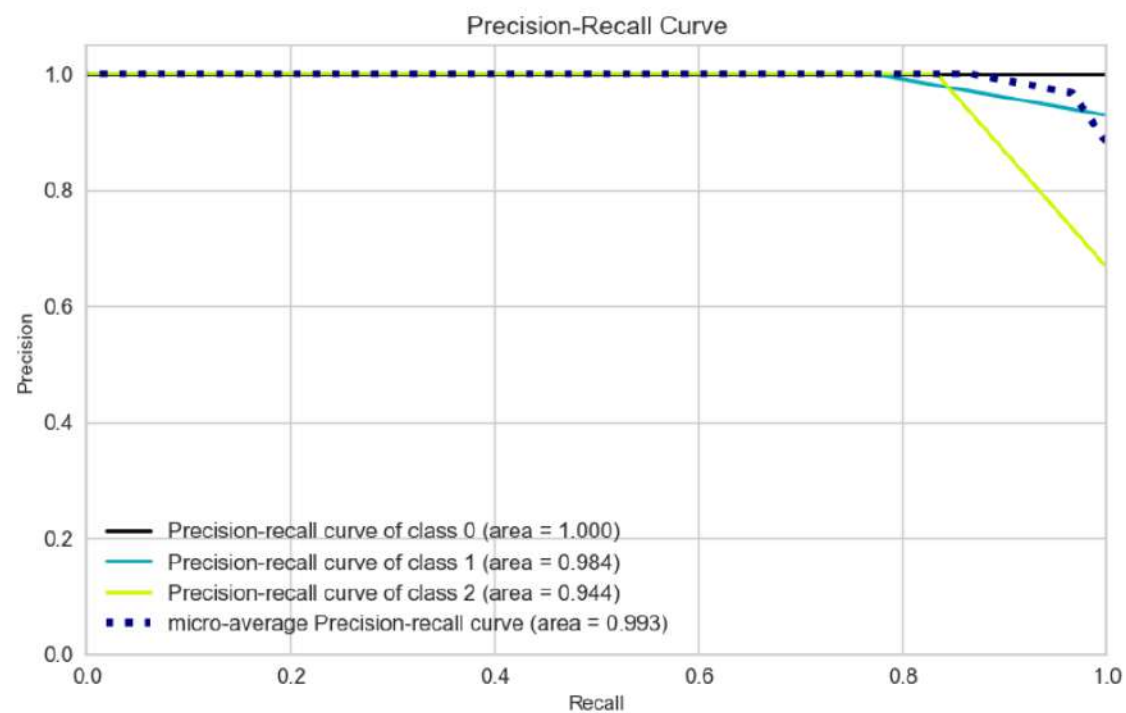
ROC Curve

```
In [68]: y_probas = knn_clf.predict_proba(X_test)
skplt.metrics.plot_roc(y_test, y_probas, figsize=(10,6), title_fontsize=14, text_fontsize=12)
plt.show()
```



Precision Recall Curve

```
In [69]: skplt.metrics.plot_precision_recall(y_test, y_probas, figsize=(10,6), title_fontsize=14, text_fontsize=12)
plt.show()
```



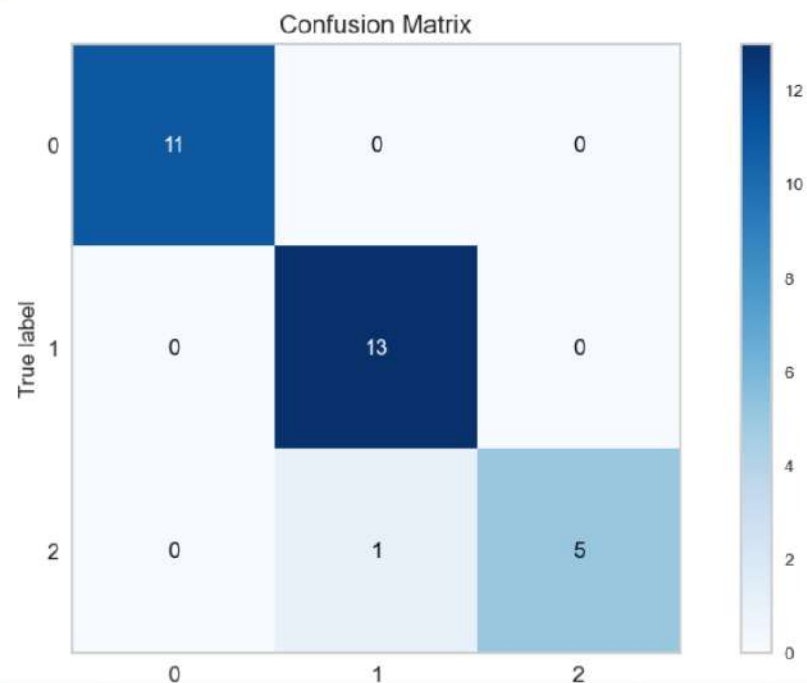
Confusion Matrix

Confusion Matrix

```
In [70]: confusion_matrix(y_test, y_pred)
```

```
Out[70]: array([[11,  0,  0],
               [ 0, 13,  0],
               [ 0,  1,  5]], dtype=int64)
```

```
In [71]: skplt.metrics.plot_confusion_matrix(y_test, y_pred, figsize=(12,6),title_fontsize=14,text_fontsize=12)
plt.show()
```

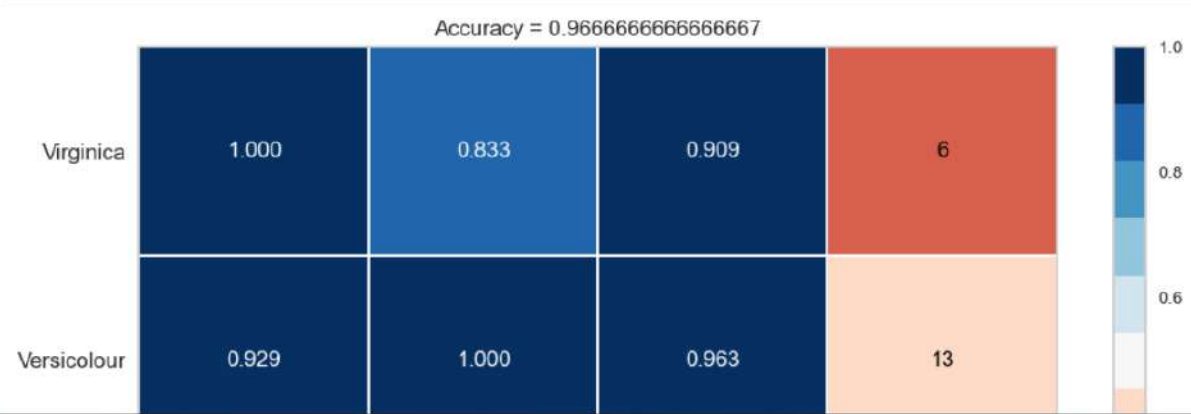


Classification Report

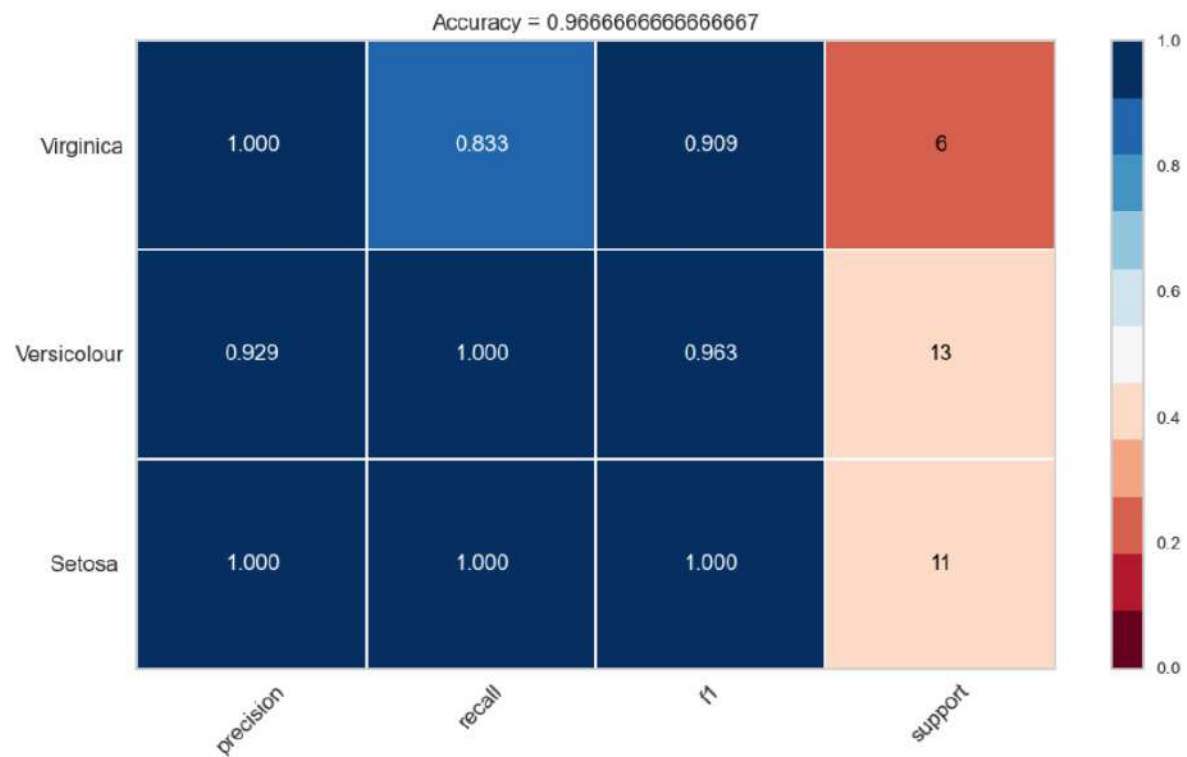
```
In [72]: print(classification_report(y_test,y_pred))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	11
1	0.93	1.00	0.96	13
2	1.00	0.83	0.91	6
accuracy			0.97	30
macro avg	0.98	0.94	0.96	30
weighted avg	0.97	0.97	0.97	30

```
In [73]: import yellowbrick as yb
plt.figure(figsize=(10,6))
visualizer = yb.classifier.classification_report(
    knn_clf, X_train, y_train, X_test, y_test, classes=classes, support=True, cmap='RdBu',
    title = 'Accuracy = ' + str(accuracy_score(y_test,y_pred))
)
plt.show()
```




```
In [73]: import yellowbrick as yb
plt.figure(figsize=(10,6))
visualizer = yb.classifier.classification_report(
    knn_clf, X_train, y_train, X_test, y_test, classes=classes, support=True, cmap='RdBu',
    title = 'Accuracy = ' + str(accuracy_score(y_test,y_pred))
)
plt.show()
```





Saving ML Models

```
In [74]: import joblib
from os import listdir
joblib.dump(knn_clf, knn_clf.__class__.__name__ + '.pkl')
for file in listdir():
    if file.endswith('.pkl'):
        print(file)
```

KNeighborsClassifier.pkl

In []: