```
In [13]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

Cleaning the data

```
In [14]: df = pd.read_csv('C:/Users/computer world/OneDrive/Desktop/IRIS.csv')
```

In [15]: df.head()

Out[15]:

	sepal_length	sepal_width	petal_length	petal_width	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 [16]: df.tail()

Out[16]:

species	petal_width	petal_length	sepal_width	sepal_length	
Iris-virginica	2.3	5.2	3.0	6.7	145
Iris-virginica	1.9	5.0	2.5	6.3	146
Iris-virginica	2.0	5.2	3.0	6.5	147
Iris-virginica	2.3	5.4	3.4	6.2	148
Iris-virginica	1.8	5.1	3.0	5.9	149

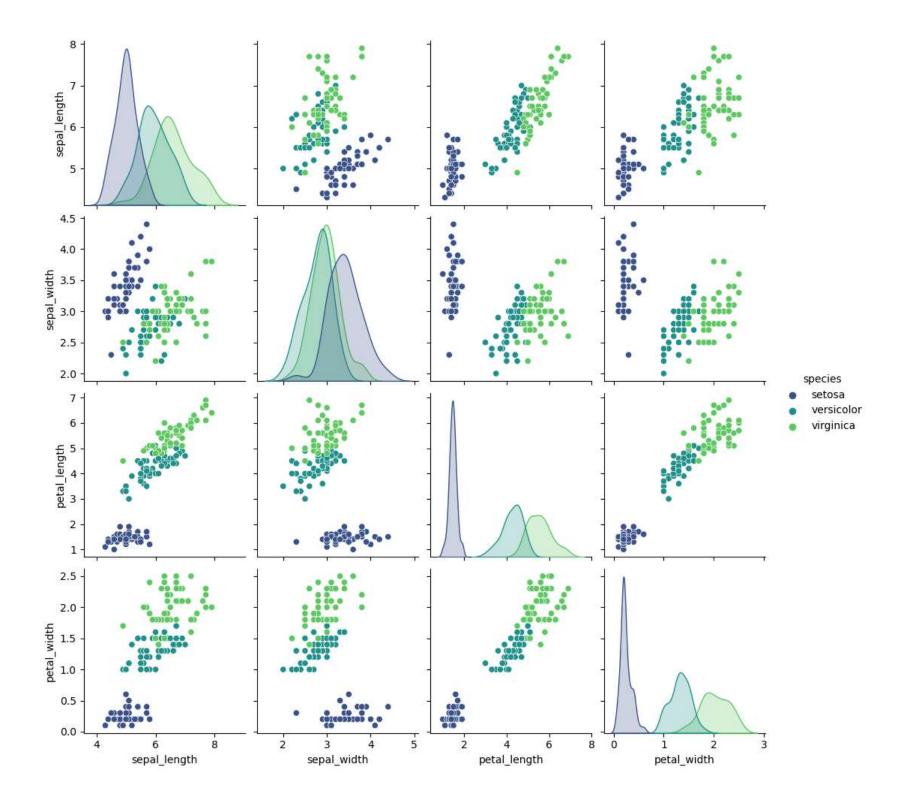
```
In [17]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 150 entries, 0 to 149
         Data columns (total 5 columns):
              Column
                            Non-Null Count Dtype
              sepal_length 150 non-null
                                            float64
          1 sepal width 150 non-null
                                           float64
              petal length 150 non-null
                                           float64
              petal_width 150 non-null
                                           float64
              species
                                           object
          4
                            150 non-null
         dtypes: float64(4), object(1)
         memory usage: 6.0+ KB
In [18]: df['species'].value_counts()
Out[18]: Iris-setosa
                            50
         Iris-versicolor
                            50
         Iris-virginica
                            50
         Name: species, dtype: int64
In [19]: df['species'] = df['species'].str.replace('Iris-','')
In [20]: df['species'].value counts()
Out[20]: setosa
                       50
         versicolor
                       50
         virginica
                       50
         Name: species, dtype: int64
```

```
In [21]: missing_values = df.isnull().sum()
    percentage_missing = (missing_values/len(df))*100
    pd.DataFrame({'missing_values':missing_values,'percentage_missing':percentage_missing})
```

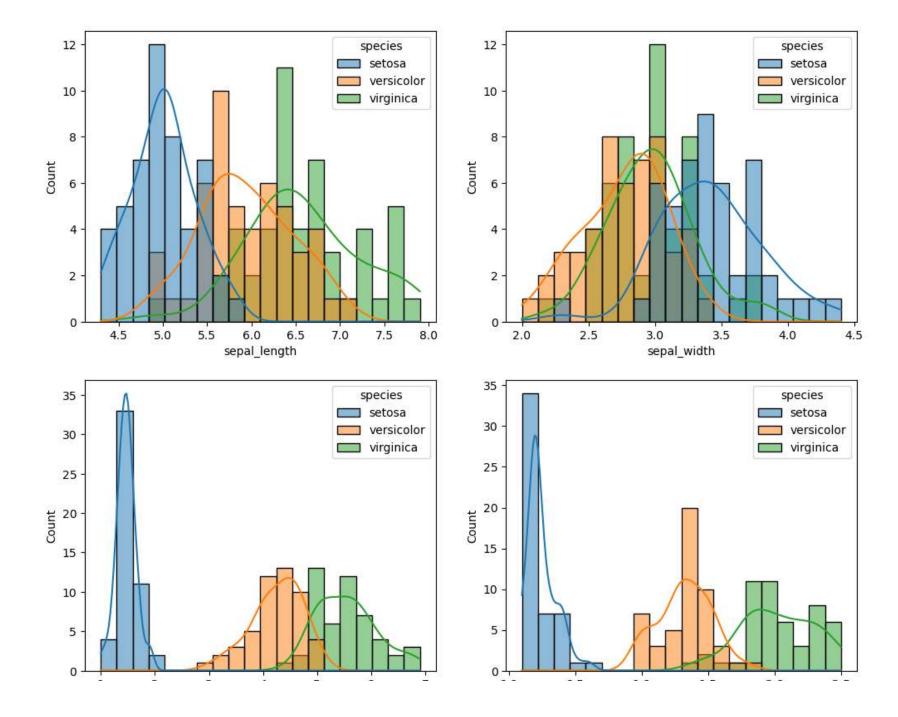
Out[21]:		missing_values	percentage_missing
	sepal_length	0	0.0
	sepal_width	0	0.0
	petal_length	0	0.0
	petal_width	0	0.0
	species	0	0.0

Data visualization

```
In [12]: green_palette = sns.color_palette("viridis",n_colors = 3)
    sns.pairplot(df,hue='species',palette=green_palette)
    plt.show()
```

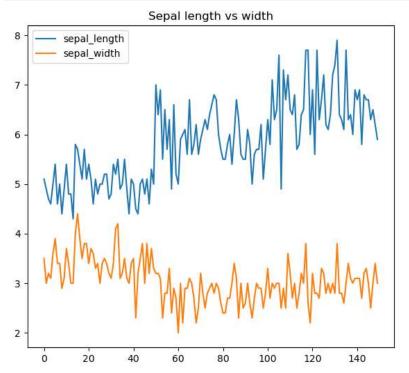


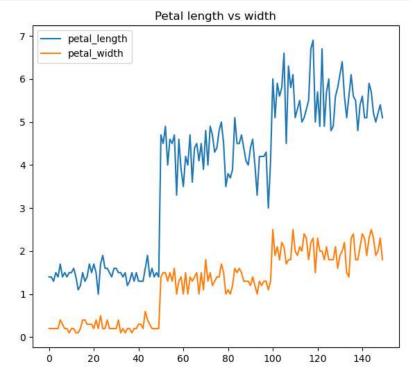
Histograms of features



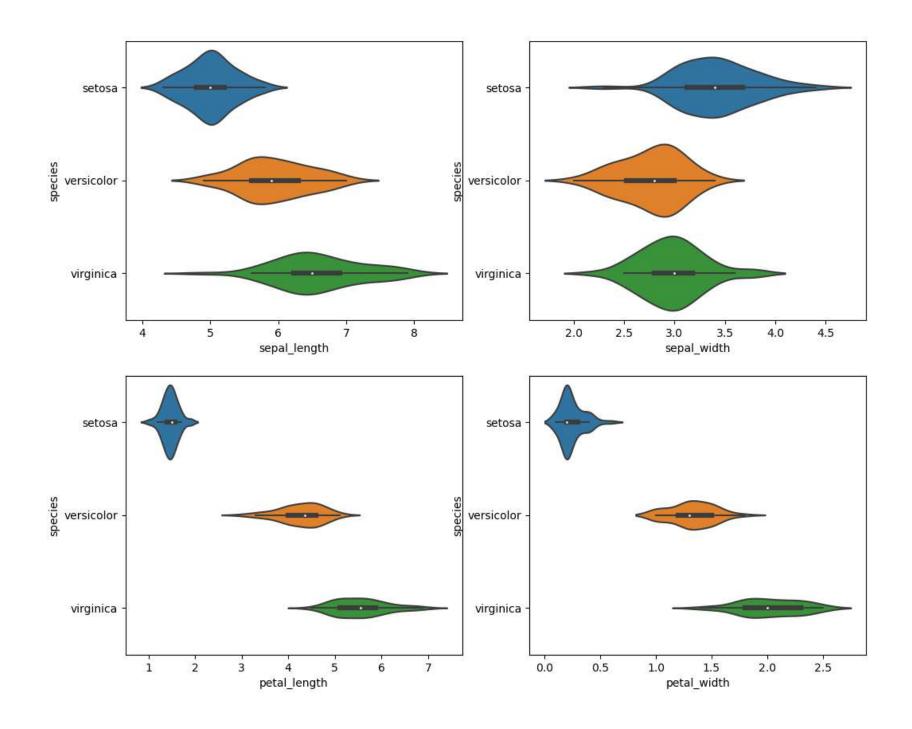
```
1 2 3 4 5 6 7 0.0 0.5 1.0 1.5 2.0 2.5 petal_length petal_width
```

```
In [23]: fig,ax = plt.subplots(1,2,figsize=(15,6))
    ax[0].plot(df['sepal_length'])
    ax[0].plot(df['sepal_width'])
    ax[0].set_title('Sepal length vs width')
    ax[0].legend(['sepal_length','sepal_width'])
    ax[1].plot(df['petal_length'])
    ax[1].plot(df['petal_width'])
    ax[1].set_title('Petal length vs width')
    ax[1].legend(['petal_length','petal_width'])
    plt.show()
```





Boxplots of features



Model Building

```
In [25]: from sklearn.model_selection import train_test_split
         from sklearn.preprocessing import StandardScaler,LabelEncoder
         from sklearn.metrics import accuracy score
         from sklearn.linear model import LogisticRegression
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.tree import DecisionTreeClassifier
In [28]: |df['species'] = LabelEncoder().fit_transform(df['species'])
         X = df.drop(['species'],axis=1)
         y = df['species']
         X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=42)
         lg = LogisticRegression()
         dt = DecisionTreeClassifier()
         knn = KNeighborsClassifier()
In [29]: models = [lg,dt,knn]
         for model in models:
             model.fit(X train,y train)
             y pred = model.predict(X test)
             print(f'Accuracy score of {model} is {accuracy score(y test,y pred)}')
         Accuracy score of LogisticRegression() is 1.0
         Accuracy score of DecisionTreeClassifier() is 1.0
         Accuracy score of KNeighborsClassifier() is 1.0
 In [ ]:
```