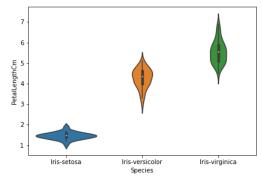
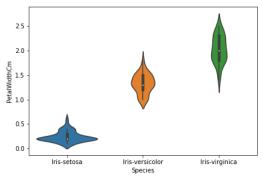
```
In [1]: import numpy as np # linear algebra
           import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
           import matplotlib.pyplot as plt
           import seaborn as sns
 In [3]: iris=pd.read_csv('exp6.csv')
In [27]: iris.head()
Out[27]:
              ld SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
                                                                            Species
                                                                      0.2 Iris-setosa
            0
                            5.1
                                          3.5
            1 2
                            4.9
                                          3.0
                                                         1.4
                                                                      0.2 Iris-setosa
              3
                            4.7
                                          3.2
                                                         1.3
                                                                      0.2 Iris-setosa
            3 4
                                                         1.5
                                                                      0.2 Iris-setosa
                            4.6
                                          3.1
              5
                            5.0
                                          3.6
                                                         1.4
                                                                      0.2 Iris-setosa
In [28]: iris['Species'].unique()
Out[28]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
g=sns.relplot(x='SepalLengthCm',y='SepalWidthCm',data=iris,hue='Species',style='Species')
g.fig.set_size_inches(10,5)
plt.show()
   4.5
   4.0
SepalWidthCm
3.0
                                                                                           Species
                                                                                           Iris-setosa
                                                                                           Iris-versicolor
                                                                                           Iris-virginica
   2.5
   2.0
                                                                 7.5
                                                        7.0
                                                                          8.0
            4.5
                     5.0
                              5.5
                                       6.0
                                               6.5
                                    SepalLengthCm
```

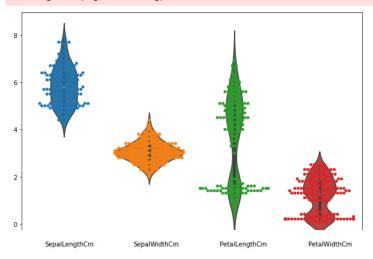
```
In [32]: plt.figure(figsize=(15,10))
  plt.subplot(2,2,1)
  sns.violinplot(x='Species',y='PetalLengthCm',data=iris)
  plt.subplot(2,2,2)
  sns.violinplot(x='Species',y='PetalWidthCm',data=iris)
  plt.subplot(2,2,3)
  sns.violinplot(x='Species',y='SepalLengthCm',data=iris)
  plt.subplot(2,2,4)
  sns.violinplot(x='Species',y='SepalWidthCm',data=iris)
  plt.show()
```



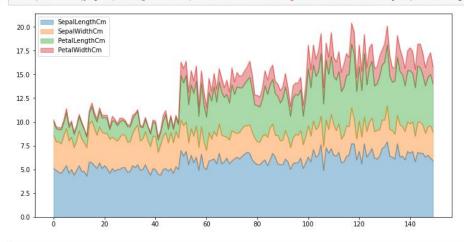


In [33]: plt.subplots(figsize=(10,7)) sns.violinplot(data=iris) sns.swarmplot(data=iris) plt.show()

C:\Users\Vijay\anaconda3\lib\site-packages\seaborn\categorical.py:1296: UserWarning: 9.3% of the points cannot be place
ay want to decrease the size of the markers or use stripplot.
 warnings.warn(msg, UserWarning)



In [34]: iris.plot.area(y=['SepalLengthCm','SepalWidthCm','PetalLengthCm','PetalWidthCm'],alpha=0.4,figsize=(12, 6));



```
In [36]: plt.subplots(figsize = (8,8))
        sns.heatmap(iris.corr(),annot=True,fmt="f").set_title("Corelation of attributes (petal length,width and sepal length,width) among plt.show()
         Corelation of attributes (petal length, width and sepal length, width) among Iris species
                    1.000000
                                                  0.817954
                                                               - 0.8
                                                               - 0.6
                              1.000000
                                                               - 0.4
                                                               - 0.2
                    0.871754
                                        1.000000
                                                  0.962757
                                                               - 0.0
In [44]: gaussian = GaussianNB()
           gaussian.fit(X_train, y_train)
           Y_pred = gaussian.predict(X_test)
           accuracy_nb=round(accuracy_score(y_test,Y_pred)* 100, 2)
           acc_gaussian = round(gaussian.score(X_train, y_train) * 100, 2)
           cm = confusion_matrix(y_test, Y_pred)
           accuracy = accuracy_score(y_test,Y_pred)
           precision =precision score(y test, Y pred,average='micro')
           recall = recall_score(y_test, Y_pred,average='micro')
           f1 = f1_score(y_test,Y_pred,average='micro')
           print('Confusion matrix for Naive Bayes\n',cm)
           print('accuracy_Naive Bayes: %.3f' %accuracy)
           print('precision_Naive Bayes: %.3f' %precision)
           print('recall_Naive Bayes: %.3f' %recall)
print('f1-score_Naive Bayes : %.3f' %f1)
           Confusion matrix for Naive Bayes
            [[16 0 0]
            [0 18 0]
```

[0 0 11]]

accuracy_Naive Bayes: 1.000 precision_Naive Bayes: 1.000 recall_Naive Bayes: 1.000 f1-score_Naive Bayes : 1.000