```
In [1]: #without using library
             import pandas as pd
             import numpy as np
             from sklearn.model_selection import train_test_split
             from sklearn.metrics import accuracy_score,classification_report
             data = pd.read_csv(r"C:\Program Files\Python311\Scripts\iris.csv",header='infer').values
             x = data[:,1:-1]
             y = data[:,-1]
             x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,shuffle=True)
             print("Shapes are: ",x_train.shape,x_test.shape,y_train.shape,y_test.shape)
             nclasses = np.unique(y_train).shape[0]
             print(nclasses)
             dist = np.zeros(shape = x_train.shape[0])
             pred = np.zeros(shape = x test.shape[0])
             classvotes = np.zeros(shape = nclasses)
             k = int(input("Enter the number of nearest neighbours to be used, i.e. k: "))
             for i in range(x_test.shape[0]):
                    dist = np.sqrt(np.sum((x_train-x_test[i])**2,axis = 1))
                   kminind = np.argpartition(dist,k)[0:k]
                   invdist = 1/(dist+10e-20)
                   denom = sum(invdist[kminind])
                   for j in range(k):
                          classvotes[int(y_train[kminind[j]])] += invdist[kminind[j]]
                   classvotes /= denom
                   pred[i] = np.argmax(classvotes)
             print(pred)
             def calc_acc(y_pred,y_true):
                   return np.sum((y_pred).astype(int) == (y_true).astype(int))/y_pred.shape[0]
             def calc_prec(y_pred,y_true):
                   classes = np.unique(y_true)
                   nclasses = classes.shape[0]
                   nrows = y_true.shape[0]
                   classprop = np.zeros(shape=nclasses)
                   for i in range(nclasses):
                          classprop[i] = np.sum(y_true == classes[i])/nrows
                   precclasswise = np.zeros(shape=nclasses)
                   prec = 0
                   for i in range(nclasses):
                          predindices = np.where((((y_pred).astype(int))) == (classes[i].astype(int))) == True)
                          trueindices = np.where((((y_true).astype(int))) == (classes[i].astype(int))) == True)
                          precclasswise[i] = ((len(predindices[0]))-(len(set(predindices[0])-set(trueindices[0]))))/(len(predindices[0]))
                          prec += precclasswise[i]*classprop[i]
                   return prec
             def calc_recall(y_pred,y_true):
                   classes = np.unique(y_true)
                   nclasses = classes.shape[0]
                   nrows = y_true.shape[0]
                   classprop = np.zeros(shape=nclasses)
                   for i in range(nclasses):
                          classprop[i] = np.sum(y_true == classes[i])/nrows
                   recallclasswise=np.zeros(shape==nclasses)
                   recall=0
                   for i in range(nclasses):
                          predindices = np.where((((y_pred).astype(int))) == (classes[i].astype(int))) == True)
                          trueindices = np.where((((y_true).astype(int)) == (classes[i].astype(int))) == True)
                          recall classwise [i] = ((len(predindices[0])) - (len(set(trueindices[0]) - set(predindices[0])))) / (len(predindices[0])) + 
                          recall += recallclasswise[i]*classprop[i]
                   return recall
             accuracy=calc_acc(pred,y_test)
             prec=calc_prec(pred,y_test)
             recall=calc_prec(pred,y_test)
             f1score=(2*prec*recall)/(prec+recall)
             print("Accuracy: ",accuracy)
             print("Precision: ",prec)
             print("Recall: ",recall)
             print("F1-score: ",f1score)
             print("Classification Report :")
             print(classification_report(y_test,pred))
             Shapes are: (120, 4) (30, 4) (120,) (30,)
             Enter the number of nearest neighbours to be used, i.e. k: 3
             [2. 1. 1. 2. 0. 0. 0. 2. 0. 1. 0. 2. 2. 2. 0. 2. 1. 1. 1. 0. 2. 2. 1. 1.
              1. 0. 1. 0. 1. 2.]
             Accuracy: 0.966666666666667
             Precision: 0.96969696969697
             Recall: 0.9696969696969697
             F1-score: 0.9696969696969697
             Classification Report :
                                   precision
                                                        recall f1-score support
                                                                                                9
                            0.0
                                           1.00
                                                           1.00
                                                                           1.00
                           1.0
                                           0.91
                                                           1.00
                                                                           0.95
                                                                                               10
                            2.0
                                           1.00
                                                           0.91
                                                                           0.95
                                                                                              11
                                                                           0.97
                                                                                               30
                    accuracy
                  macro avg
                                           0.97
                                                           0.97
                                                                           0.97
                                                                                               30
```

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0.97

weighted avg

0.97

0.97

30

```
In [2]: #with Library
        #using sklearn
        import pandas as pd
        import numpy as np
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import accuracy_score,classification_report
        from sklearn.neighbors import KNeighborsClassifier
        data=pd.read_csv(r"C:\Program Files\Python311\Scripts\iris.csv",header='infer').values
        x=data[:,1:-1]
        y=data[:,-1]
        x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,shuffle=True)
        k=int(input("Enter number of nearest neighbors: "))
        model=KNeighborsClassifier(n_neighbors=k,weights='distance')
        model.fit(x_train,y_train)
        pred=model.predict(x_test)
        accuracy=accuracy_score(y_test,pred)
        print('Accuracy: ',accuracy)
        print(classification_report(y_test,pred))
        Enter number of nearest neighbors: 3
        Accuracy: 0.93333333333333333
                      precision
                                  recall f1-score support
                 0.0
                          1.00
                                    1.00
                                              1.00
                                                          11
                1.0
                          1.00
                                    0.80
                                              0.89
                                                          10
                 2.0
                          0.82
                                    1.00
                                              0.90
                                                           9
                                              0.93
                                                          30
            accuracy
                                    0.93
                          0.94
                                              0.93
                                                          30
           macro avg
        weighted avg
                          0.95
                                    0.93
                                              0.93
                                                          30
In [ ]:
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In [ ]:
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