```
In [3]: from sklearn.datasets import make_classification
    from sklearn.model_selection import train_test_split
    from sklearn.preprocessing import StandardScaler
    import numpy
    from tqdm import tqdm
    import numpy as np
    from sklearn.metrics.pairwise import euclidean_distances

        x,y = make_classification(n_samples=10000, n_features=2, n_informative=2, n_redundant= 0,n_clusters_per_class=1, rando
        m_state=60)
        X_train, X_test, y_train, y_test = train_test_split(x,y,stratify=y,random_state=42)

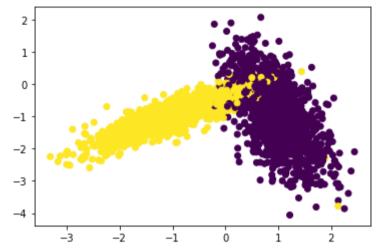
In [4]: length_x_train=len(X_train)
    length_x_train

Out[4]: 7500

In [5]: length_y_train=len(y_train)
    length_y_train=len(y_train)
    length_y_train=len(y_train)
```

Out[5]: 7500

```
In [6]: %matplotlib inline
    import matplotlib.pyplot as plt
    colors={0:'red',1:'blue'}
    plt.scatter(X_test[:,0], X_test[:,1],c=y_test)
    plt.show()
```



## **Implementing Custom RandomSearchCV**

```
In [7]: from sklearn.metrics import accuracy score #importing for accuracy
        import numpy as np #we importing numpy for scientific calculation purpose
        # we writing this function for breaking the train data into cross validation and next train data.
        def train test data splitter(no of datapoints, folds):
            list of test data = []
            spliting form of train data = []
            for i in range(folds+1):
                list of test data.append(i*(no of datapoints/folds))
            for j in range(folds):
                spliting form of train data.append((int(list of test data[j]),int(list of test data[j+1])))
            return spliting form of train data
        def RandomSearchCV(x train,y train,classifier, params, folds):
            no of datapoints = len(x train)
            index provider = train test data splitter(no of datapoints, folds)
            trainscores = []
            testscores = []
            for k in tqdm(params):
                trainscores folds=[]
                                              #initialize two list for store accuracy value
                testscores folds =[]
                for j in range(folds):
                    test indices=index provider[j] #provide the index of tuple which store in list
                    test indices=list(test indices) #we converting in list for finding the range
                    X test = x train[test indices[0]:test indices[1]]
                    Y test = y train[test indices[0]:test indices[1]]
                    # here we finding X train and Y train part
                    X train=x train[list(set(list(range(0,len(x train))))-set(list(range(test indices[0],test indices[1]))))]
                    Y train=y train[list(set(list(range(0,len(y train))))-set(list(range(test indices[0],test indices[1]))))]
```

```
classifier.n_neighbors = k

#making the model
    classifier.fit(X_train,Y_train)

# find testscore predicted value
    Y_predicted = classifier.predict(X_test)

# # find prediction of x_train data
    testscores_folds.append(accuracy_score(Y_test, Y_predicted))

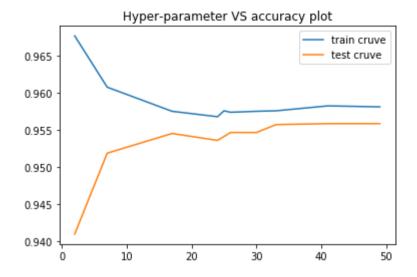
# find prediction of x_train data
    Y_predicted = classifier.predict(X_train)

# find accuracy and append the value
    trainscores_folds.append(accuracy_score(Y_train,Y_predicted))

# here we first finding the mean and then append the value in trainscore and testscore
    trainscores.append(np.mean(np.array(trainscores_folds)))

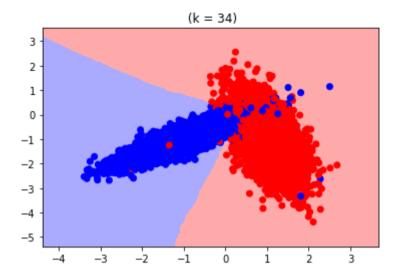
return trainscores,testscores # function return values
```

```
In [11]: from sklearn.metrics import accuracy score #for accuracy
         from sklearn.neighbors import KNeighborsClassifier #importing classifier
         import matplotlib.pyplot as plt #for ploting the graph
         import random # for randomly select the numbers
         import warnings
         warnings.filterwarnings('ignore') #ignore the warnings messages
         neigh=KNeighborsClassifier()
         #generate 10 value randomly between 1 to 50
         params=random.sample(range(1,50), 10)
         # sorting of generated value
         params.sort()
         #initialize the number of folds
         folds=3
         # calling the function
         trainscores, testscores=RandomSearchCV(X train, y train, neigh, params, folds)
         plt.plot(params,trainscores, label='train cruve')
         plt.plot(params, testscores, label='test cruve')
         plt.title('Hyper-parameter VS accuracy plot')
         plt.legend()
         plt.show()
```



```
In [14]: | ####sir this code take from your reference book####
         def plot decision boundary(X1, X2, y, clf):
                 # this code for coloring the graph
              cmap light = ListedColormap(['#FFAAAA', '#AAFFAA', '#AAAAFF'])
             cmap bold = ListedColormap(['#FF0000', '#00FF00', '#0000FF'])
             x \min, x \max = X1.\min() - 1, X1.\max() + 1
             y \min, y \max = X2.min() - 1, X2.max() + 1
             xx, yy = np.meshgrid(np.arange(x min, x max, 0.02), np.arange(y min, y max, 0.02))
             Z = clf.predict(np.c [xx.ravel(), yy.ravel()])
             Z = Z.reshape(xx.shape)
             plt.figure()
             plt.pcolormesh(xx, yy, Z, cmap=cmap light)
             plt.scatter(X1, X2, c=y, cmap=cmap bold)
             plt.xlim(xx.min(), xx.max())
             plt.ylim(yy.min(), yy.max())
             plt.title(" (k = %i)" % (clf.n neighbors))
             plt.show()
```

```
In [15]: from matplotlib.colors import ListedColormap
    neigh = KNeighborsClassifier(n_neighbors = 34)
    neigh.fit(X_train, y_train)
    plot_decision_boundary(X_train[:, 0], X_train[:, 1], y_train, neigh)
```



```
In [ ]:
```