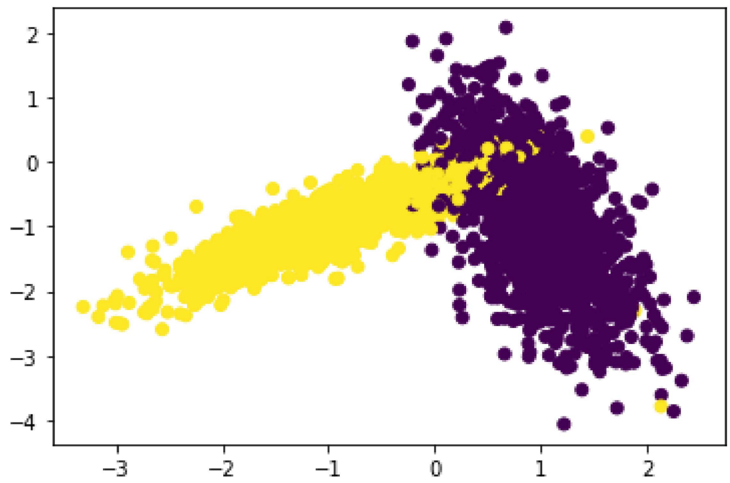
```
from sklearn.datasets import make_classification
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
import numpy
from tqdm import tqdm
import numpy as np
from sklearn.metrics.pairwise import euclidean_distances
```

```
x,y = make_classification(n_samples=10000, n_features=2, n_informative=2, n_redundant= 0,
X_train, X_test, y_train, y_test = train_test_split(x,y,stratify=y,random_state=42)

# del X_train,X_test
```

```
%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

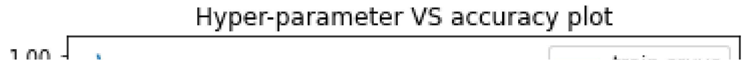
```
def RandomSearchCV(x_train,y_train,classifier, param_range, folds):
    # x_train: its numpy array of shape, (n,d)
    # y_train: its numpy array of shape, (n,) or (n,1)
    # classifier: its typically KNeighborsClassifier()
    # param_range: its a tuple like (a,b) a < b
    # folds: an integer, represents number of folds we need to devide the data and test


    #1.generate 10 unique values(uniform random distribution) in the given range "param_
    # ex: if param_range = (1, 50), we need to generate 10 random numbers in range 1 to
    #2.devide numbers ranging from  0 to len(X_train) into groups= folds
```

```
    # ex: folds=3, and len(x_train)=100, we can devide numbers from 0 to 100 into 3 grou
      group 1: 0-33, group 2:34-66, group 3: 67-100
    #3.for each hyperparameter that we generated in step 1:
        # and using the above groups we have created in step 2 you will do cross-validat

        # first we will keep group 1+group 2 i.e. 0-66 as train data and group 3: 67-100
          test accuracies

        # second we will keep group 1+group 3 i.e. 0-33, 67-100 as train data and group
          train and test accuracies

        # third we will keep group 2+group 3 i.e. 34-100 as train data and group 1: 0-33
          test accuracies
        # based on the 'folds' value we will do the same procedure

        # find the mean of train accuracies of above 3 steps and store in a list "train_
        # find the mean of test accuracies of above 3 steps and store in a list "test_sc
    #4. return both "train_scores" and "test_scores"

# 5. call function RandomSearchCV(x_train,y_train,classifier, param_range, folds) and st
# 6. plot hyper-parameter vs accuracy plot as shown in reference notebook and choose the
# 7. plot the decision boundaries for the model initialized with the best hyperparameter
```

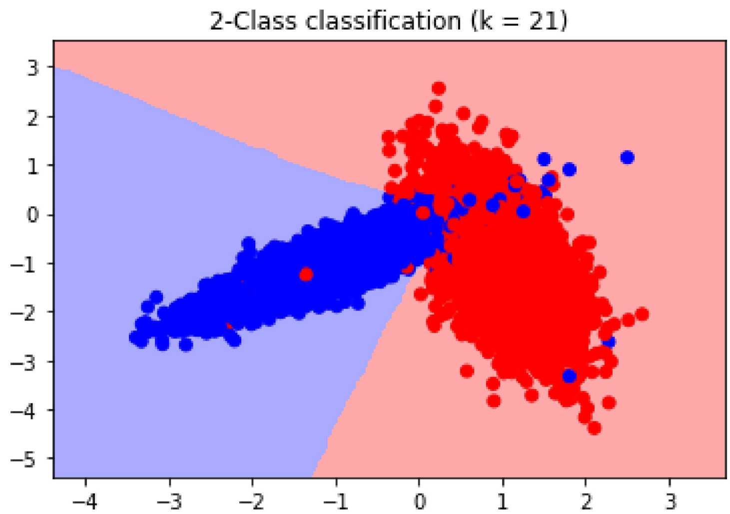
```
from sklearn.metrics import accuracy_score
from sklearn.neighbors import KNeighborsClassifier
import matplotlib.pyplot as plt
import random
import warnings
warnings.filterwarnings("ignore")

def RandomSearchCV(x_train,y_train,classifier, params, folds):
    trainscores = []
    testscores  = []
    splitted_x_train = np.array(np.array_split(x_train,folds))
    splitted_y_train = np.array(np.array_split(y_train,folds))

    for k in tqdm(params['n_neighbors']):
        trainscores_folds = []
        testscores_folds  = []
        for j in range(0, folds):
            # selecting the data points based on the folds
            X_train = np.zeros(splitted_x_train[0].shape)
            Y_train = np.zeros(splitted_y_train[0].shape)
            for count in range(folds):
              if count == j:
                continue
              else:
                X_train = np.append(X_train, splitted_x_train[count],axis=0)
```

```
                Y_train = np.append(Y_train, splitted_y_train[count],axis=0)
            X_train = np.delete(X_train,list(range(0,len(splitted_x_train[0]))),axis=0)
            Y_train = np.delete(Y_train,list(range(0,len(splitted_x_train[0]))),axis=0)
            X_test  = splitted_x_train[j]
            Y_test  = splitted_y_train[j]

            classifier.n_neighbors = k
            classifier.fit(X_train,Y_train)

            Y_predicted = classifier.predict(X_test)
            testscores_folds.append(accuracy_score(Y_test, Y_predicted))

            Y_predicted = classifier.predict(X_train)
            trainscores_folds.append(accuracy_score(Y_train, Y_predicted))
        trainscores.append(np.mean(np.array(trainscores_folds)))
        testscores.append(np.mean(np.array(testscores_folds)))
    return trainscores,testscores


neigh = KNeighborsClassifier()
param_lst = []
while len(param_lst) < 10:
  num = random.randrange(1,25)
  if num%2 == 1:
    if num in param_lst:
      continue
    else:
      param_lst.append(num)

params = {'n_neighbors':sorted(param_lst)}
folds = 3
print(params)

trainscores,testscores = RandomSearchCV(X_train, y_train, neigh, params, folds)

plt.plot(params['n_neighbors'],trainscores, label='train cruve')
plt.plot(params['n_neighbors'],testscores, label='test cruve')
plt.title('Hyper-parameter VS accuracy plot')
plt.legend()
plt.show()
```

```
{'n_neighbors': [1, 3, 5, 7, 9, 11, 13, 17, 19, 21]}
100%|██████████| 10/10 [00:07<00:00,  1.40it/s]
```




```
def plot_decision_boundary(X1, X2, y, clf):
        # Create color maps
    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)
    # Plot also the training points
    plt.scatter(X1, X2, c=y, cmap=cmap_bold)

    plt.xlim(xx.min(), xx.max())
    plt.ylim(yy.min(), yy.max())
    plt.title("2-Class classification (k = %i)" % (clf.n_neighbors))
    plt.show()

from matplotlib.colors import ListedColormap
neigh = KNeighborsClassifier(n_neighbors = 21)
neigh.fit(X_train, y_train)
plot_decision_boundary(X_train[:, 0], X_train[:, 1], y_train, neigh)
```



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