

```

1 from sklearn.datasets import make_classification
2 from sklearn.model_selection import train_test_split
3 from sklearn.preprocessing import StandardScaler
4 import numpy
5 from tqdm import tqdm
6 import numpy as np
7 from sklearn.metrics.pairwise import euclidean_distances
8
9
10 x,y = make_classification(n_samples=10000, n_features=2, n_informative=2, n_redundant= 0,
11 X_train, X_test, y_train, y_test = train_test_split(x,y,stratify=y,random_state=42)
12 print(X_train)

```

```

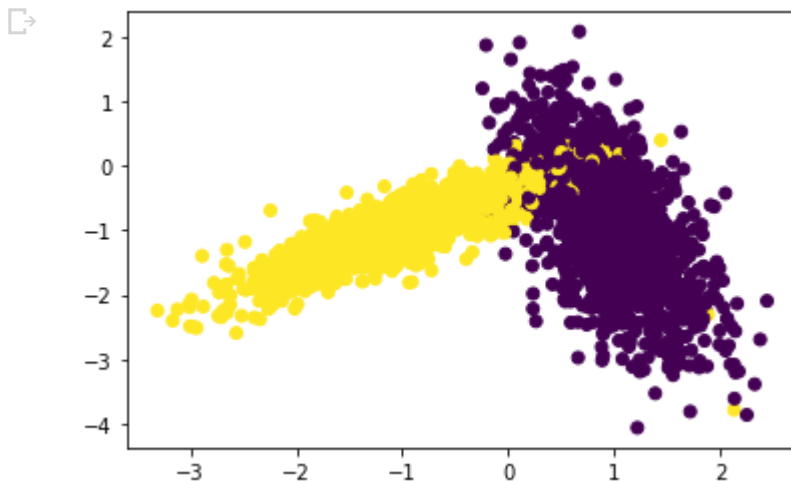
[[ 0.45267141 -1.42381257]
 [ 0.61696406 -0.00418956]
 [-0.60025705 -0.72979921]
 ...
 [ 0.63107723 -0.4743162 ]
 [-2.09387761 -1.76791586]
 [ 1.07909424 -1.67541279]]

```

```

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

```



▼ Implementing Custom RandomSearchCV

```

1 from sklearn.metrics import accuracy_score
2 import random
3 from tqdm import tqdm
4

```

```

5 def random_params_range_1_to_len(params_range):
6     sort_values = random.sample(range(1, params_range),10)
7     sort_values.sort()
8     return sort_values
9
10 def RandomSerachCV(x_train, y_train, classifier, params, folds):
11     trainscores = []
12     testscores = []
13
14     #Randomly selected numbers from params_range
15     params_list= random_params_range_1_to_len(params_range)
16     #printing the random paramter values
17     print(params_list)
18
19     params = {'n_neighbors': params_list}
20
21     for k in tqdm(params['n_neighbors']):
22
23         trainscores_folds = []
24         testscores_folds = []
25
26         for j in range(0, folds): #fold = [1,2,3]
27             #formulae for finding length
28             Values = (len(x_train)/ (folds))
29             #covert into integer values
30             boundary = int(Values)
31
32
33             test_indices=list(set(list(range((boundary*j), (boundary*(j+1))))))
34             train_indices = list(set(list(range(0, len(x_train)))) - set(test_indices))
35             # selecting the data points based on the train_indices and test_indices
36
37             X_train = x_train[train_indices]
38             Y_train = y_train[train_indices]
39             X_test = x_train[test_indices]
40             Y_test = y_train[test_indices]
41
42             classifier.n_neighbors = k
43             classifier.fit(X_train,Y_train)
44
45             Y_predicted = classifier.predict(X_test)
46             testscores_folds.append(accuracy_score(Y_test, Y_predicted))
47
48             Y_predicted = classifier.predict(X_train)
49             trainscores_folds.append(accuracy_score(Y_train, Y_predicted))
50             trainscores.append(np.mean(np.array(trainscores_folds)))
51             testscores.append(np.mean(np.array(testscores_folds)))
52     return trainscores,testscores,params

```

```
1 from sklearn.metrics import accuracy_score
```

```

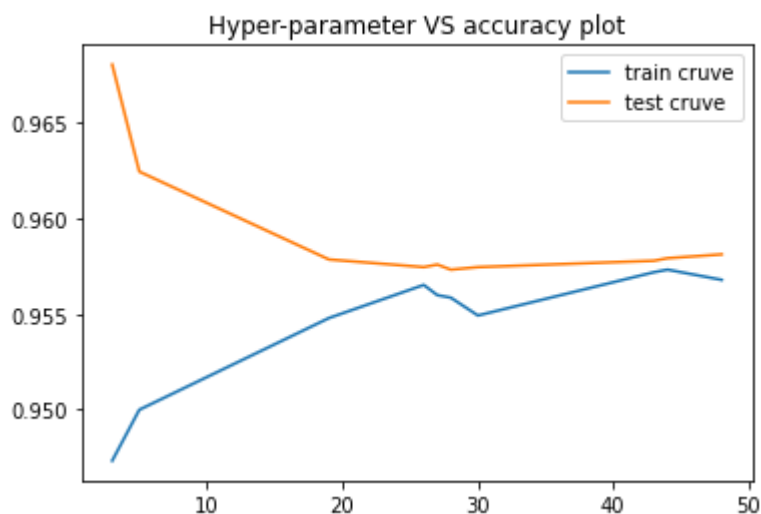
2 from sklearn.neighbors import KNeighborsClassifier
3 import matplotlib.pyplot as plt
4 import random
5 import warnings
6 warnings.filterwarnings("ignore")
7 neigh = KNeighborsClassifier()
8 params_range = 50
9 folds = 3
10 testscores, trainscores, params = RandomSerachCV(X_train, y_train, neigh, params_range, fo
11 print(params)
12 print(trainscores)
13 print(testscores)
14 plt.plot(params['n_neighbors'],trainscores, label='train cruve')
15 plt.plot(params['n_neighbors'],testscores, label='test cruve')
16 plt.title('Hyper-parameter VS accuracy plot')
17 plt.legend()
18 plt.show()

```

```

[3, 5, 19, 26, 27, 28, 30, 43, 44, 48]
100%|██████████| 10/10 [00:07<00:00, 1.29it/s]{'n_neighbors': [3, 5, 19, 26, 27, 28, 30, 43, 44, 48],
trainscores: [0.9473333333333334, 0.9500000000000001, 0.9548, 0.9565333333333333, 0.956, 0.9558666666666667, 0.9680666666666667, 0.9624666666666667, 0.9578666666666668, 0.9574666666666667],
testscores: [0.9473333333333334, 0.9500000000000001, 0.9548, 0.9565333333333333, 0.956, 0.9558666666666667, 0.9680666666666667, 0.9624666666666667, 0.9578666666666668, 0.9574666666666667]}

```



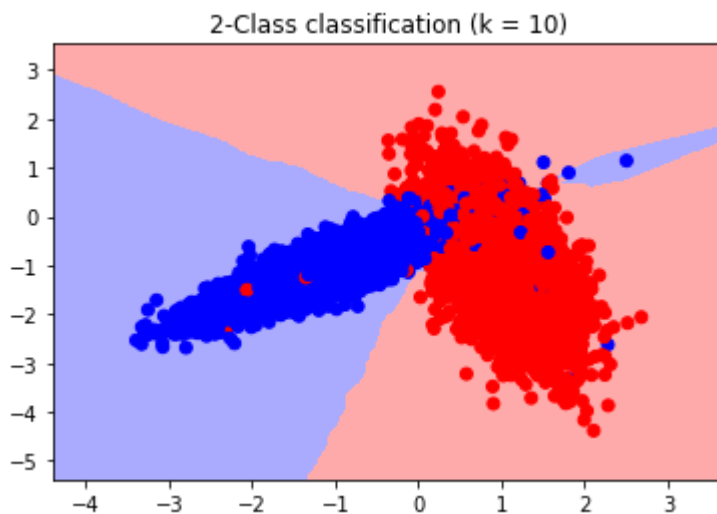
```

1 # taking it from reference
2 def plot_decision_boundary(X1, X2, y, clf):
3     # Create color maps
4     cmap_light = ListedColormap(['#FFAAAA', '#AAFFAA', '#AAAAFF'])
5     cmap_bold = ListedColormap(['#FF0000', '#00FF00', '#0000FF'])
6
7     x_min, x_max = X1.min() - 1, X1.max() + 1
8     y_min, y_max = X2.min() - 1, X2.max() + 1
9
10    xx, yy = np.meshgrid(np.arange(x_min, x_max, 0.02), np.arange(y_min, y_max, 0.02))
11    Z = clf.predict(np.c_[xx.ravel(), yy.ravel()])
12    Z = Z.reshape(xx.shape)

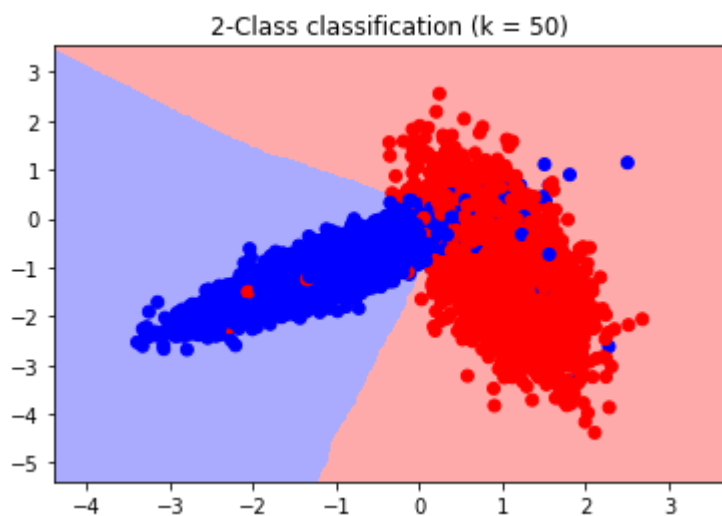
```

```
13
14 plt.figure()
15 plt.pcolormesh(xx, yy, Z, cmap=cmap_light)
16 # Plot also the training points
17 plt.scatter(X1, X2, c=y, cmap=cmap_bold)
18
19 plt.xlim(xx.min(), xx.max())
20 plt.ylim(yy.min(), yy.max())
21 plt.title("2-Class classification (k = %i)" % (clf.n_neighbors))
22 plt.show()
```

```
1 from matplotlib.colors import ListedColormap
2 neigh = KNeighborsClassifier(n_neighbors = 10)
3 neigh.fit(X_train, y_train)
4 plot_decision_boundary(X_train[:, 0], X_train[:, 1], y_train, neigh)
```



```
1 from matplotlib.colors import ListedColormap
2 neigh = KNeighborsClassifier(n_neighbors = 50)
3 neigh.fit(X_train, y_train)
4 plot_decision_boundary(X_train[:, 0], X_train[:, 1], y_train, neigh)
```



✓ 6s completed at 12:32 PM

