

Assignment_4_Instructions

April 8, 2022

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[ ]: 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

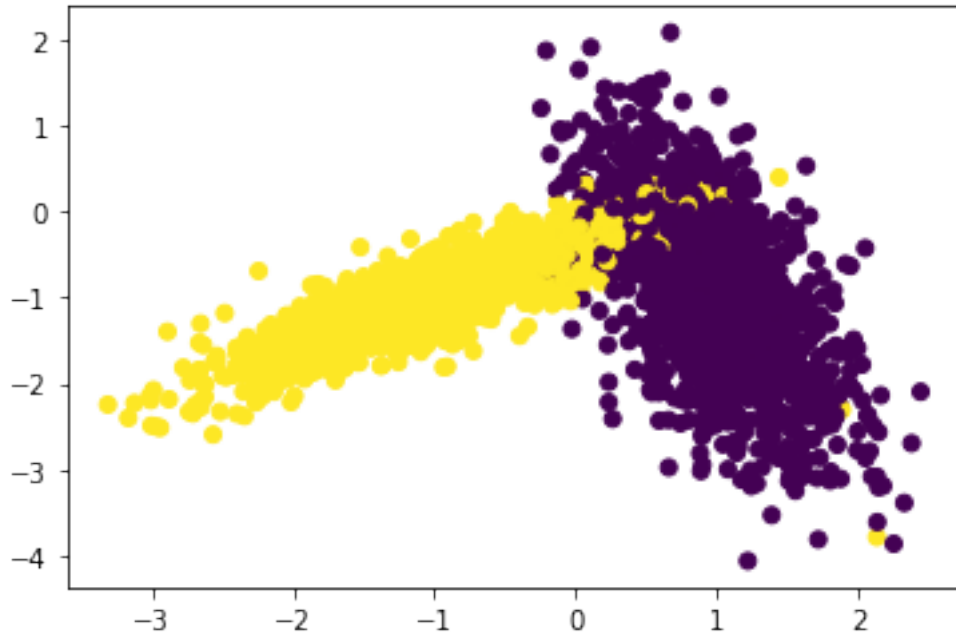
from sklearn.metrics import accuracy_score
from sklearn.neighbors import KNeighborsClassifier
import matplotlib.pyplot as plt
import random as r
import warnings
warnings.filterwarnings("ignore")

neigh = KNeighborsClassifier()

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

# del X_train,X_test
```

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[ ]: %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()
```



```
[ ]: #Ref
#https://moonbooks.org/Articles/
#How-to-create-a-list-of-tuples-with-random-numbers-in-python-/
#https://www.youtube.com/watch?v=8YJRtyx10vw
#https://stackoverflow.com/questions/61818704/
#building-a-custom-randomsearchcv-using-python
#https://machinelearningmastery.com/implement-resampling-methods-scratch-python/

def RandomSearchcv(x_train,y_train,classifier,param_range,folds):

    train_scores = []
    test_scores = []
    #print(x_train.shape, 'Shape train')
    #print(y_train.shape, 'Shape test')

    params = sorted(r.sample(range(1,param_range),10))
    print(params)

    # [1,2,3,4,5,6,7,8,9] ---> [[1, 2], [3, 4], [5, 6], [7, 8]]
    # Dividing the whole data to k group ie 3 grps
    x_train_group = []
    y_train_group = []
```

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step = int(len(X_train)/folds)

for i in range(0,len(X_train),step):
    #print(i)
    x_train_group.append(X_train[i:i + step])
    y_train_group.append(y_train[i:i + step])

#print(len(x_train_group),'thi')

for parameter in tqdm(params):
    # Storing the avg train and test accuracy for each parameter after cv
    train_score_fold = []
    test_score_fold = []

    # [[1, 2], [3, 4], [5, 6]] ---> train :[[1,2],[3,4]] cv [5,6]
    for group_num in range(len(x_train_group)):
        train_set_x = np.concatenate(x_train_group[:group_num] +
    ↪x_train_group[group_num + 1:])
        cv_set_x = x_train_group[group_num]
        train_set_y = np.concatenate(y_train_group[:group_num] +
    ↪y_train_group[group_num + 1:])
        cv_set_y = y_train_group[group_num]

        #Fitting to model
        neigh.n_neighbors = parameter
        neigh.fit(train_set_x,train_set_y)

        train_pred = neigh.predict(train_set_x)
        train_score_fold.append(accuracy_score(train_set_y,train_pred))

        test_pred = neigh.predict(cv_set_x)
        test_score_fold.append(accuracy_score(cv_set_y,test_pred))

    train_scores.append(np.mean(train_score_fold))
    test_scores.append(np.mean(test_score_fold))

    return train_scores,test_scores,params

train,test,params = RandomSearchcv(X_train,y_train,neigh,50,3)
print(train)
print(test)

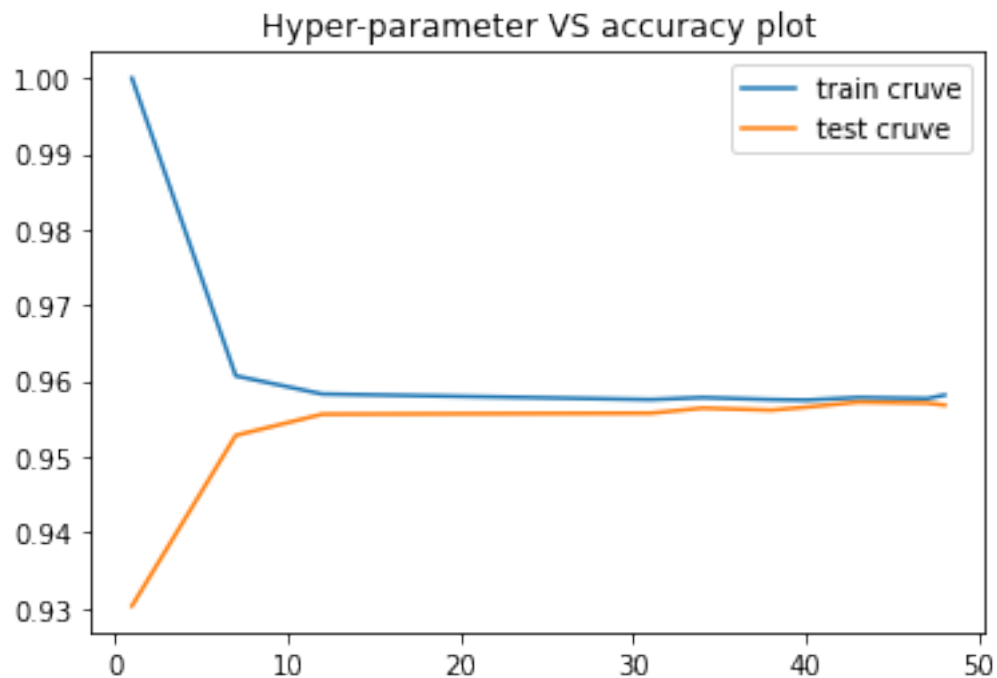
```

[1, 7, 12, 31, 34, 38, 40, 43, 47, 48]

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```
[1.0, 0.9606666666666666, 0.9582666666666667, 0.9575333333333332,
0.9577999999999999, 0.9575333333333335, 0.9574666666666666, 0.9578000000000001,
0.9576666666666668, 0.9581333333333334]
[0.9302666666666667, 0.9528, 0.9556, 0.9557333333333333, 0.9564,
0.9561333333333334, 0.9565333333333333, 0.9571999999999999, 0.9570666666666666,
0.9568]
```

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



```
[ ]: # understanding this code line by line is not that important
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)

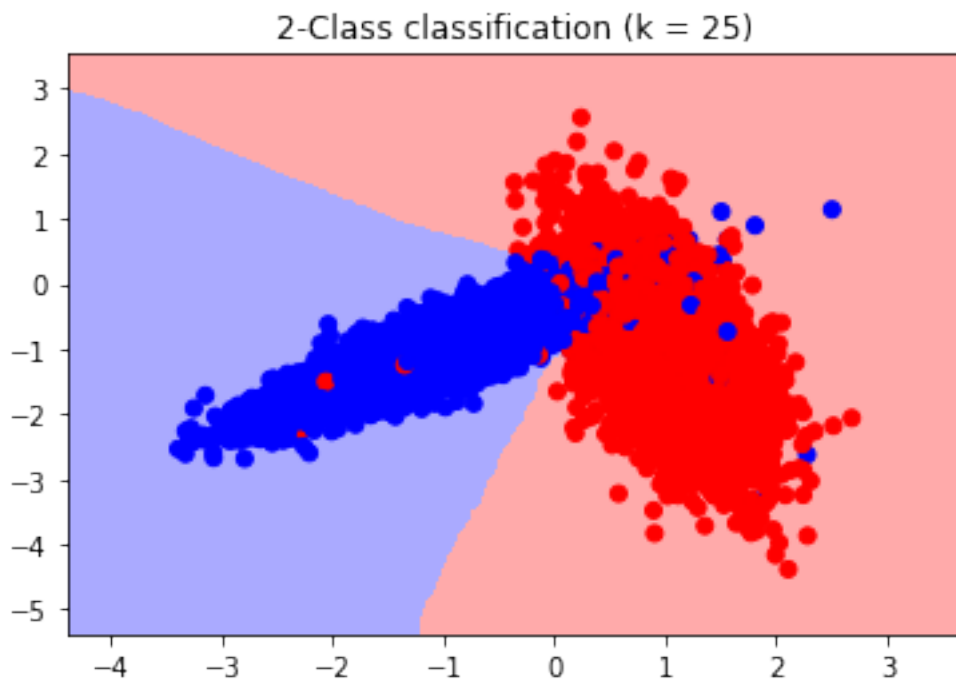
plt.xlim(xx.min(), xx.max())
plt.ylim(yy.min(), yy.max())
plt.title("2-Class classification (k = %i)" % (clf.n_neighbors))
plt.show()

```

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[ ]: from matplotlib.colors import ListedColormap
neigh = KNeighborsClassifier(n_neighbors = 25)
neigh.fit(X_train, y_train)
plot_decision_boundary(X_train[:, 0], X_train[:, 1], y_train, neigh)

```



1 Implementing Custom RandomSearchCV

```
[ ]: !wget -nc https://raw.githubusercontent.com/Bluedevil-hub/colab-to-pdf/master/
      ↪colab_pdf.py
      from colab_pdf import colab_pdf
      colab_pdf('Assignment_4_Instructions.ipynb')
```

```
--2022-04-08 13:21:40-- https://raw.githubusercontent.com/Bluedevil-hub/colab-
to-pdf/master/colab_pdf.py
Resolving raw.githubusercontent.com (raw.githubusercontent.com)...
185.199.108.133, 185.199.109.133, 185.199.110.133, ...
Connecting to raw.githubusercontent.com
(raw.githubusercontent.com)|185.199.108.133|:443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 1864 (1.8K) [text/plain]
Saving to: 'colab_pdf.py'
```

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colab_pdf.py          100%[=====>]    1.82K  --.-KB/s    in 0s
```

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2022-04-08 13:21:41 (24.8 MB/s) - 'colab_pdf.py' saved [1864/1864]
```

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[ ]:
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