

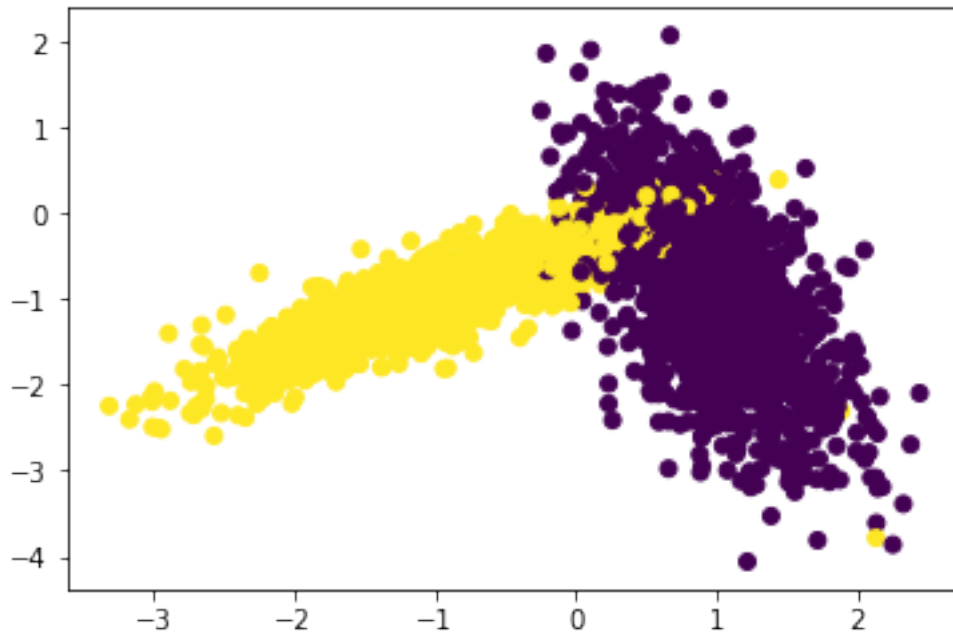
# RandomsearchCV Assignment

April 19, 2020

```
[1]: 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, random_state=60)
X_train, X_test, y_train, y_test =
    ↳train_test_split(x,y,stratify=y,random_state=42)
```

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[2]: %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()
```



## 1 Implementing Custom RandomSearchCV

```
[3]: from sklearn.metrics import accuracy_score

def RandomSearchCV(x_train,y_train,clf,param_range,folds):
    low=param_range[0]
    high=param_range[1]
    params=sorted(random.sample(range(low,high),10)) #randomly generated param_
    ↪ values
    trainscores=[]
    cvscores=[]
    for k in tqdm(params): #for every k value
        trainscores_folds=[]
        cvscores_folds=[]
        n=len(x_train) #length of train dataset is 7500
        part=int(n/folds) #number of parts the train dataset should divide into
        for j in range(0,n,part): #in 1st iterate cv_indices will_
            ↪ be (0 to 2499) and remaining are train_indices
                cv_indices=list(range(j,j+part)) #simillary in 2nd iterate_
            ↪ cv_indiecs(2500 to 4999) and 3rd iterate (5000 to 7499)
                train_indices=list((set(list(range(1,n))))-(set(cv_indices)))

        X_train=x_train[train_indices]
        Y_train=y_train[train_indices]
```

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X_cv=x_train[cv_indices]
Y_cv=y_train[cv_indices]

clf.n_neighbors=k
clf.fit(X_train,Y_train)  #fitting model KNeighborsClassifier

Y_predict=clf.predict(X_train) #predicting class label for train_
↪dataset
trainscores_folds.
↪append(accuracy_score(Y_train,Y_predict))#accuracy for trainscores

Y_predict=clf.predict(X_cv)  #predicting class label for cv dataset
cvscores_folds.append(accuracy_score(Y_cv,Y_predict))  #accuracy_
↪for cvscores

trainscores.append(np.mean(np.array(trainscores_folds)))  #accuracy of_
↪train and cv scores for
cvscores.append(np.mean(np.array(cvscores_folds)))  #different_
↪values of k
return trainscores,cvscores,params

```

```

[4]: from sklearn.metrics import accuracy_score
from sklearn.neighbors import KNeighborsClassifier
import matplotlib.pyplot as plt
import random

neigh = KNeighborsClassifier()
param_range=(1,50)
folds=3

trainscores,cvscores,params=RandomSearchCV(X_train,y_train,neigh,param_range,folds)

plt.plot(params,trainscores,label="Train curve")
plt.plot(params,cvscores,label="CV curve")
plt.scatter(params,trainscores,label="Train Accuracy points")
plt.scatter(params,cvscores,label="CV Accuracy points")

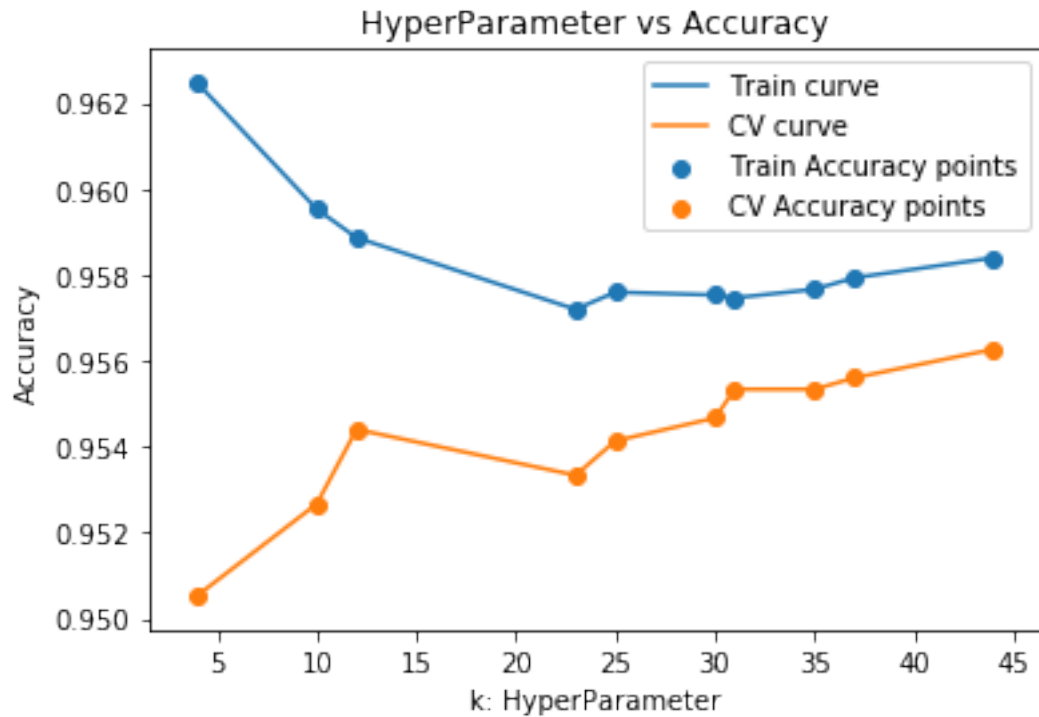
plt.legend()
plt.xlabel("k: HyperParameter")
plt.ylabel("Accuracy")
plt.title("HyperParameter vs Accuracy")
plt.show()

print("k values {}".format(params))

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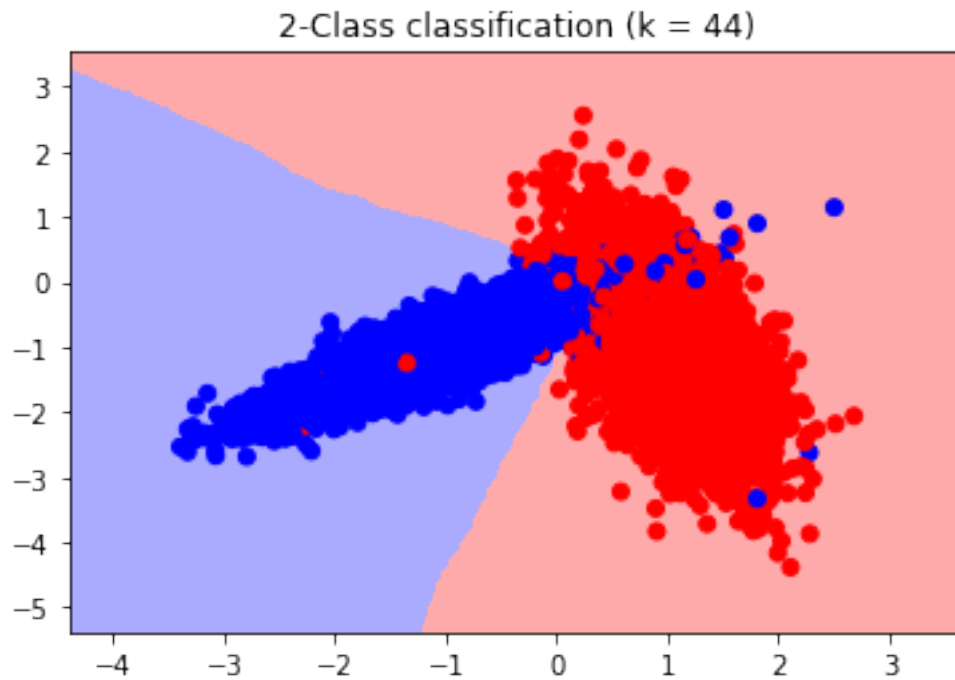


k values [4, 10, 12, 23, 25, 30, 31, 35, 37, 44]

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[5]: 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())
```

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

```
[6]: from matplotlib.colors import ListedColormap
neigh = KNeighborsClassifier(n_neighbors=44)
neigh.fit(X_train,y_train)
plot_decision_boundary(X_train[:, 0],X_train[:, 1],y_train,neigh)
```



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