

WE CAN WRITE OUR OWN RANDOM GRID SEARCH CV. IT IS A CROSS VALIDATION SPLIT TECHNIQUE IN ORDER TO FIND THE GOOD HYPERPARAMETERS.

Double-click (or enter) to edit

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from sklearn.datasets import make_classification
from sklearn.model_selection import train_test_split
import random
import matplotlib.pyplot as plt
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score
import numpy as np
import warnings
warnings.filterwarnings("ignore")

#####Random Search cv definition#####
def RandomSearchCV(x_train,y_train,classifier,k_val,folds):
    X_train=[];Y_train=[];cvscores=[];trainscores=[];testscores=[]
    #####DIVIDE x_train into k fold#####

    #The idea is simple. If we have 5 folds that implies we divide the the data into 5 par

    #x_train[j*int((len(x_train)/folds)):(j+1)*int((len(x_train)/folds))] -->It takes sele
        # over these 5 parts, we assign a variable j and move it part by part. P

    for j in range(0,folds):
        if ((j+1)*int((len(x_train)/folds)))<= int(len(x_train)) and ((j+1)*int((len(y_tra
            X_train.append(x_train[j*int((len(x_train)/folds)):(j+1)*int((len(x
            Y_train.append(y_train[j*int((len(y_train)/folds)):(j+1)*int((len(y

    # So every part is being appended in X_train and Y_train. Seperate steps are given b

    #Now we have data divided into parts. We will then take a k neighbour and apply trai

    #What is cv Data? Simple for 5 parts of your train every 4 parts goes to train and r

    #Start with K values now
    for k in (k_val):          ##### for every k neighbor value#####

        trainscores_folds = [];testscores_folds = [] # these will handle the mean accuracy

        for j in range(0, folds):          #####for every fold/part#####
            #####selecting data points accordingly, also X_train,Y_train,Xtrain,Ytrain:list #
                Xtrain=[];Ytrain=[];Xtest=[];Ytest=[]
                for t in range(0,folds):
                    #the below eqn has RHS which tells the part in the data taken, ex:j=
                    if folds-1-t != j:          ##### when this value becomes equal then da

                                Xtrain.append(X_train[t]);Ytrain.append(Y_train[t])
                else:
                    Xtest.append(X_train[folds-1-j]);Ytest.append(Y_train[fo
```

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##### fitting the k neighbors for the jth part being cv and the remaini
classifier.n_neighbors = k

#Note: Please fit the x data in 2-d array, y data(class label) will be
classifier.fit(np.array(Xtrain).reshape((folds-1)*len(Xtrain[0]),2),np

##### prediction and accuracy scores for test#####
Y_predicted = classifier.predict(np.array(Xtest).reshape(len(Xtest[0])
testscores_folds.append(accuracy_score(np.array(Ytest).reshape(len(Ytr

##### prediction and accuracy scores for train #####
Y_predicted = classifier.predict(np.array(Xtrain).reshape((folds-1)*le
trainscores_folds.append(accuracy_score(np.array(Ytrain).reshape((fold

#Mean scores for every k neighbour value
testscores.append(np.mean(np.array(testscores_folds)))# average CV score for all t
trainscores.append(np.mean(np.array(trainscores_folds)))
return trainscores,testscores

sample_size=int(input('please enter number of samples in the dataset : '))
x,y = make_classification(n_samples=sample_size, n_features=2, n_informative=2,n_redundant
x_train, x_test, y_train, y_test = train_test_split(x,y,stratify=y,random_state=42)

#####GET # OF FOLDS#####
folds=int(input('please enter number of folds you need < 10 :  '))

low_lim=int(input('define k range: please enter the lower limit:  '))
hig_lim=int(input('define k range: please enter the high limit > 10:  '))

#####GET 10 UNIQUE K VALUES#####
k_val=random.sample(range(low_lim,hig_lim),10);k_val=sorted(k_val,reverse=False)

#Considering simple k neighbours algorithm
neigh = KNeighborsClassifier()
trainscores,cvscor=RandomSearchCV(x_train,y_train,neigh,k_val,folds)

##### plotting the accuracy plot#####
plt.plot(k_val,trainscores, label='train cruve')
plt.plot(k_val,cvscor, label='cv cruve')
plt.title('Hyper-parameter VS accuracy plot')
plt.legend()
plt.show()

```

The plot shows training accuracy (blue line) and cross-validation accuracy (orange line) for various hyper-parameters. The x-axis represents different hyper-parameters, and the y-axis represents accuracy. The training accuracy is generally higher than the cross-validation accuracy, and both accuracies decrease as the hyper-parameters move from left to right.

Hyper-parameter	train cruce	cv cruce
1	0.88	0.85
2	0.88	0.87
3	0.87	0.86
4	0.87	0.87
5	0.87	0.88
6	0.87	0.87
7	0.86	0.85
8	0.84	0.80
9	0.71	0.71

0.64

0.85333333333333334