

Assignment_4_Instructions

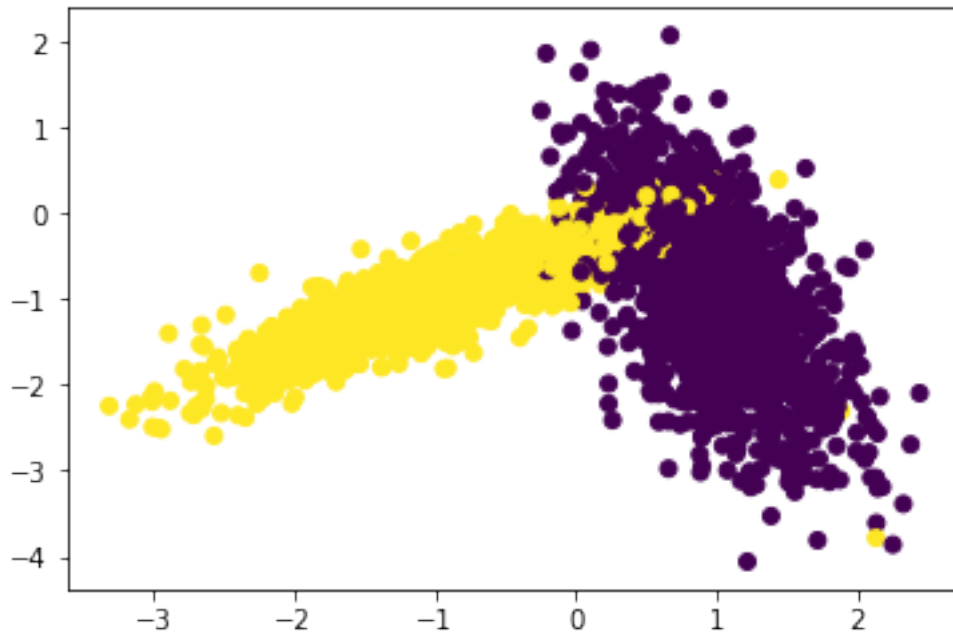
December 31, 2019

```
In [1]: from sklearn.datasets import make_classification
        from sklearn.model_selection import train_test_split
        from sklearn.preprocessing import StandardScaler
        from itertools import combinations
        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=
        X_train, X_test, y_train, y_test = train_test_split(x,y,stratify=y,random_state=42)

        # del X_train,X_test

In [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

In [3]: *# Generate n random value between a and b*

```
def get_n_unique_random_value(a,b,n):
    return random.sample(range(a,b),n)
```

In [4]: *# it will take classifier and set of values for hyper prameter in dict type dict({hyperparameter: value})*
we are implementing this only for KNN, the hyper parameter should n_neighbors
from sklearn.metrics import accuracy_score

```
def RandomSearchCV(x_train,y_train,classifier, neighbors, folds):
    trainscores = []
    testscores = []
    # Storing folds+1 equidistant index for train data
    folds_value = [int(i) for i in np.linspace(0,len(x_train)-1,folds+1)]
    group = []
    # Storing indexes of train data
    indices = [i for i in range(0,folds)]
    # Storing combinations of groups index of train data
    group_idx = list(combinations(indices, folds-1))
    # Storing index of train data for each group on which training needs to be done
    for i in range(0, folds):
        if i==0:
            idx = list(range(folds_value[i],folds_value[i+1]+1))
        else:
```

```

        idx = list(range(folds_value[i]+1,folds_value[i+1]+1))
        group.append(idx)
    for k in tqdm(neighbors):
        trainscores_folds = []
        testscores_folds = []
        for j in range(0, folds):
            train_idx = []
            for i in group_idx[j]:
                [train_idx.append(value) for value in group[i]]
            test_idx = list(set(indices)-set(group_idx[j]))

            X_train = x_train[train_idx]
            Y_train = y_train[train_idx]

            X_test = x_train[group[test_idx[0]]]
            Y_test = y_train[group[test_idx[0]]]

            #             print(X_train.shape)
            #             print(Y_train.shape)
            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

```

```

In [5]: from sklearn.metrics import accuracy_score
        from sklearn.neighbors import KNeighborsClassifier
        import matplotlib.pyplot as plt
        import random
        import warnings
        warnings.filterwarnings("ignore")

        neigh = KNeighborsClassifier()

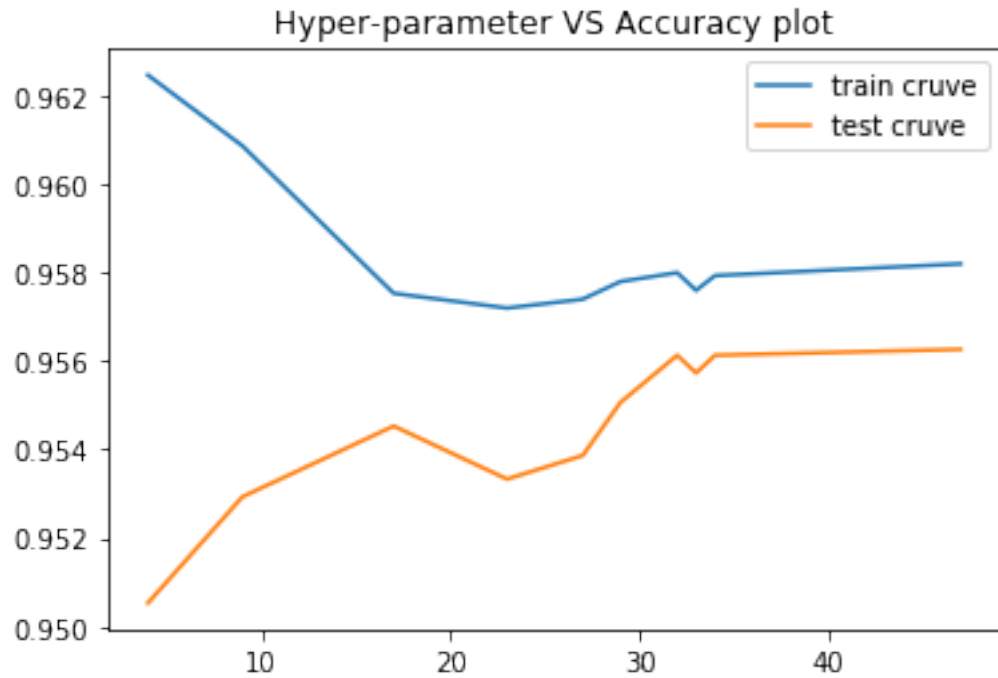
        params = {'n_neighbors':[1,50]}
        neighbors = get_unique_random_value(params['n_neighbors'][0],params['n_neighbors'][1],
        neighbors.sort()
        folds = 3

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

```

```
plt.plot(neighbors, trainscores, label='train cruve')
plt.plot(neighbors, testscores, label='test cruve')
plt.title('Hyper-parameter VS Accuracy plot')
plt.legend()
plt.show()
```

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```
In [6]: # len(trainscores)
# print(trainscores)
# print(testscores)
# neighbors.sort()
# print(neighbors)
```