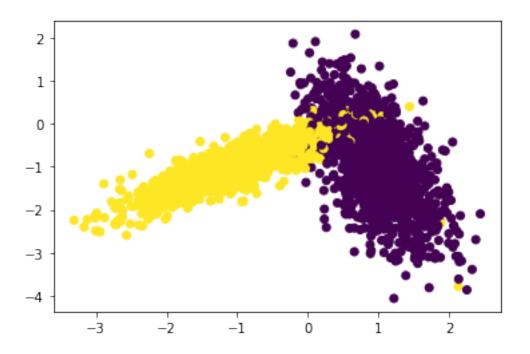
Assignment_4_Instructions

December 19, 2021

```
[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
   from tqdm.notebook 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)
   # del X_train, X_test
    # print(type(X_train), type(y_train))
   # len(X_train)
[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()
```



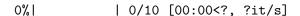
1 Implementing Custom RandomSearchCV

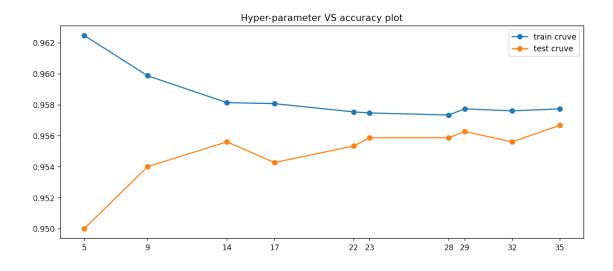
```
[3]: def RandomSearchCV(x_train,y_train,classifier, param_range, folds):
        , , ,
        x_{train}: its numpy array of shape, (n,d)
        y_train: its numpy array of shape, (n,) or (n,1)
        classifier : its typically KNeighborsClassifier()
        param_range: its a tuple like (a,b) a < b
                  : an integer, represents number of folds we need to devide the ...
     \rightarrow data and test our model
        111
        #1. qenerate 10 unique values (uniform random distribution) in the given
     →range "param_range" and store them as "params"
        ten_value_param_range = sorted(np.random.
     →randint(param_range[0],param_range[1], 10))
        params = {'n_neighbors': ten_value_param_range }
        \#2. devide numbers ranging from 0 to len(X_train) into groups k folds
        111
        https://stackoverflow.com/a/43106405
```

```
import numpy as np
   xy = np.array_split(range(100), 3)
   for i in xy:
        print(i)
    [ \ 0 \ \ 1 \ \ 2 \ \ 3 \ \ 4 \ \ 5 \ \ 6 \ \ 7 \ \ 8 \ \ 9 \ 10 \ 11 \ 12 \ 13 \ 14 \ 15 \ 16 \ 17 \ 18 \ 19 \ 20 \ 21 \ 22 \ 23 \ 24 {\sqcup} 
→25 26 27 28 29 30 31 32 33]
    [ 34\ 35\ 36\ 37\ 38\ 39\ 40\ 41\ 42\ 43\ 44\ 45\ 46\ 47\ 48\ 49\ 50\ 51\ 52\ 53\ 54\ 55\ 56\ 57\ 58_{\sqcup} 
→59 60 61 62 63 64 65 66]
    \begin{smallmatrix} 67 & 68 & 69 & 70 & 71 & 72 & 73 & 74 & 75 & 76 & 77 & 78 & 79 & 80 & 81 & 82 & 83 & 84 & 85 & 86 & 87 & 88 & 89 & 90 & 91 \\ \end{smallmatrix} 
→92 93 94 95 96 97 98 99]
   111
   x_train_split = np.array_split(x_train, folds)
   y_train_split = np.array_split(y_train, folds)
   x train folds = []
   y_train_folds = []
   for xt,yt in zip(x_train_split,y_train_split):
        x_train_folds.append(xt)
        y_train_folds.append(yt)
    print(len(x_train_split)) # ==> folds
   # 3. for each hyperparameter that we generate values
   train scores = []
   test scores = []
   for k in tqdm(params['n_neighbors']):
        train_scores_folds = []
        test_scores_folds = []
        111
        https://stackoverflow.com/a/61819104
        x_train_fold_input = [[5], [7], [9]]
        for xy in range(len(x_train_fold_input)):
            print('CV data : ', x_train_fold_input[xy])
            print('Train data : ', x_train_fold_input[0:xy] +__
\rightarrow x_train_fold_input[xy+1:])
            print('\n')
        CV data : [5]
        Train data : [[7], [9]]
        CV data : [7]
        Train data : [[5], [9]]
```

```
CV data : [9]
        Train data : [[5], [7]]
       for index in range(len(x_train_folds)):
         K-Fold splitting
           x_fold_train = []
           y_fold_train = []
            x_train_folds_cv = x_train_folds[index]
           x_train_folds_train = x_train_folds[0:index] +__
→x train folds[index+1:]
            y_train_folds_cv = y_train_folds[index]
            y_train_folds_train = y_train_folds[0:index] +__
→y_train_folds[index+1:]
             print(len(x_train_folds_train)) # ==> folds-1, means [[...], [...
\rightarrow]] : if fold ==3
                                                      But we need [.....]
            if len(x_train_folds_train) == folds-1 :
                for each in x_train_folds_train:
                    x_fold_train.extend(each)
            if len(y_train_folds_train) == folds-1:
                for each in y_train_folds_train:
                    y_fold_train.extend(each)
#
             print(len(x_fold_train)) # ==> [.....]
         Applying Classifier: Assignment_4_Reference
            classifier.n_neighbors = k
            classifier.fit(x_fold_train,y_fold_train)
          Train score Computation
            Y_predicted = classifier.predict(x_fold_train)
            train_scores_folds.append(accuracy_score(y_fold_train, Y_predicted))
          Test score Computation
            Y_predicted = classifier.predict(x_train_folds_cv)
            test_scores_folds.append(accuracy_score(y_train_folds_cv,_
→Y predicted))
       train_scores.append(np.mean(np.array(train_scores_folds)))
```

```
test_scores.append(np.mean(np.array(test_scores_folds)))
        return train_scores, test_scores, params
[4]: | #https://colab.research.google.com/drive/13NoPSgtq_OEqFiUbAmzjVsbMwWUayofG :__
     \rightarrow Assignment_4_Reference
    from sklearn.metrics import accuracy_score
    from sklearn.neighbors import KNeighborsClassifier
    import matplotlib.pyplot as plt
    import random
    import warnings
    warnings.filterwarnings("ignore")
    classifier = KNeighborsClassifier()
    param_range = (1, 50)
    folds = 3
    train_scores,cv_scores, params = RandomSearchCV(X_train,y_train,classifier,_
     →param_range, folds)
    plt.figure(figsize = (12,5), dpi =125)
    plt.plot(params['n_neighbors'],train_scores, 'o-', label='train cruve')
    plt.plot(params['n_neighbors'],cv_scores, 'o-', label='test cruve')
    plt.title('Hyper-parameter VS accuracy plot')
    plt.xticks(params['n_neighbors'])
    plt.legend()
    plt.show()
```





```
[5]: #https://colab.research.google.com/drive/13NoPSqtq_OEqFiUbAmzjVsbMWWUayofG:
    \rightarrow Assignment_4_Reference
   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,_
     \rightarrow 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()
[]: #https://colab.research.google.com/drive/13NoPSgtq_OEqFiUbAmzjVsbMWWUayofG:
    →Assignment_4_Reference
   from matplotlib.colors import ListedColormap
   neigh = KNeighborsClassifier(n_neighbors = 35)
   neigh.fit(X_train, y_train)
   plot_decision_boundary(X_train[:, 0], X_train[:, 1], y_train, neigh)
[7]: | !jupyter nbconvert --to PDF "Assignment_4_Instructions.ipynb"
```