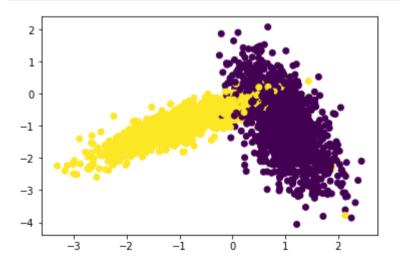
```
In [26]: from sklearn.datasets import make_classification
    from sklearn.model_selection import train_test_split
    from sklearn.preprocessing import StandardScaler
    import numpy
    from tqdm import tqdm
    import numpy as np
    from sklearn.metrics.pairwise import euclidean_distances

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

## In [27]: %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()



In [28]: import random

```
In [29]: def random_dist(param_range):
    unique_values=random.sample(range(1, param_range), 10)
    unique_values.sort()
    return unique_values

In [31]: def partition(xtrain,ytrain,folds):
    division = len(xtrain) / float(folds)
        xtrain=xtrain.tolist()
    ytrain=ytrain.tolist()
    group=([ xtrain[int(round(division * i)): int(round(division * (i + 1)))] for i in range(folds) ])
    label=([ ytrain[int(round(division * i)): int(round(division * (i + 1)))] for i in range(folds) ])
    return group,label
```

```
In [32]: def RandomSerachCV(x train, y train, classifier, param range, folds):
             train_scores = []
             test scores = []
              param list=random dist(param range)
             print(param list)
              params={'n neighbors': param list}
             group,label=partition(x train,y train,folds)
             for k in tqdm(params['n neighbors']):
                 trainscores folds = []
                 testscores folds = []
                  for i in reversed(range(folds)):
                     X train = [group[iter] for iter in range(3) if iter != i]
                     X train = [item for sublist in X train for item in sublist]
                     Y train = [label[iter] for iter in range(folds) if iter != i]
                     Y train = [j for sublist in Y train for j in sublist]
                     X test = group[i]
                     Y test = label[i]
                      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))
                 train scores.append(np.mean(np.array(trainscores folds)))
                  test scores.append(np.mean(np.array(testscores folds)))
              return train_scores, test_scores, params
```

In [41]: 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() param range=50 folds = 3train scores,cv scores,params= RandomSerachCV(X train,y train,neigh,param range,folds) print(params) print(train scores) print(cv scores) plt.plot(params['n neighbors'],train scores, label='train cruve') plt.plot(params['n neighbors'],cv scores, label='test cruve') plt.title('Hyper-parameter VS accuracy plot') plt.legend() plt.show()

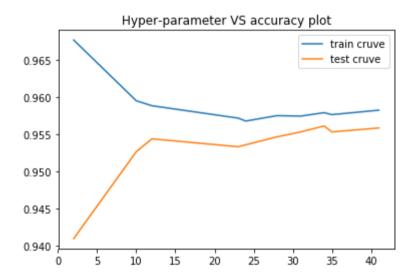
0% | 0/10 [00:00<?, ?it/s]

[2, 10, 12, 23, 24, 28, 31, 34, 35, 41]

{'n\_neighbors': [2, 10, 12, 23, 24, 28, 31, 34, 35, 41]}

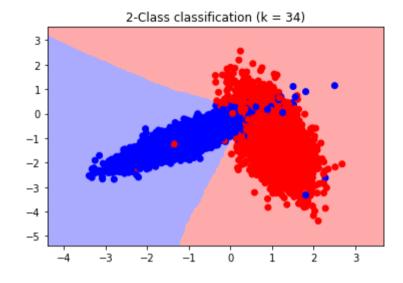
[0.9677333333333333, 0.95953333333333, 0.95886666666666, 0.9571999999999, 0.9568, 0.95753333333335, 0.957466666666668, 0.95793333333333334, 0.957666666666668, 0.95826666666667]

[0.9409333333333, 0.952666666666667, 0.9544, 0.953333333335, 0.95360000000001, 0.95466666666667, 0.95533333333334, 0.95613333333333, 0.9553333333334, 0.95586666666668]



In [42]: 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()

In [43]: from matplotlib.colors import ListedColormap
 neigh = KNeighborsClassifier(n\_neighbors = 34)
 neigh.fit(X\_train, y\_train)
 plot\_decision\_boundary(X\_train[:, 0], X\_train[:, 1], y\_train, neigh)



In [44]: !jupyter nbconvert --to html KNN\_solve.ipynb

[NbConvertApp] Converting notebook KNN\_solve.ipynb to html [NbConvertApp] Writing 360538 bytes to KNN\_solve.html