```
In [50]: train_data = 'train_data.txt'
         train_label = 'train_label.txt'
         test_data = 'test_data.txt'
         test_label = 'test_label.txt'
         import pandas as pd
         import numpy as np
         from cvxopt import matrix
         from cvxopt import solvers
         import matplotlib.pyplot as plt
In [51]: mean = [0]
         std = [0]
         # -- helper functions
         def getNumpyArray(file_name, preprocess = False, train = True):
             x = pd.read_csv(file_name, sep=" ", header=None)
             if preprocess:
                 if train:
                     mean[0] = x.mean(axis = 0)
                 x = x - mean[0]
                 if train:
                      std[0] = x.std(axis = 0)
                 x = x / std[0]
             x = x.to_numpy()
             return x
         def I(n):
             return np.eye(n)
In [52]: X_train = getNumpyArray(train_data, preprocess = True)
         Y_train = getNumpyArray(train_label).reshape(-1)
         X test = getNumpyArray(test data, preprocess = True, train = False)
         Y test = getNumpyArray(test label).reshape(-1)
         C = 0.5
In [53]: def train svm(X, Y, C, normalized = True):
             if not normalized:
                 mean[0] = X.mean(axis = 0)
                 X = X - mean[0]
                 std[0] = X.std(axis = 0)
                 X = X / std[0]
             N = X.shape[0]
             D = X.shape[1]
             assert len(Y.shape) == 1, 'dim error for label'
             assert Y.shape[0] == N, 'label shape[0] != N'
             temp = Y.reshape(-1,1) * X
             H = np.dot(temp , temp.T)
             P = matrix(H)*1.0 # - make it d type
             q = matrix(np.ones((N)) * -1)
             G = matrix(np.vstack((-I(N), I(N))))
             h = matrix(np.hstack((np.zeros(N), np.ones(N)*C)))
```

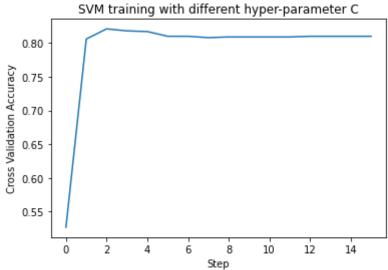
```
A = Y.reshape(1,-1)
    A = A.astype('float')
   A = matrix(A)
    b = matrix(np.array([0.0]))
    solvers.options['show progress'] = False # turn off the output
    sol = solvers.qp(P,q,G,h,A,b)
    alpha = sol['x']
    alpha = np.array(alpha).reshape(-1)
    assert len(alpha.shape) == 1, 'dim erro on alpha'
    assert alpha.shape[0] == N, 'alpha.shape[0] != N'
   w = np.dot((alpha * Y).T,X)
   w = w.reshape(-1)
   b = []
    for i in range(len(alpha)):
        if 10**-4 < alpha[i] < C:
            b.append(Y[i] - np.dot(w,X[i].T))
    assert len(w.shape) == 1, 'dim error on w'
    assert w.shape[0] == D, 'w.shape[0] != D'
   return w,b
def test_svm(X, Y, w, b, need_normalized = False):
    if need normalized:
        X = X - mean[0]
        X = X / std[0]
    best = 0
    for b in b:
       pred y = np.sign(np.dot(X,w)+b)
       Y=Y.reshape(-1)
        acc = np.sum(pred y==Y)/Y.shape[0]
        if acc > best:
            best = acc
    return best
```

```
In [54]: step = []
    acc_list = []

X_train = getNumpyArray(train_data)
Y_train = getNumpyArray(train_label).reshape(-1)

best_C = np.inf
best_acc = 0
# - choose C with cross-validation
for i,c in enumerate(range(-6, 10)):
    step.append(i)
    C = 4**c
    total_acc = []
    for f in range(5):
        lo, hi = f*200, (f+1)*200
```

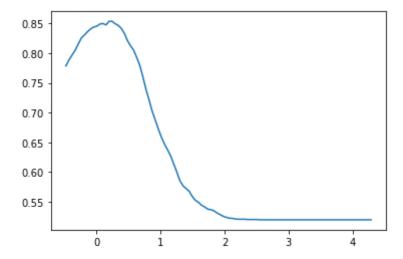
```
validation X = X train[lo:hi]
                   validation_Y = Y_train[lo:hi]
                   if lo > 0:
                       train_X = np.vstack((X_train[0:lo], X_train[hi:]))
                       train_Y = np.hstack((Y_train[0:lo], Y_train[hi:]))
                   else:
                       train X = X train[hi:]
                       train_Y = Y_train[hi:]
                   w,b = train_svm(train_X, train_Y, C, normalized = False)
                   acc_vali = test_svm(validation_X, validation_Y, w, b, need_normalized =
                   if acc_vali > best_acc:
                       best C = C
                       best_acc = acc_vali
                   total_acc.append(acc_vali)
              total acc = np.mean(total acc)
              acc_list.append(total_acc)
              print('C = {%5.4f} validation acc: {%5.4f}' %(C,total_acc))
          plt.title('SVM training with different hyper-parameter C')
          plt.xlabel('Step')
          plt.ylabel('Cross Validation Accuracy')
          plt.plot(step, acc_list)
          C = \{0.0002\} validation acc: \{0.5270\}
          C = \{0.0010\} validation acc: \{0.8060\}
          C = \{0.0039\} validation acc: \{0.8210\}
          C = \{0.0156\} validation acc: \{0.8180\}
          C = \{0.0625\} validation acc: \{0.8170\}
          C = \{0.2500\} validation acc: \{0.8100\}
          C = \{1.0000\}\ validation acc: \{0.8100\}\ 
          C = \{4.0000\} \text{ validation acc: } \{0.8080\}
          C = \{16.0000\} validation acc: \{0.8090\}
          C = \{64.0000\} validation acc: \{0.8090\}
          C = \{256.0000\} validation acc: \{0.8090\}
          C = \{1024.0000\} validation acc: \{0.8090\}
          C = \{4096.0000\} validation acc: \{0.8100\}
          C = \{16384.0000\} validation acc: \{0.8100\}
          C = \{65536.0000\}  validation acc: \{0.8100\}
          C = \{262144.0000\} validation acc: \{0.8100\}
          [<matplotlib.lines.Line2D at 0x7f8718087d30>]
Out [54]:
                    SVM training with different hyper-parameter C
```



```
In [55]: w,b = train_svm(X_train, Y_train, best_C)
B = np.linspace(min(b), max(b), 100)
```

```
acc_list = []
for _b in B:
    acc = test_svm(X_test, Y_test, w, [_b])
    acc_list.append(acc)
plt.plot(B, acc_list)
```

Out[55]: [<matplotlib.lines.Line2D at 0x7f87621c85b0>]



```
In [ ]:
```