

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import cvxpy as cp # I couldn't get this one to work
from cvxopt import matrix as cv_matrix
from cvxopt import solvers as cv_solvers
from sklearn.svm import SVC # for checking
```

## Problem 3 - SVM Implementation

- Implement the SVM (both domain adaptation SVM and normal SVM) using a quadratic program
- Do not need to build a classifier (i.e. solve for bias)
- Find  $W_T$  (target hyperplane) and  $W_S$  (source hyperplane)
- To find weight vectors, solve a quadratic programming problem and learn how to solve optimization task:
  - CVXOPT(<https://cvxopt.org/>)
  - PyCVX (<https://www.cvxpy.org/install/>)

## Resources:

- Quadratic Program Thoughts
- General SVM: <https://scikit-learn.org/stable/modules/svm.html>
- [https://xavierbourretsicotte.github.io/SVM\\_implementation.html](https://xavierbourretsicotte.github.io/SVM_implementation.html)

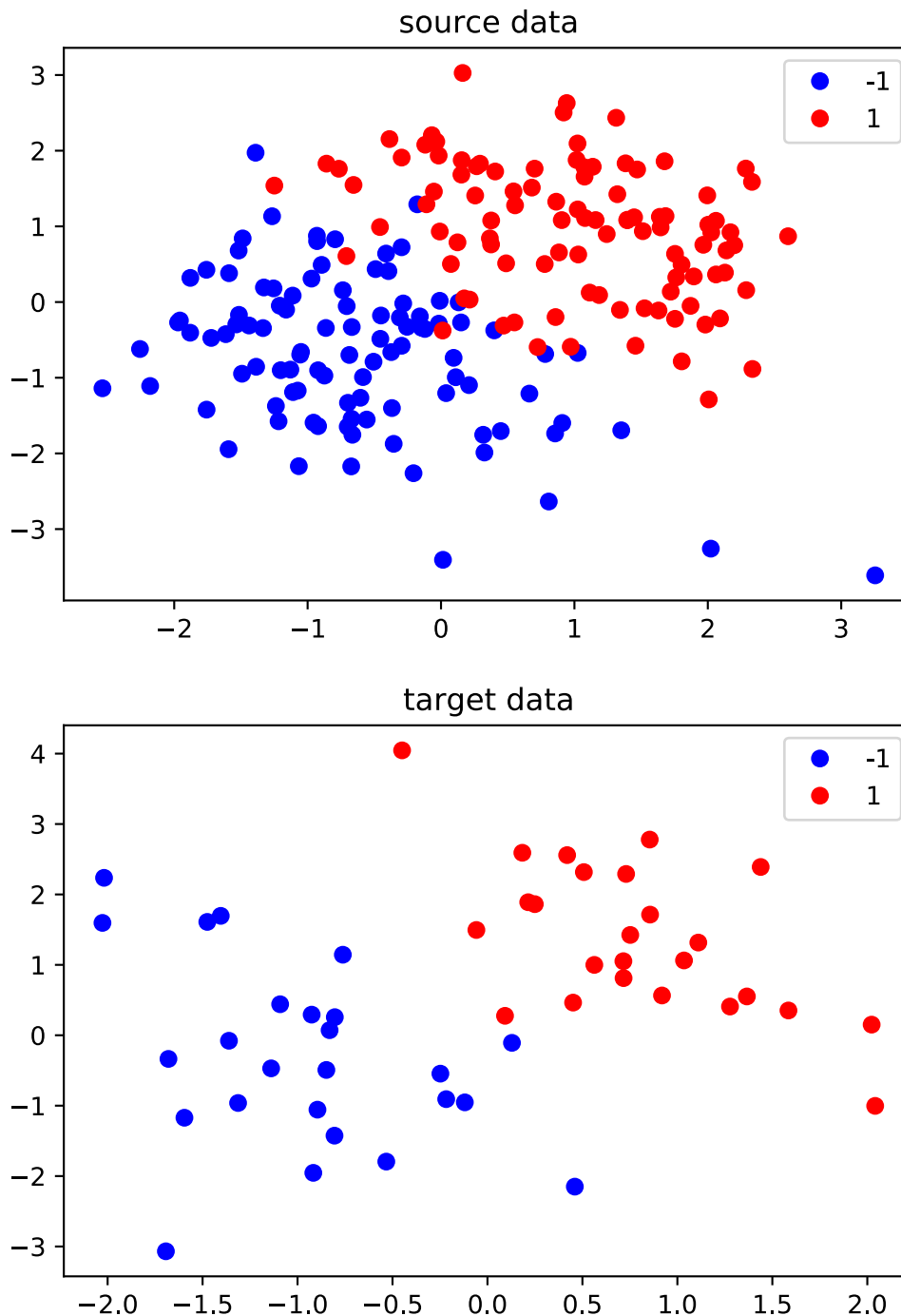
```
In [3]: # read data
s_arr = pd.read_csv('hw3_data/source_train.csv', header=None).to_numpy()
t_arr = pd.read_csv('hw3_data/target_train.csv', header=None).to_numpy()

# plot
# color list for graphing
color_list = ['b', 'r']
cla_l = [-1, 1]
for cla in range(2):
    lab = str(cla_l[cla])
    plt.plot(s_arr[:,0][s_arr[:,1] == cla_l[cla]], s_arr[:,1][s_arr[:,1] == cla_l[cla]], color_list[cla])

plt.legend()
plt.title('source data')
plt.show()

# plot
# color list for graphing
color_list = ['b', 'r']
cla_l = [-1, 1]
for cla in range(2):
    lab = str(cla_l[cla])
    plt.plot(t_arr[:,0][t_arr[:,1] == cla_l[cla]], t_arr[:,1][t_arr[:,1] == cla_l[cla]], color_list[cla])

plt.legend()
plt.title('target data')
plt.show()
```



1. solve for alpha source using a quad program

```
In [4]: # 1. solve for alpha source using a quad program
        # form:  $\min (1/2 * \alpha.T * H * \alpha + F.T * \alpha)$ 
        # s.t.  $A * \alpha = b, Q * \alpha \leq d$ 

        # free parameters
        C, B = 10, 0 # B = 1 if domain adaptive SVM, else B = 0

        X = s_arr[:, :-1]
        # print(X.shape)
        y = s_arr[:, -1]
        m, n = X.shape
        y = y.reshape(-1, 1) * 1.
        # print(y.shape)
```

```

# Hij = (yi * yj * xi.T * xj) --> shape = m*m
H = cv_matrix(np.matmul(y,y.T) * np.matmul(X,X.T))
# F = vector of -1 for generic SVM --> shape = m,1
F = cv_matrix(np.ones((m, 1)) * -1)
# A = vector of y, b = 0 (from constraint sum(yi*alpha) = 0) --> A shape = 2
A, b = cv_matrix(y.reshape(1, -1)), cv_matrix(np.zeros(1))
# Q = I matrix of -1 and 1, stacked; d = vector matrix of 0 and C --> Q sha
# note more commonly referred to as G = Q, and h = d
Q, d = cv_matrix(np.vstack((np.eye(m) * -1, np.eye(m)))), cv_matrix(np.hstack((n

solution = cv_solvers.qp(H, F, Q, d, A, b)
alphas = np.array(solution['x'])

```

	pcost	dcost	gap	pres	dres
0:	-2.8056e+02	-1.3709e+04	4e+04	8e-01	2e-14
1:	-1.5968e+02	-3.5932e+03	5e+03	7e-02	2e-14
2:	-1.2499e+02	-7.8459e+02	8e+02	9e-03	2e-14
3:	-1.7810e+02	-3.7504e+02	2e+02	2e-03	1e-14
4:	-1.9469e+02	-3.3774e+02	2e+02	1e-03	1e-14
5:	-2.1025e+02	-2.9579e+02	9e+01	6e-04	1e-14
6:	-2.0893e+02	-2.8104e+02	7e+01	3e-04	1e-14
7:	-2.1499e+02	-2.7779e+02	6e+01	2e-04	1e-14
8:	-2.3154e+02	-2.5248e+02	2e+01	5e-05	2e-14
9:	-2.3172e+02	-2.4828e+02	2e+01	2e-05	2e-14
10:	-2.3606e+02	-2.4070e+02	5e+00	7e-07	2e-14
11:	-2.3652e+02	-2.4052e+02	4e+00	5e-07	1e-14
12:	-2.3765e+02	-2.3861e+02	1e+00	2e-08	2e-14
13:	-2.3810e+02	-2.3811e+02	1e-02	2e-10	2e-14
14:	-2.3810e+02	-2.3810e+02	1e-04	2e-12	1e-14

Optimal solution found.

1. solve for ws using closed form ws

```

In [5]: # 2. solve for ws using closed form ws
# ws = sum(alphai*yi*xi) (should be a 1xm vector)

ws = np.zeros(2)
for j in range(m):
    # print(alphas[j])
    ws += alphas[j]*y[j]*X[j,:]

print('w source using for loop', ws)

w_s = ((y * alphas).T @ X).reshape(-1,1)
S = (alphas > 1e-4).flatten()
b = y[S] - np.dot(X[S], w_s)

print('w source using Q_Prog= ', w_s.flatten())

clf = SVC(C = 10, kernel = 'linear')
clf.fit(X, y.ravel())

print('w source using SKLearn ',clf.coef_.flatten())

w source using for loop [2.31500325 2.14257169]
w source using Q_Prog= [2.31500325 2.14257169]
w source using SKLearn [2.3138911 2.14154075]

```

1. solve for alpha target using a quad program

```
In [6]: # 3. solve for alpha target using a quad program

# free parameters
C, B = 10, 1 # B = 1 if domain adaptive SVM, else B = 0

X = t_arr[:, :-1]
# print(X.shape)
y = t_arr[:, -1]
m, n = X.shape
y = y.reshape(-1, 1) * 1.
# print(y.shape)

# Hij = (yi * yj * xi.T * xj) --> shape = m*m
H = cv_matrix(np.matmul(y, y.T) * np.matmul(X, X.T))
# F = vector of -(1 - B*yi*xi.T*w_s) for DA SVM --> shape = m, 1
F = np.zeros(m).reshape(-1, 1)
for i in range(F.shape[0]):
    # print(y[i].shape)
    # print(w_s.shape)
    # print(X[i, :].shape)
    # print(y[i]*np.matmul(X[i, :].T, w_s))
    F[i] = -(1 - B*y[i]*np.matmul(X[i, :].T, w_s))

print(F.shape)
F = cv_matrix(F)
# A = vector of y, b = 0 (from constraint sum(yi*alpha) = 0) --> A shape = 2
A, b = cv_matrix(y.reshape(1, -1)), cv_matrix(np.zeros(1))
# Q = I matrix of -1 and 1, stacked; d = vector matrix of 0 and C --> Q sha
# note more commonly referred to as G = Q, and h = d
Q, d = cv_matrix(np.vstack((np.eye(m) * -1, np.eye(m)))), cv_matrix(np.hstack((n

solution = cv_solvers.qp(H, F, Q, d, A, b)
alphas = np.array(solution['x'])

(50, 1)
      pcost      dcost      gap      pres      dres
0: -2.5826e+01 -3.9444e+03 1e+04 1e+00 2e-15
1:  4.4777e+01 -1.1862e+03 2e+03 8e-02 4e-15
2:  3.6709e+01 -1.3060e+02 2e+02 7e-03 3e-15
3:  4.9820e+00 -3.6256e+01 5e+01 1e-03 7e-16
4: -8.3141e+00 -1.5697e+01 8e+00 2e-04 7e-16
5: -1.1894e+01 -1.2316e+01 4e-01 5e-06 6e-16
6: -1.2191e+01 -1.2195e+01 5e-03 5e-08 5e-16
7: -1.2194e+01 -1.2194e+01 5e-05 5e-10 6e-16
8: -1.2194e+01 -1.2194e+01 5e-07 5e-12 6e-16
Optimal solution found.
```

1. solve for wt using closed form wt

```
In [7]: # 4. solve for wt using closed form wt
# wt = sum(alphai*yi*xi) - Bws (should be a 1xn vector)

wt = np.zeros(2)
for j in range(m):
    # print(alphas[j])
    # print(y[j])
    # print(X[j, :])
    # print(w_s.shape)
    wt += alphas[j]*y[j]*X[j, :] - B*w_s.flatten()

print('w target using for loop', wt)
```

```
w target using for loop [-113.94564524 -106.3400826 ]
```

In [ ]: