

# M4-L1 Problem 1

In this problem, you will perform support vector classification on a linearly separable dataset. You will do so without using an SVM package

That is, you will be solving the large margin linear classifier optimization problem:

$$\min_{w,b} \quad \frac{1}{2} \|w\|^2$$

$$\text{subject to: } y_i(w^T x_i + b) \geq 1$$

As described in lecture, you will convert the problem into a form compatible with the quadratic programming solver in the `cvxopt` package in Python:

$$\min \quad \frac{1}{2} x^T P x + q^T x$$

$$\text{subject to: } Gx \preceq h; Ax = b$$

Your job in this notebook is to define `P`, `q`, `G`, and `h` from above.

Please install the `cvxopt` package. (You can do that in the notebook directly with `!pip install cvxopt`) Then run the next cell to make the necessary imports.

```
In [8]: # Import modules
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap

from cvxopt import matrix, solvers
solvers.options['show_progress'] = False

def plot_boundary(x, y, w1, w2, b, e=0.1):
    x1min, x1max = min(x[:,0]), max(x[:,0])
    x2min, x2max = min(x[:,1]), max(x[:,1])

    xb = np.linspace(x1min, x1max)
    y_0 = 1/w2*(-b-w1*xb)
    y_1 = 1/w2*(1-b-w1*xb)
    y_m1 = 1/w2*(-1-b-w1*xb)

    cmap = ListedColormap(["purple", "orange"])

    plt.scatter(x[:,0], x[:,1], c=y, cmap=cmap)
    plt.plot(xb, y_0, '--', c='blue')
    plt.plot(xb, y_1, '--', c='green')
    plt.plot(xb, y_m1, '--', c='green')
    plt.xlabel('$x_1$')
```

```
plt.ylabel('$x_2$')
plt.axis((x1min-e,x1max+e,x2min-e,x2max+e))
```

## Load the data

```
In [9]: x1 = np.array([0.0478, 1.4237, 0.2514, 0.2549, 0.3378, 0.5349, 0.7319, 0.7768,
    0.6593, 0.9807, 0.877 , 0.8321, 0.6524, 1.4231, 1.2814, 1.3021,
    1.1915, 1.0913, 1.4438, 0.0959, 0.0752, 0.1789, 0.2549, 0.324 ,
    0.4934, 0.5971, 0.6005, 0.718 , 0.5452, 0.2272, 0.7802, 0.9565,
    0.1028, 0.0579, 0.1927, 0.3862])

x2 = np.array([ 0.9555, -0.396 ,  0.8968,  0.7987,  0.7251,  0.5453,  0.5371,
    0.7088,  0.8028,  0.766 ,  0.439 ,  0.1733,  0.3082,  0.9213,
    0.6515,  0.3777,  0.1896, -0.1374,  0.112 ,  0.3368,  0.1569,
    0.2101,  0.3368,  0.2509,  0.1651,  0.0343, -0.1169, -0.2355,
    -0.3009, -0.3091, -0.3418, -0.3377, -0.3091, -0.0188,  0.0547, -0.3091])

y = np.array([ 1,  1,  1,  1,  1,  1,  1,  1,  1,  1,  1,  1,  1,  1,  1,  1,  1,
    1,  1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1])

X = np.vstack([x1,x2]).T
```

## Quadratic Programming

Create the P, q, G, and h matrices as described in the lecture:

- **P** (3x3): Identity matrix, but with 0 instead of 1 for the bias (third) row/column
- **q** (3x1): Vector of zeros
- **G** (Nx3): Negative y multiplied element-wise by [ x1 , x2 , 1 ]
- **h** (Nx1): Vector of -1

Make sure the sizes of your matrices match the above. Use numpy arrays. These will be converted into `cvxopt` matrices later.

```
In [10]: # YOUR CODE GOES HERE
# Define P, q, G, h
P = np.eye(3)
P[2,2] = 0

q = np.zeros(3)
G = -y.reshape(-1, 1) * np.concatenate([X, np.ones((X.shape[0], 1))], axis=1)
h = -np.ones(X.shape[0])

print("P: ",P.shape)
print("q: ",q.shape)
print("G: ",G.shape)
print("h: ",h.shape)
```

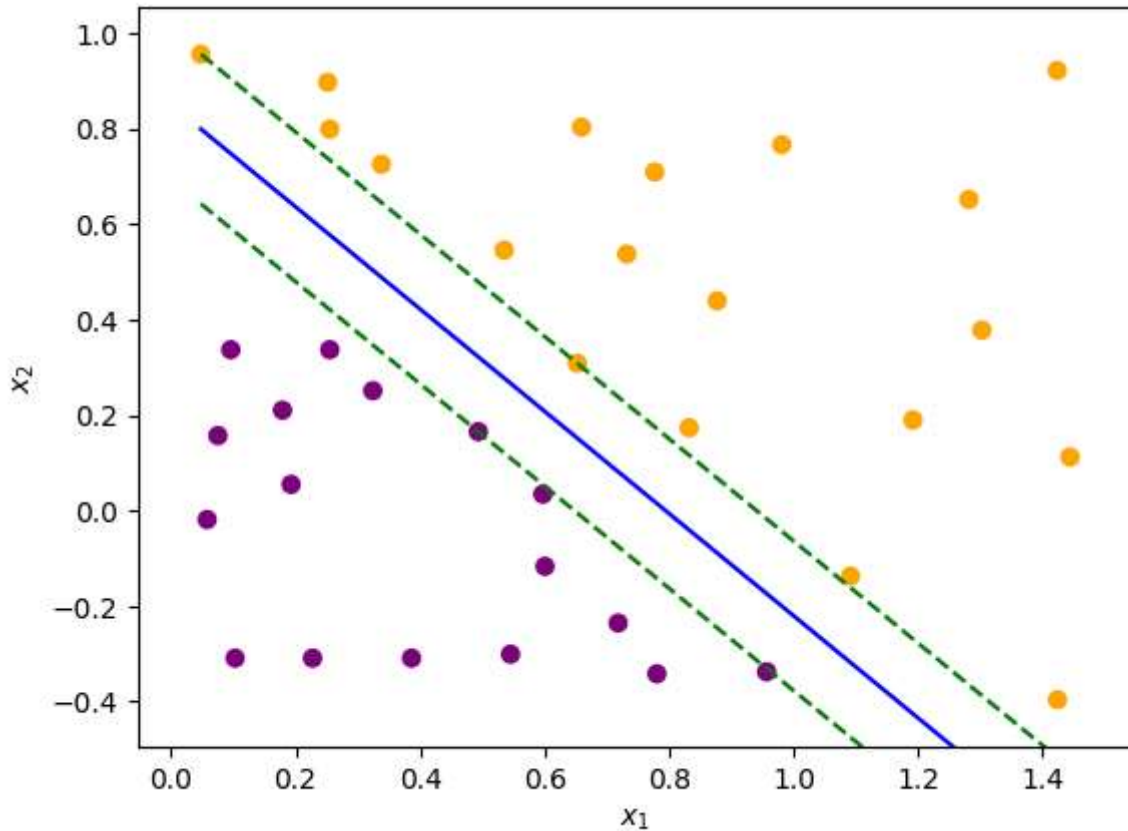
P: (3, 3)  
 q: (3,)  
 G: (36, 3)  
 h: (36,)

## Using cvxopt for QP

Now we convert these arrays into `cvxopt` matrices and solve the quadratic programming problem. Then we get the weights `w1`, `w2`, and `b` and plot the decision boundary.

```
In [11]: z = solvers.qp(matrix(P),matrix(q),matrix(G),matrix(h))
w1 = z['x'][0]
w2 = z['x'][1]
b = z['x'][2]

plot_boundary(X, y, w1, w2, b)
```



## Using the SVM

Finally, we will generate a grid of  $(x_1, x_2)$  points and evaluate our support vector classifier on each of these points. Given the array `X_grid`, determine `y_grid`, the class of each point in `X_grid` according to the support vector machine you trained.

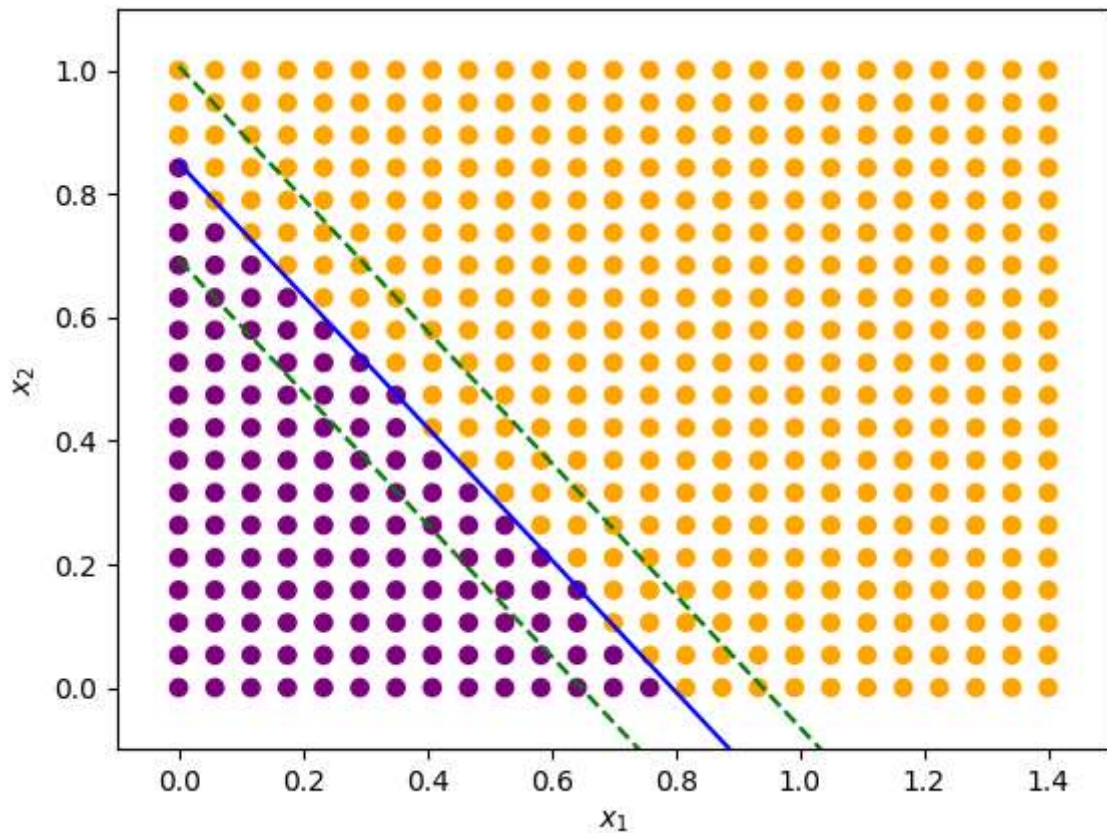
```

In [15]: x1vals = np.linspace(0,1.4,25)
x2vals = np.linspace(0,1,20)
x1s, x2s = np.meshgrid(x1vals, x2vals)
X_grid = np.vstack([x1s.flatten(),x2s.flatten()]).T

# YOUR CODE GOES HERE
# Get y_grid
y_grid = np.sign(w1*X_grid[:,0] + w2*X_grid[:,1] + b)

plot_boundary(X_grid, y_grid, w1, w2, b)

```



In [ ]: