## CS 4O03 Assignment 1

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```
In [1]: import scipy.optimize as opt
    import numpy as np
    import matplotlib.pyplot as plt
    import math
```

## **Question 1**

We are trying to formulate a plane s.t.

```
Ax >= B for all points in S1 Ax <= B for all points in S2
```

(or vice versa)

Since strict inequalities are not allowed and A != (0, 0, 0), that is to say simply formulating

min. 0

subject to.

```
Ax - B >= 0

Ax - B <= 0
```

Running the above solution through an optimizer will allow for A = (0, 0, 0) and therefore, we reach an inadequate solution. We reformulate the problem as

```
Ax >= B + 1

Ax <= B - 1
```

This, in the context of linear classification, will determine if the set of points are *linearly separable*. After changing both inequalities to upper bounds with a constant RHS

```
Ax + B <= -1

Ax - B <= -1
```

Now we can formulate the problem

Decision variables - Let:

```
A[i] represent the coefficient corresponding to x[i] B represent the constant of the plane
```

Therefore our LP becomes

min. 0

subject to.

We are *not* trying to find an optimal hyperplane separating the two classes, we are simply trying to find *a* hyperplane which separates them.

```
In [2]: # question 1 pt. 1
        A_ub = np.array([
        [-1, -2, -3, 1],
        [-3, -1, -2, 1],
        [-2, -3, -1, 1],
        [0, 2, 0, -1],
        [4, 2, 4, -1],
        [math.pi, math.log(math.pi), math.sqrt(2), -1]])
        b_ub = np.array([-1, -1, -1, -1, -1, -1])
        c = np.array([0, 0, 0, 0])
        res = opt.linprog(c, A_ub=A_ub, b_ub=b_ub,
        bounds=((None, None), (None, None), (None, None), (0, None)))
        print(res)
     fun: 4.602353985700042
 message: 'Optimization failed. Unable to find a feasible starting point.'
     nit: 6
 status: 2
 success: False
       x: nan
```

We see from the above optimization run, the problem we have formulated is infeasible as the optimizer cannot find a starting point, and therefore, there exists *no* plane which can separate the points and they are not *linearly separable*. To show that no perfect plane exists with a terminating optimization problem, we will reformulate the optimization problem into a linear classification problem such that

```
Ax - B >= 1 - u

Ax - B <= - (1 - v)
```

Where u and v represent a margin from a optimal hyperplane and the points. We then formulate the LP with:

Decision variables - Let

```
A[i] represent the coefficient corresponding to x[i] B represent the constant of the plane u[i] represent the distance between the plane + 1 and each point in S1 v[i] represent the distance between the plane - 1 and each point in S2 min. u[1] + u[2] + u[3] + v[1] + v[2] + v[3]
```

subject to.

```
+ B - u[1] <= -1
-1*A[0] - 2*A[1]
                       - 3*A[2]
-3*A[0] - 1*A[1]
                       -2*A[2]
                                      + B - u[2] <= -1
-2*A[0] - 3*A[1]
                        - 1*A[2]
                                       + B - u[3] <= -1
          2*A[1]
                                       - B - v[1] <= -1
 4*A[0] + 2*A[1]
                        + 4*A[2]
                                      - B - v[2] <= -1
pi*A[0] + log(pi)*A[1] + sqrt(2)*A[2] - B - v[3] <= -1
u[i], v[i] >= 0
```

We are now trying the minimize the number of points which are on the wrong side of the hyperplane (or in other words, misclassified). If the optimal objective value is 0, then we have found a perfect separating hyperplane. Otherwise the optimal objective value after optimization is the best we can do in terms of classification. Therefore, there is no perfect separating hyperplane *and* the parameters returned is *best* plane which separates as many points correctly as possible.

```
In [3]: # question 2 pt. 2
```

```
A_ub = np.array([
       [-1, -2, -3, 1, -1, 0, 0, 0, 0, 0]
       [-3, -1, -2, 1, 0, -1, 0, 0, 0, 0],
       [-2, -3, -1, 1, 0, 0, -1, 0, 0, 0],
       [0, 2, 0, -1, 0, 0, 0, -1, 0, 0],
       [4, 2, 4, -1, 0, 0, 0, 0, -1, 0],
       [math.pi, math.log(math.pi), math.sqrt(2), -1, 0, 0, 0, 0, -1]])
       b_ub = np.array([-1, -1, -1, -1, -1, -1])
       c = np.array([0, 0, 0, 0, 1, 1, 1, 1, 1, 1])
       res = opt.linprog(c, A_ub=A_ub, b_ub=b_ub,
       bounds=((None, None), (None, None), (None, None), (0, None), (0, None),
               (0, None), (0, None), (0, None), (0, None), (0, None)))
       print(res)
    fun: 4.602353985700042
message: 'Optimization terminated successfully.'
   nit: 11
 slack: array([0., 0., 0., 0., 0., 0.])
 status: 0
success: True
     x: array([0.04823451, 1.25294248, 0.6505885, 3.50588496, 0.
     1.80706196, 0.
                                        , 2.79529203, 0.
                                                               ])
                     , 0.
```

The optimal function value is ~4.6, therefore there exists no perfect separating hyperplane. The difference between the current LP and the previous LP is that the previous LP required a perfect separating hyperplane, whereas the current LP requires the most *optimal* hyperplane. If the problem could be formulated as a quadratic program, we could add a hyperparameter which regulates the margins.

Sources consulted:

https://en.wikipedia.org/wiki/Support\_vector\_machine https://yalmip.github.io/tutorial/linearprogramming/

## **Question 2**

```
Decision variables - Let
x[0] represent the number of tables we sell
x[1] represent the number of desks we sell
x[2] represent the number of chairs we sell
We formulate the LP as:
max. 100*x[0] + 50*x[1] + 10*x[2]
subject to.
25*[0] + 15*x[1] + 10*x[2] <= 4000
50*[0] + 30*x[1] + 10*x[2] <= 5000
90*[0] + 50*x[1] + 25*x[2] <= 6000
4*x[0] + x[1] \le x[2] # 4x chairs as tables + 1x chairs as desks <= total number of chairs we set
x[0], x[1], x[2] >= 0
In [4]: # question 2 pt. 2
        A_ub = np.array([
        [25, 15, 10],
        [50, 30, 10],
        [90,50,25],
        [4, 1, -1]]
        b_ub = np.array([4000, 5000, 6000, 0])
        c = np.array([100, 50, 10])
        # default is a min problem. max f(x) = -min f(-x)
        res = opt.linprog(-c, A_ub=A_ub, b_ub=b_ub,
        bounds=(0, None))
        print(res)
     fun: -4800.0
 message: 'Optimization terminated successfully.'
   slack: array([2000., 1800.,
                                          0.])
  status: 0
success: True
       x: array([ 0., 80., 80.])
```

The maximum revenue is \$4800. The optimal product mix is 80 desks and 80 chairs.

## **Question 3**

```
Decision variables - Let
```

```
x[i] be the number of trees we can buy normally in week i
z[3] be the number of extra trees we can buy in week 3
z[4] be the number of extra trees we can buy in week 4

Other variables - Let
y[1] be the number of trees left over in week 1
y[2] be the number of trees left over in week 2
y[3] be the number of trees left over in week 3
```

\* there should be no trees left over in week 4

We are trying to minimize the cost in order to meet the demand of trees every week, therefore We formulate the LP as:

```
min. 100*(x[1] + x[2]) + 150*(x[3] + x[4]) + 3*(y[1] + y[2] + y[3]) + 200*(z[3] + z[4]) subject to.
```

```
x[1] - 70
                        == y[1]
x[2] + y[1] - 80
                        == y[2]
x[3] + y[2] + z[3] - 90 == y[3]
x[4] + y[3] + z[4] - 100 == 0 # no leftover trees after 4th week
x[1], x[2], x[3], x[4]
                        <= 90
z[3], z[4]
                        <= 20
x[1], x[2], x[3], x[4]
                        >= 0
y[1], y[2], y[3]
                        >= 0
z[3], z[4]
                        >= 0
```