hw6

September 29, 2018

1 Week 6 HW

1.1 Question 1

Implement the stochastic inventory control model in CVXPY with the following data.

- (a) Demand d can take 5 values: $d_1 = 10$, $d_2 = 20$, $d_3 = 30$, $d_4 = 40$, $d_5 = 50$, with probability $p_1 = 0.1$, $p_2 = 0.15$, $p_3 = 0.3$, $p_4 = 0.25$, $p_5 = 0.2$, respectively.
- (b) Unit cost c = 10, retail price r = 15, discount price s = 5.
- (c) Production capacity $\bar{x} = 75$.

Solution The information above indicates there are **5** scenarios. Therefore we will have 11 total variables (*1 for the 1st stage and 10 for the second stage.*)

```
In [22]: import cvxpy as cp
         import numpy as np
         # given values
         n = 5 \# scenario count
         d = np.array([10., 20., 30., 40., 50.])
         p = np.array([0.1, .15, .30, .25, 0.2])
         c = 10.
         r = 15.
         s = 5.
         xbar = 75.
         #lp variables
         x = cp.Variable()
         y = cp.Variable(n,1)
         z = cp.Variable(n,1)
         #problem setup
         objective = cp.Minimize(c*x + p*(-r*y-s*z))
         constraints = [x >= 0., x <= xbar, y >= 0., y <= d, z >= 0., -x + y + z <= 0.]
```

```
#solve
         prob = cp.Problem(objective, constraints)
         result = prob.solve()
         #output results
         print('The problem status is', prob.status)
         print('The objective value is', round(result, 2))
         #print(x.value)
         #print(y.value)
         #print(z.value)
         print('The production quanity is', x.value.round(2))
         print('The retail qty for each scenario is', y.value.round(2))
         print('The sale qty for each scenario is', z.value.round(2))
The problem status is optimal
The objective value is -115.0
The production quanity is 30.0
The retail qty for each scenario is [10. 20. 30. 30. 30.]
The sale qty for each scenario is [20. 10. 0. 0. 0.]
```

1.2 Question 2

Given a set of training data...build the classifier using the absolute deviation regression (ADR).

- (a) Is the objective function of (ADR) a convex function in 0, . . . , n? ADR is convex because it is a sum of convex functions.
- **(b) Write a linear programming reformulation of (ADR).** In the given data, the *X* array contains 2 features with 100 observations. Therefore we need 3 variables β_0 , β_1 , and β_2 . We also need 100 variables to support the LP formulation.

$$\min_{\beta_0,\beta_1,\beta_2} \sum_{i=0}^{100} z_i$$

$$\text{s.t.}\beta_0 + \sum_{j=1}^2 x_{ij}\beta_j - y_i \le z_i \forall i \in \{1, 2, ..., 100\}$$

$$\beta_0 + \sum_{j=1}^2 x_{ij}\beta_j - y_i \ge -z_i \forall i \in \{1, 2, ..., 100\}$$

(c) Code your LP reformulation of (ADR) in CVXPY, using the data file provided. I ended up with 2 implementations. The first is a direct implementation of the ADR. The other is the LP formulation. Both yield the same result.

```
In [26]: # direct implementation of ADR
    import numpy as np
```

```
# load the file data (I created these files from the given file)
         xs = np.loadtxt('x.dat')
         #print(xs.shape)
         ys = np.loadtxt('y.dat')
         #print(ys.shape)
         #lp variables
         x = cp.Variable(3,1)
         ones = np.ones([xs.shape[0], 1])
         A = np.hstack([ones, xs])
         #print(A)
         #problem setup
         objective = cp.Minimize(cp.sum(cp.abs(ys - A*x)))
         constraints = None
         #solve
         prob = cp.Problem(objective, constraints)
         result = prob.solve()
         #output results
         print('The problem status is', prob.status)
         print('The objective value is', round(result, 2))
         params = x.value
         print('The hyperplane parameters are', params.round(4))
The problem status is optimal
The objective value is 26.52
The hyperplane parameters are [ 0.4036 0.185 -0.1995]
In [27]: # lp formulation
        m = xs.shape[1] + 1
         n = xs.shape[0]
         b = cp.Variable(m, 1)
         z = cp.Variable(n, 1)
         ones = np.ones([n, 1])
         A = np.hstack([ones, xs])
         #problem setup
         objective = cp.Minimize(cp.sum(z))
         constraints = [A*b - ys \le z, A*b - ys \ge -z]
         #solve
```

```
prob = cp.Problem(objective, constraints)
    result = prob.solve()

#output results
    print('The problem status is', prob.status)
    print('The objective value is', round(result, 2))
    params = b.value
    print('The hyperplane parameters are', params.round(4))

The problem status is optimal
The objective value is 26.52
The hyperplane parameters are [ 0.4036  0.185  -0.1995]
```

yy = a * xx + (.5 - params[0]) / params[2]

In [29]: import matplotlib.pyplot as plt

plt.plot(xx, yy, 'k-');

plt.show();

(d) Write a Python code to plot the data points and the hyperplane obtained from (ADR).

```
plt.scatter(xs[:,0], xs[:,1], c=np.where(ys == 1, 'y', 'k'));
plt.xlabel("x1");
plt.ylabel("x2");

xx = np.linspace(-2, 4.5)
# plot b0 + b1*x1 + b2*x2 >= .5 which is x2 >= -b1/b2*x1 + (.5 - b0) / b2
a = -params[1] / params[2]
```

