

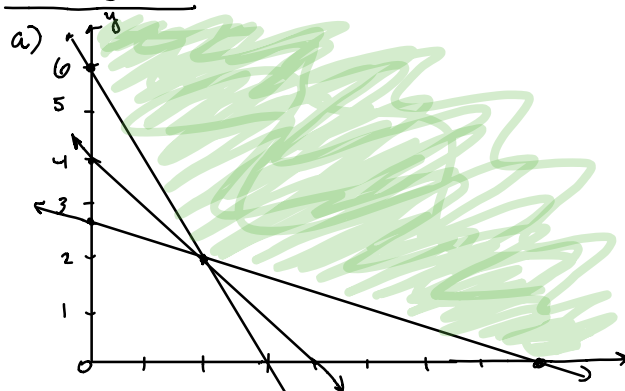
PROBLEM 1

- a) The lowest value that the scroll bars will adjust to for the sum of the residuals is 77.947. The optimal linear estimator is $y = -0.67x + 20.61$ with the sum of the residuals = 77.894.
- b) The lowest value that the scroll bars will adjust to is 849.14. The optimal linear estimator is $y = 0.46x + 33.79$ with the sum of the residuals = 849.14.

PROBLEM 2

- a) decision variables: amount of radiation dose to healthy tissue voxels V_1, V_3, V_4 , and V_9
 Objective function: minimize $f = V_1 + V_3 + V_4 + V_5 + V_9$
 constraints:
 $V_1 = 1.0(b_1 + b_4)$ $V_3 = 2.0(b_1 + b_6)$ $V_4 = 1.0(b_2 + b_4)$ $V_9 = 2.5(b_3 + b_6)$
 $V_2, V_6, V_7, V_8 \geq 7 \rightarrow 2(b_1 + b_5), 2.5(b_2 + b_6), 1.5(b_3 + b_4), 1.5(b_3 + b_5) \geq 7$
 $V_5 \leq 5 \rightarrow 2(b_2 + b_5) \leq 5$
- b) Optimal solution = 31.017 with V_2 receiving 4 doses, V_5 receiving 4, and V_8 receiving 3. This does make sense intuitively because you are mostly hitting the cancerous voxels and hitting the other voxels less.
- c) Max dose: 4 optimal solution: 31.7667
 3 opt. soln: 32.517
 2 opt. soln: 33.267
 1 opt. soln: 34.017
 0 opt. soln: 34.767
- The trade-off for hitting less of the spinal cord means that you will hit more of the surrounding healthy tissue, since the optimal solution increases as you decrease the max. dose to the spinal cord.

PROBLEM 3



extreme pts: $(2, 2)$, $(8, 0)$, $(0, 6)$

b) the optimal solution occurs at $(2, 2)$ and is equal to 4

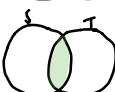
c) $x + y \geq 4$ is redundant

d) there is only one optimal solution because the objective function value at $(8, 0)$ and $(0, 6)$ is greater

e) since the region is not bounded on all sides, it is not possible to find the optimal solution for maximizing. You could keep increasing x and y to ∞ .

PROBLEM 4

a) i.  not convex

ii.  convex

iii.  not convex

b) i. True

ii. True

iii. False

c) i. True

ii. True

iii. True

iv. False

PROBLEM 5

- a) the optimal solution is 2.0 with $x_1 = 0$ and $x_2 = 2$ when the constraint = 6. When the constraint = 7, the optimal solution is 2.333 with $x_1 = 0$ and $x_2 = 2.333$.

```
1 #PROBLEM 5A
2
3 using JuMP, Gurobi
4 model = Model(with_optimizer(Gurobi.Optimizer))
5
6 #variables
7 @variable(model, x1>=0)
8 @variable(model, x2>=0)
9
10 @objective(model, Max, 2x1+x2)
11
12 #constraints
13 @constraint(model, 12x1+3x2<=7)
14 @constraint(model, -3x1+x2<=7)
15 @constraint(model, x2<=10)
16
17 model
```

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```
max 2x1 + x2
Subject to
    x1 ≥ 0.0
    x2 ≥ 0.0
    12x1 + 3x2 ≤ 7.0
    - 3x1 + x2 ≤ 7.0
    x2 ≤ 10.0
```

```
1 optimize!(model)
2
3 println("Objective Value: ", objective_value(model))
4 println("x1: ", value(x1))
5 println("x2: ", value(x2))
```

Optimize a model with 3 rows, 2 columns and 5 nonzeros

Coefficient statistics:

```
Matrix range [1e+00, 1e+01]
Objective range [1e+00, 2e+00]
Bounds range [0e+00, 0e+00]
RHS range [7e+00, 1e+01]
```

Presolve removed 3 rows and 2 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	2.3333333e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds

Optimal objective 2.3333333333333335

Objective Value: 2.3333333333333335

x1: 0.0

x2: 2.3333333333333335

- b) The optimal solution is 0.3007 with $x_1 = 2.4$, $x_2 = 0$, $x_3 = 8.1$, $x_4 = 0$, $x_5 = 0.5$, and $y = 1$. With the given changes, the optimal solution is 0.2887 with $x_1 = 1.9$, $x_2 = 0$, $x_3 = 8.75$, $x_4 = 0$, $x_5 = 0.35$, and $y = 1$.

```

In [15]: 1 #PROBLEM 5B
2
3 using JuMP, Gurobi, CSV
4 model = Model(with_optimizer(Gurobi.Optimizer))
5
6 data = CSV.read("ps2_Portfolio1.csv", header = true)
7 rating = data[:,4]
8 maturity = data[:,5]
9 afterTax = data[:,8]
10
11 AvgBankQuality = 1.3
12 AvgYearsToMaturity = 4.8
13
14
15 @variable(model, x[1:5]>=0)
16 @variable(model, 0<=y<=1)
17
18 println(float(rating[1]))
19 @constraint(model, (sum(x[1:5])-y)<=10)
20 @constraint(model, x[2]+x[3]+x[4]>=4)
21 @constraint(model, sum((rating[i])*x[i] for i in 1:5)<=AvgBankQuality*sum(x[1:5]))
22 @constraint(model, sum(maturity[i]*x[i] for i in 1:5)<=AvgYearsToMaturity*sum(x[1:5]))
23
24 @objective(model, Max, (sum(afterTax[i]*x[i] for i in 1:5)-0.0275*y))
25
26 optimize!(model)
27
28 println("Objective Value: ", objective_value(model))
29 for i in 1:5
30     println("x", i, ": ", value(x[i]))
31 end
32 println("y: ", value(y))

```

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2.0

Optimize a model with 4 rows, 6 columns and 19 nonzeros

Coefficient statistics:

```

Matrix range      [3e-01, 1e+01]
Objective range    [2e-02, 4e-02]
Bounds range       [1e+00, 1e+00]
RHS range          [4e+00, 1e+01]

```

Presolve time: 0.00s

Presolved: 4 rows, 6 columns, 19 nonzeros

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	1.6200000e+29	7.250000e+30	1.620000e-01	0s
3	2.8870000e-01	0.000000e+00	0.000000e+00	0s

Solved in 3 iterations and 0.00 seconds

Optimal objective 2.887000000e-01

Objective Value: 0.2887

x1: 1.8999999999999995

x2: 0.0

x3: 8.750000000000002

x4: 0.0

x5: 0.3500000000000003

y: 1.0

PROBLEM 6

a) decision variables: x_0, y_0 = the x,y coordinates of the coordinates of the warehouse.

$$\text{Objective function: Min. } z = \sum_{i=1}^5 \text{shipment}_i \sqrt{(x_0 - x_i)^2 + (y_0 - y_i)^2}$$

shipment_i = # shipments/year for customer i

x_i, y_i = x,y coordinates of customer i

b) optimal location: (3.068, 7.076)

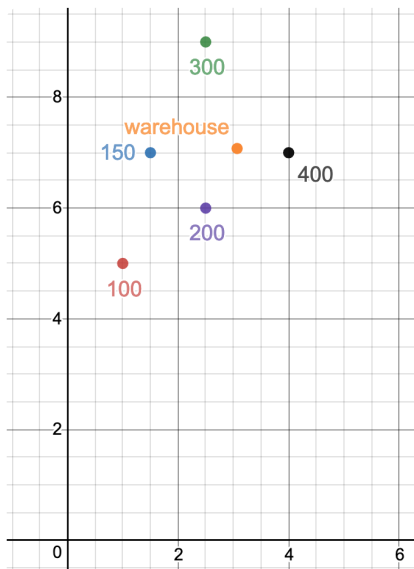
minimized total distance value: 1747.71

```
1 using JuMP, Ipopt
2 model = Model(with_optimizer(Ipopt.Optimizer))
3
4 x = [1, 1.5, 2.5, 2.5, 4]
5 y = [5, 7, 9, 6, 7]
6 shipments = [100, 150, 300, 200, 400]
7
8 @variable(model, x0)
9 @variable(model, y0)
10
11 @NLObjective(model, Min, sum(shipments[i]*sqrt((x0-x[i])^2+(y0-y[i])^2) for i in 1:5))
12
13 optimize!(model)
```

```
1 println("Objective Value: ", objective_value(model))
2 println("x0: ", value(x0))
3 println("y0: ", value(y0))
```

Objective Value: 1747.7096891328363
x0: 3.0679192152498205
y0: 7.076400277458581

c)



This does seem correct because the warehouse is located closer to the customers with higher shipment weights.