ISyE524: Introduction to Optimization Problem Set #1

Due: February 11, 2024 at 11:59PM

Notes and Deliverables:

- Submit solutions to all problems in the form of a *single PDF file* of the IJulia notebook to the course dropbox.
- If you need to write answers on paper, you can take a picture of you work and add that to your IJulia notebook

1 Graphical Method Extravaganza

Consider the following linear programming problem:

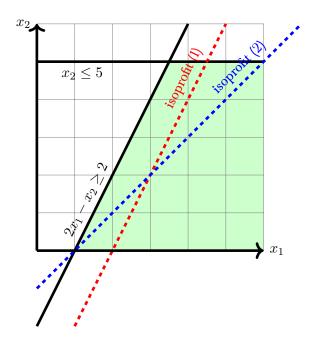
$$\max -4x_1 + 2x_2$$
 subject to:
$$x_2 \le 5$$

$$2x_1 - x_2 \ge 2$$

$$x_1, \quad x_2 \ge 0$$

1-1 Problem Find an optimal solution to this linear program using the graphical solution method. (Show the feasible region as part of your answer.)

Answer:



I drew the feasible region and isoprofit line in red. One optimal solution is $(x_1, x_2) = (1, 0)$ with optimal objective value -4.

1-2 Problem How many optimal solutions does this linear program have? If more than one, provide another optimal solution.

Answer:

There are infinitely many optimal solutions to this problem. Another optimal solution is $(x_1, x_2) = (3.5, 5)$, again with optimal objective value -4. In fact, any solution on the line segment between these two points is an optimal solution.

1-3 Problem Now suppose the objective is changed to maximize $-x_1 + x_2$. Answer questions 1-1 and 1-2 again.

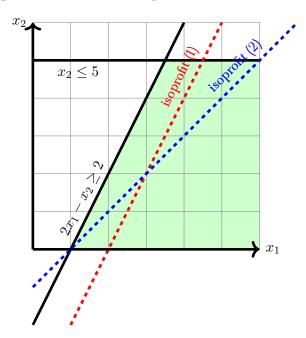
Answer:

I drew the isoprofit line for this problem in blue. Now there is a unique optimal solution at $(x_1, x_2) = (3.5, 5)$ with optimal objective value 1.5.

1-4 Problem Now suppose the objective is changed to minimize $-x_1 + x_2$. Does the LP now have an optimal solution? If so, provide one. If not, what do we call this LP?

Answer:

The feasible region is plotted below, and an isoprofit line is in blue.



The isoprofit line is the same slope as in Problem 1-3, however in this case, the value gets smaller as we move the line to the right. We can keep moving the line to the right infinitely far and continue intersecting the feasible region. The LP does not have an optimal solution. It is *unbounded*.

2 Standard Form

Consider the following LP:

$$\begin{aligned} \min 2x_1 - x_2 + 3x_3 - 2x_4 \\ \text{s.t. } 3x_2 - x_3 + 4x_4 &\geq 10 \\ 4x_1 - 7x_2 + x_3 - x_4 &= 5 \\ x_1 - 3x_3 + 8x_4 &\leq 3 \\ x_1, x_4 &\geq 0 \\ x_2, x_3 \text{ Unrestricted in Sign (Free)} \end{aligned}$$

2-1 Problem Implement the instance and Julia/JuMP and solve it, printing the optimal solution and optimal solution value.

Answer:

```
_ Julia Code -
      using JuMP, HiGHS
      m = Model(HiGHS.Optimizer)
3
      # I made an array of variables, and index them as necessary
       @variable(m, x[1:4])
5
6
       Objective(m, Min, 2x[1] - x[2] + 3x[3] - 2x[4])
7
       @constraint(m, x[1] >= 0)
8
       @constraint(m, x[4] >= 0)
       0constraint(m, 3x[2] - x[3] + 4x[4] >= 10)
10
       0constraint(m, 4x[1] -7x[2] + x[3] - x[4] == 5)
11
       0constraint(m, x[1] - 3x[3]+ 8x[4] <=3)
12
       optimize!(m)
13
      println("objective: ", objective_value(m))
14
15
      println("solution: ", value.(x)) # dot operator on value() returns value of each entry in the vector
```

Gives

2-2 Problem Convert the problem to the standard form:

$$\max_{x \in \mathbb{R}^n} \qquad c^\top x$$
 subject to:
$$Ax \le b$$

$$x \ge 0$$

What are A, b, c,and x in this problem? Clearly indicate how the decision variables of your transformed LP relate to those of the original LP.

Answer:

First, identify the issues. This is a min problem, and be want it to be max. We have two free variables (x_2, x_3) . We also have one nonpositive variable (x_4) . There is an equality constraint and a \geq constraint.

We'll start with the variables: Let $x_2 = v_2 - w_2$. Let $x_3 = v_3 - w_3$. Then x_1, v_2, w_2, v_3, w_3 , and $x_4 \ge 0$. Replace everywhere x_2 and x_3 appear:

$$\min 2x_1 - (v_2 - w_2) + 3(v_3 - w_3) - 2x_4$$

$$\text{s.t. } 3(v_2 - w_2) - (v_3 - w_3) + 4x_4 \ge 10$$

$$4x_1 - 7(v_2 - w_2) + (v_3 - w_3) - x_4 = 5$$

$$x_1 - 3(v_3 - w_3) + 8x_4 \le 3$$

$$x_1, v_2, w_2, v_3, w_3, x_4 \ge 0$$

Now let's simplify and fix the objective:

$$-\max -2x_1 + v_2 - w_2 - 3v_3 + 3w_3 + 2x_4$$
 s.t.
$$3v_2 - 3w_2 - v_3 + w_3 + 4x_4 \ge 10$$

$$4x_1 - 7v_2 + 7w_2 + v_3 - w_3 - x_4 = 5$$

$$x_1 - 3v_3 + 3w_3 + 8x_4 \le 3$$

$$x_1, v_2, w_2, v_3, w_3, x_4 \ge 0$$

Finally, fix the constraints:

$$\begin{aligned} &-\max -2x_1+v_2-w_2-3v_3+3w_3+2x_4\\ \text{s.t.} &&-3v_2+3w_2+v_3-w_3-4x_4\leq -10\\ &&4x_1-7v_2+7w_2+v_3-w_3-x_4\leq 5\\ &&-4x_1+7v_2-7w_2-v_3+w_3+x_4\leq -5\\ &&x_1&-3v_3+3w_3+8x_4\leq 3\\ &&x_1,v_2,w_2,v_3,w_3,x_4\geq 0 \end{aligned}$$

And we're done! We have

$$x = \begin{bmatrix} x_1 \\ v_2 \\ w_2 \\ v_3 \\ w_3 \\ x_4 \end{bmatrix} \quad A = \begin{bmatrix} 0 & -3 & 3 & 1 & -1 & -4 \\ 4 & -7 & 7 & 1 & -1 & -1 \\ -4 & 7 & -7 & -1 & 1 & 1 \\ 1 & 0 & 0 & -3 & 3 & 8 \end{bmatrix}$$

$$c = \begin{bmatrix} -2 & 1 & -1 & -3 & 3 & 2 \end{bmatrix}, \quad b = \begin{bmatrix} -10 \\ 5 \\ -5 \\ 3 \end{bmatrix}.$$

Note we also have $x_2 = v_2 - w_2$ and $x_3 = v_3 - w_3$.

2-3 Problem Solve the standard-form LP in Julia and report the objective value and the value of each decision variable in an optimal solution to the original LP.

Answer:

```
_ Julia Code _
     # Here I will do a "matrix form" of the LP,
     # showing off some more of Julia and Jump's operators,
2
     # especially for those who like matlab
3
4
5
     #standard form
     A = [0 -3 3 1 -1 -4; 4 -7 7 1 -1 -1; -4 7 -7 -1 1 1; 1 0 0 -3 3 8]
     c = [-2; 1; -1; -3; 3; 2]
8
    b = [-10; 5; -5; 3]
9
    using JuMP, HiGHS
10
    m = Model(HiGHS.Optimizer)
11
12
    @variable(m, x[1:6] >= 0)
13
    @objective(m, Max, c'*x)
    @constraint(m, A*x .<= b) # Note the '.' makes entrywise <=</pre>
14
15
    optimize!(m)
16
     # Here I report the translated values! So we should get exactly the same answer
    println("x1 = ", value(x[1]))
18
    println("x2 = ", value(x[2])-value(x[3]))
19
    println("x3 = ", value(x[4])-value(x[5]))
20
    println("x4 = ", value(x[6]))
21
     println("objective: ", -(objective_value(m)))
```

Gives matching solution:

```
Julia Answer

x1 = 7.593750000000002

x2 = 3.84375

x3 = 1.5312500000000007

x4 = 0.0

objective: 15.93750000000005
```

3 Mmmmmmmmmmmmm. Pork.

Jacob's Meat Packing Plant produces 480 hams, 400 pork bellies, and 230 picnic hams every day. Each of these products can be sold either fresh or smoked. The total number of hams, bellies, and picnics that can be smoked during a normal working day is 420. In addition, up to 250 products can be smoked on overtime, at a higher cost (and therefore a lower net profit). The net profits (in \$ per product) are the following:

	Fresh	Smoked (Regular)	Smoked (Overtime)
Hams	8	14	11
Bellies	4	12	7
Picnics	4	13	9

3-1 Problem Write a linear program that will maximize the total net profit.

Answer:

Define the following decision variables

- h_f : # of hams to sell fresh each day
- h_r : # of hams to smoke in regular time each day
- h_o : # of hams to smoke in overtime each day
- b_f : # of bellies to sell fresh each day
- b_r : # of bellies to smoke in regular time each day
- b_o : # of bellies to smoke in overtime each day
- p_f : # of picnics to sell fresh each day
- p_r : # of picnics to smoke in regular time each day
- p_o : # of picnics to smoke in overtime each day

The the following linear programming instance maximize the daily profit.

$$\max 8h_f + 14h_r + 11h_o + 4b_f + 12b_r + 7b_o + 4p_f + 13p_r + 9p_o.$$
s.t.
$$h_f + h_r + h_o \le 480$$

$$b_f + b_r + b_o \le 400$$

$$p_f + p_r + p_o \le 230$$

$$h_r + b_r + p_r \le 420$$

$$h_o + b_o + p_o \le 250$$

$$h_f, h_r, h_o, b_f, b_r, b_o, p_f, p_r, p_o \ge 0$$

3-2 Problem Create your instance from Problem 3-1 in the JuMP modeling language and solve it, printing the profit and number of each type of meat product to produce.

Answer:

```
_ Julia Code _
         using JuMP, Clp
2
         m = Model(Clp.Optimizer)
3
         @variable(m, hf >= 0)
4
         @variable(m, hr >= 0)
5
         @variable(m, ho >= 0)
6
         @variable(m, bf >= 0)
7
         @variable(m, br >= 0)
8
         @variable(m, bo >= 0)
         @variable(m, pf >= 0)
10
         @variable(m, pr >= 0)
11
         @variable(m, po >= 0)
12
13
         @objective(m, Max, 8*hf + 14*hr + 11*ho + 4*bf + 12*br + 7*bo + 4*pf + 13*pr + 9*po)
14
15
16
         @constraint(m, hf + hr + ho <= 480)
         @constraint(m, bf + br + bo \le 400)
17
         @constraint(m, pf + pr + po <= 230)</pre>
18
         @constraint(m, hr + br + pr <= 420)</pre>
19
         @constraint(m, ho + bo + po \le 250)
20
21
         optimize!(m)
22
23
         println("total profit: ", objective_value(m))
24
         println("Fresh Hams: ", value(hf))
25
         println("Hams Smoked (Regular) ", value(hr))
26
         println("Hams Smoked (Overtime) ", value(ho))
27
         println("Fresh Bellies: ", value(bf))
28
         println("Bellies Smoked (Regular) ", value(br))
29
         println("Bellies Smoked (Overtime) ", value(bo))
30
         println("Fresh Picnics: ", value(pf))
31
         println("Picnics Smoked (Regular) ", value(pr))
32
         println("Picnics Smoked (Overtime) ", value(po))
33
```

```
total profit: 10910.0
Fresh Hams: 440.0
Hams Smoked (Regular) 0.0
Hams Smoked (Overtime) 40.0
Fresh Bellies: 0.0
Bellies Smoked (Regular) 400.0
Bellies Smoked (Overtime) 0.0
Fresh Picnics: 0.0
Picnics Smoked (Regular) 20.0
Picnics Smoked (Overtime) 210.0
```

4 Mission Impossible

Lisa: "I'm going to become a vegetarian"

Homer: "Does that mean you're not going to eat any pork?"

Lisa: "Yes"

Homer: "Bacon?"

Lisa: "Yes Dad"

Homer: "Ham?"

Lisa: "Dad all those meats come from the same animal!"

Homer: "Right Lisa, some wonderful, magical animal!"

Jacob's Meat Alternative Plant (jmap) produces a set of plant-based pork products P using Impossible Meat. (For example, it may be that $P = \{\text{Ham, Bacon, Pork Belly}\}$.) A product $p \in P$ can be prepared (and then sold) using a cooking method $c \in C$. (For example, it may be that $C = \{\text{Fresh, Smoked, Fried}\}$.) jmap produces b_p products of type $p \in P$ per day. The total number of products that can be prepared using cooking method $c \in C$ each day is given by the parameter u_c . The net revenue to jmap of product $p \in P$ when prepared by cooking method $c \in C$ is h_{pc} . Each cooking method $c \in C$ has an associated cost of q_c for any product sold using that method.

4-1 Problem Write a linear program that will maximize the total net profit for jmap. Just write the mathematical model using general notation.

Answer:

Let x_{pc} be the number of pork product $p \in P$ to cook with cooking method $c \in C$ on each day. Then a linear program that will maximize the profit is the following:

$$\max \sum_{p \in P} \sum_{c \in C} (h_{pc} - q_c) x_{pc}$$

s.t.
$$\sum_{c \in C} x_{pc} \leq b_p \quad \forall p \in P$$

$$\sum_{p \in P} x_{pc} \leq u_c \quad \forall c \in C$$

$$x_{pc} \geq 0 \qquad \forall p \in P, c \in C$$

4-2 Problem Implement your instance from Problem 4-1 in JuMP on the data in the data file pork-general.csv. You can either re-type the data, or use Julia DataFrames and CSV to build.

If you wish, you can use this code to help read the data from the CSV file:

```
# You may need to Pkg.add() these
using DataFrames, CSV, NamedArrays
df = CSV.read("pork-general.csv", DataFrame, delim =',')

cook_type = propertynames(df)[3:end]
product = convert(Array, df[3:end,1])
```

```
max_daily_product = Dict(zip(product,df[3:end,2]))
max_daily_cook = Dict(zip(cook_type,df[1,3:end]))
proc_cost = Dict(zip(cook_type,df[2,3:end]))

h = Matrix(df[3:end,3:end])
h_NA = NamedArray(h, (product, cook_type), ("Product", "Cook Type"));
```

Answer:

```
_ Julia Code
     using DataFrames, CSV, NamedArrays
1
     df = CSV.read("pork-general.csv", DataFrame, delim =',')
2
3
     cook_type = propertynames(df)[3:end]
     product = convert(Array, df[3:end,1])
5
6
     max_daily_product = Dict(zip(product,df[3:end,2]))
7
     max_daily_cook = Dict(zip(cook_type,df[1,3:end]))
8
     proc_cost = Dict(zip(cook_type,df[2,3:end]))
10
11
     h = Matrix(df[3:end,3:end])
     h_NA = NamedArray(h, (product, cook_type), ("Product", "Cook Type"));
12
13
14
     using JuMP, HiGHS
15
16
     m = Model(HiGHS.Optimizer)
     set_silent(m)
17
18
     # Number of product p to process using cooking method c
19
     @variable(m, x[product, cook_type] >= 0)
20
21
22
     # Objective
23
     @expression(m, profit, sum( (h_NA[p,c]-proc_cost[c])*x[p,c] for p in product, c in cook_type))
     @objective(m, Max, profit)
24
25
     # Do not exceed the max daily product for p
26
27
     @constraint(m, [p in product], sum(x[p,c] for c in cook_type) <= max_daily_product[p])</pre>
28
     # Do not exceed max daily cook
29
     @constraint(m, [c in cook_type], sum(x[p,c] for p in product) <= max_daily_cook[c])</pre>
30
31
32
     optimize!(m)
     for p in product
33
         for c in cook_type
34
             if value(x[p,c]) > 0.01
35
                  println("Process ", value(x[p,c]), " of product ", p, " using cooking method ", c)
36
37
              end
         end
38
39
     end
```

```
println("Max total profit: ", objective_value(m))
```

```
Process 380.0 of product Ham using cooking method Fresh
Process 100.0 of product Ham using cooking method Cured
Process 230.0 of product PorkBelly using cooking method SmokedRegular
Process 170.0 of product PorkBelly using cooking method Pickled
Process 230.0 of product Picnic using cooking method Pickled
Process 310.0 of product Bacon using cooking method Fresh
Process 190.0 of product Bacon using cooking method SmokedRegular
Process 245.0 of product Tenderloin using cooking method SmokedOvertime
Max total profit: 14725.0
```

5 Optiyours and Optimine

A mining company ("OptiMine") is planning mine operations for the next week. There is a set I of mine locations available for ore extraction. Each mine location $i \in I$ has u_i tons of ore available. Mining at each location $i \in I$ has an associated cost per ton c_i . Finally, there is a set of elements E, and each mine location $i \in I$ has a different percentage of each type of element. The value α_{ei} is the percentage of element $e \in E$ in the ore mined from location $i \in I$. The Optimine company must mine at least K tons of ore total, and each element $e \in E$ must have at least a percentage β_e .

5-1 **Problem** Formulate a linear program to help OptiMine plan mining operations to meet requirements at a minimal cost. Give a general form (parameters only – no numbers) of the math model.

Answer:

Define the decision variables x_i as the tons of ore mined from location $i \in I$. Then we can write a linear program to help OptiMine plan mining operations to meet requriments at a minimal cost as the following:

$$\min \sum_{i \in I} c_i x_i$$
s.t.
$$\sum_{i \in I} x_i \ge K$$

$$\sum_{i \in I} \alpha_{ei} x_i \le \beta_e \sum_{i \in I} x_e \quad \forall e \in E$$

$$0 < x_i < u_i \quad \forall i \in I$$

5-2 **Problem** Here is ChaptGPT (v4.0's) answer for this problem, when I asked to write it in Julia and JuMP:

```
Julia Code

using JuMP
using GLPK # Or any other LP solver
```

```
# Define the data structures
4
         I = # Set of mine locations
5
6
         E = # Set of elements
         u = # Dictionary mapping each location i \in I to the available tons of ore u\_i
7
         c = # Dictionary mapping each location i \in I to the cost per ton c_i
8
9
         \alpha = # Dictionary mapping each (element, location) pair (e, i) to the percentage \alpha_{-}{ei}
         \beta = # Dictionary mapping each element e \in E to the required minimum percentage \beta_e
10
11
         K = # Minimum total tons of ore to be mined
12
         # Create a new model
13
         model = Model(GLPK.Optimizer)
14
15
         # Decision variables
16
         @variable(model, x[i in I] >= 0) # Tons of ore to mine from each location
17
18
         # Objective: Minimize total cost
19
         @objective(model, Min, sum(c[i] * x[i] for i in I))
20
21
         # Constraint: Mine at least K tons of ore in total
22
         @constraint(model, sum(x[i] for i in I) >= K)
23
24
         # Constraints: Ensure the percentage of each element
25
         for e in E
26
              @constraint(model, sum(\alpha[(e, i)] * x[i] for i in I) / sum(x[i] for i in I) >= \beta[e])
27
         end
28
29
         # Solve the model
30
         optimize! (model)
31
32
         # Check results
33
         if termination_status(model) == MOI.OPTIMAL
34
              println("Optimal solution found")
35
              for i in I
36
                  println("Mine ", i, " - Tons of ore: ", value(x[i]))
37
38
              end
         else
39
              println("No optimal solution found")
40
          end
41
```

Is this a correct linear programming model for the problem? If it is not correct, please explain why and fix the model. If it is a correct model, please comment on any differences between your mathematical model and ChatGPT (v4.0)'s Julia/JuMP implementation.

Answer:

This is *not* a linear programming model. The "Ensure the percentage of each element" constraint is *not written in a linear fashion*. Using the fact that we know that $\sum_{i \in I} x_i \ge K > 0$, we can multiply both slides of ChaptGPT's inequality by $\sum_{i \in I} x_i \ge K > 0$ and get an equivalent linear inequality. Its implementation in Jump should look like this:

5-3 Problem Implement and solve this instance of the model in Julia/JuMP. Display the optimal objective value and the optimal variable values.

The cost, ore available, tons of each element required, and percentage of elements in the ore at each mining location are given in the .csv file "mine.csv." A sample of the data is given in the table below. There is a minimum percentage of each element required across the total ore extracted, also given in the .csv file.

The data is provided in the "optimine.csv" on Canvas. You can use the code snippet provided below to read the .csv files and load them into Julia.

```
Julia Code
         #You might need to run "Pkg.add(...)" before using these packages
         using DataFrames, CSV, NamedArrays
2
3
         #Load the CSV data file (should be in same directory as notebook)
         df = CSV.read("optimine.csv",DataFrame,delim=',');
5
6
7
         # create a list of mines
         mines = convert(Array, df[2:end,1])
8
9
         # create a list of elements
10
         # here we take from the DataFrame header (into Julia Symbol)
11
12
         elements = propertynames(df)[2:6]
13
         # create a dictionary of the total cost of mining at each location
14
         c = Dict(zip(mines,df[2:end,7]))
         # create a dictionary of the max tons available at each location
17
         u = Dict(zip(mines,df[2:end,8]))
18
19
         # create a dictionary of the amount required of each element
20
         \beta = Dict(zip(elements,df[1,2:6]))
21
         # create a matrix of the % of each element at each loation
23
         mine_element_matrix = Matrix(df[2:end,2:6])
24
25
26
         # rows are mines, columns are elements
27
         \alpha = NamedArray(mine_element_matrix, (mines, elements), ("mines", "elements"))
28
         K = 3000;
29
30
```

Answer:

Here is my Julia Code

```
_ Julia Code -
          #You might need to run "Pkg.add(...)" before using these packages
         using DataFrames, CSV, NamedArrays
2
3
         #Load the data file
         df = CSV.read("optimine.csv",DataFrame,delim=',');
5
         # create a list of mines
7
         mines = convert(Array, df[2:end,1])
8
9
10
         # create a list of elements,
11
         \#elements = 1:5
12
         elements = propertynames(df)[2:6]
13
         # create a dictionary of the total cost of mining at each location
14
15
         c = Dict(zip(mines,df[2:end,7]))
16
         # create a dictionary of the max tons available at each location
17
         u = Dict(zip(mines,df[2:end,8]))
18
19
         # create a dictionary of the amount required of each element
20
         \beta = Dict(zip(elements, df[1,2:6]))
21
22
         # create a matrix of the % of each element at each loation
23
         mine_element_matrix = Matrix(df[2:end,2:6])
24
25
         # rows are mines, columns are elements, here the names/indices are just integers, but they don't have to be
26
         \alpha = NamedArray(mine_element_matrix, (mines, elements), ("mines", "elements"))
27
28
29
         # We are required to mine K = 3000 tons
         K = 3000
30
31
32
         using JuMP, HiGHS
33
34
         m = Model(HiGHS.Optimizer)
35
         @variable(m, x[mines] >= 0)
36
37
38
         @objective(m, Min, sum(c[i]*x[i] for i in mines)) # minimize costs
39
         # Data was written intranspose format, so get subscripts right here.
40
         \texttt{@constraint(m, meet\_req[e in elements], sum}(\alpha[i,e]*x[i] \text{ for i in mines}) >= \beta[e] * sum(x[i]) \text{ for i in mines}(\alpha[i,e]*x[i])
41
         @constraint(m, max_con[i in mines], x[i] \le u[i]) # don't exceed tons available
42
43
         @constraint(m, dem_con, sum(x[i] for i in mines) >= K) # meet total demand
44
          set_silent(m) # You can turn off HiGHS output if you want
45
         optimize!(m)
46
48
         println("Solution: mine at each location as follows:")
         for i in mines
49
              @printf("%s: %.4f\n", i, value(x[i]))
50
```

```
end

©printf("This has a cost of: %.2f\$", objective_value(m))
```

Which gives the following output:

```
Solution: mine at each location as follows:

1: 254.6667
2: 628.0000
3: 117.0000
4: 547.0000
5: 0.0000
6: 0.0000
7: 0.0000
8: 737.0000
9: 111.0000
10: 605.3333
This has a cost of: 62556.33$
```