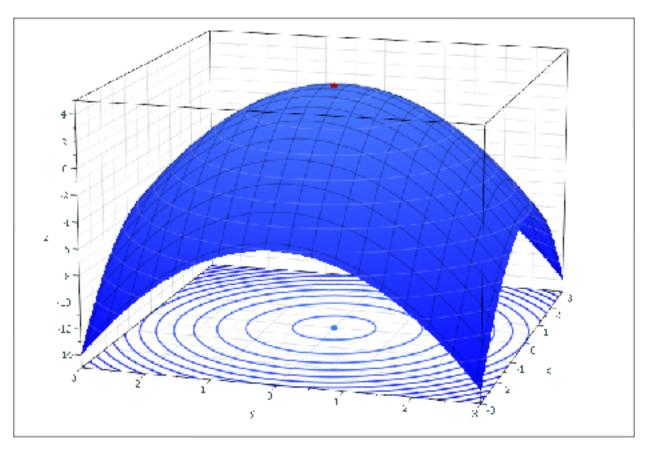
ISYE 4133: Advanced Optimization



Homework 1

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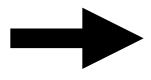
Fall 2019

Problem 1: (10 points) Transforms the following linear programs in the standard form.

(a)

minimize
$$x_1 + 2x_2$$

subject to
 $x_1 + x_2 \ge 10$
 $2x_1 + 5x_2 \le 40$
 $x_1, x_2 \ge 0$



minimize
$$x_1 + 2x_2$$

subject to
 $x_1 + x_2 - s_1 = 10$
 $2x_1 + 5x_2 - s_2 = 40$
 $x_1, x_2, s_1, s_2 \ge 0$

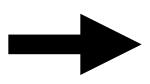
(b) maximize
$$x_1 - x_2$$

subject to
$$x_1 + x_2 \le 4$$

$$2x_1 + 5x_2 = 30$$

$$x_1 \le 0$$

$$x_2 \ge 0$$



minimize
$$x_1 + x_2$$

subject to
 $-x_1 + x_2 + s_1 = 4$
 $-2x_1 + 5x_2 = 30$
 $x_1, x_2, s_1 \ge 0$

Problem 2: (10 points) Implement and solve the linear programs in the above question in the given form as well as the standard form in Gurobi. Note the optimal solutions of the given LP and the LP in its standard form. Show how these solutions are related

```
(a)
Original Code:
from gurobipy import *
    # Create my model
    m = Model('hw1p1a_regular')
    # Create my variables
    x = m.addVars(2, vtype= GRB.CONTINUOUS, name=['x1','x2'])
    # Set an objective function
    m.setObjective(x[0]+2*x[1], GRB.MINIMIZE)
    # Add regular constraints
    m.addConstr(x[0]+x[1] >= 10)
    m.addConstr(2*x[0]+5*x[1] <= 40)
    # Add non-negativity constraints
    m.addConstrs((x[i] >= 0 for i in range(2)))
    # Optimize model
    m.optimize()
    for v in m.getVars():
        print('%s %g' % (v.varName, v.x))
```

```
print('Obj: %g' % m.objVal)
except GurobiError as e:
    print('Error code ' + str(e.errno) + ": " + str(e))
except AttributeError:
    print('Encountered an attribute error')
Original Output:
Optimize a model with 4 rows, 2 columns and 6 nonzeros
Coefficient statistics:
 Matrix range [1e+00, 5e+00]
 Objective range [1e+00, 2e+00]
  Bounds range [0e+00, 0e+00]
                   [1e+01, 4e+01]
  RHS range
Presolve removed 2 rows and 0 columns
Presolve time: 0.00s
Presolved: 2 rows, 2 columns, 4 nonzeros
             Objective |
Iteration
                             Primal Inf.
                                             Dual Inf.
                                                             Time
            0.000000e+00
                            2.500000e+00
       0
                                            0.000000e+00
                                                               0 S
            1.0000000e+01
                            0.00000e+00
                                            0.00000e+00
                                                               0s
Solved in 1 iterations and 0.00 seconds
Optimal objective 1.000000000e+01
x1 10
x2 0
Obi: 10
Standard Code:
from {\sf gurobipy} {\sf import} *
try:
   # Create my model
   m = Model('hw1p1a standard')
```

```
# Create my variables
    x = m.addVars(2, vtype= GRB.CONTINUOUS, name=['x1','x2'])
    s = m.addVars(2, vtype= GRB.CONTINUOUS, name=['s1','s2'])
    # Set an objective function
    m.setObjective(x[0]+2*x[1], GRB.MINIMIZE)
    # Add regular constraints
    m.addConstr(x[0]+x[1]-s[0] == 10)
    m.addConstr(2* x[0]+5*x[1]+s[1] == 40)
    # Add non-negativity constraints
    m.addConstrs((x[i] >= 0 for i in range(2)))
    m.addConstrs((s[i] >= 0 for i in range(2)))
    # Optimize model
    m.optimize()
    for v in m.getVars():
        print('%s %g' % (v.varName, v.x))
    print('Obj: %g' % m.objVal)
except GurobiError as e:
    print('Error code ' + str(e.errno) + ": " + str(e))
except AttributeError:
    print('Encountered an attribute error')
Standard Output:
Optimize a model with 6 rows, 4 columns and 10 nonzeros
Coefficient statistics:
 Matrix range [1e+00, 5e+00]
 Objective range [1e+00, 2e+00]
Bounds range [0e+00, 0e+00]
```

```
RHS range [1e+01, 4e+01]

Presolve removed 4 rows and 2 columns

Presolve time: 0.00s

Presolved: 2 rows, 2 columns, 4 nonzeros
```

```
IterationObjectivePrimal Inf.Dual Inf.Time00.0000000e+002.500000e+000.000000e+000s11.0000000e+010.000000e+000.000000e+000s
```

```
Solved in 1 iterations and 0.00 seconds
Optimal objective 1.000000000e+01
x1 10
x2 0
s1 0
s2 20
Obj: 10
```

Explanation:

Both answers are quite similar, except for the second slack variable being used as a 20 point shift.

(b)

Original Code:

```
from gurobipy import *
```

try:

```
# Create my model
m = Model('hw1a1b_regular')

# Create my variables
```

```
x1 = m.addVar(name="x1",lb = -GRB.INFINITY, ub = 0.0)
x2 = m.addVar(name="x2",lb = 0.0)
```

```
# Set an objective function
m.setObjective(x1-x2, GRB.MAXIMIZE)
```

Add regular constraints

```
m.addConstr(x1+x2 <= 4)</pre>
    m.addConstr(2*x1+5*x2 == 30)
    # Optimize model
    m.optimize()
    print(' worked')
    for v in m.getVars():
        print('%s %g' % (v.varName, v.x))
    print('Obj: %g' % m.objVal)
except GurobiError as e:
    print('Error code ' + str(e.errno) + ": " + str(e))
except AttributeError:
    print('Encountered an attribute error')
Original Output:
Optimize a model with 2 rows, 2 columns and 4 nonzeros
Coefficient statistics:
  Matrix range [1e+00, 5e+00]
 Objective range [1e+00, 1e+00]
 Bounds range [0e+00, 0e+00]
RHS range [4e+00, 3e+01]
Presolve removed 2 rows and 2 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration Objective
                             Primal Inf.
                                             Dual Inf.
                                                              Time
           -1.0666667e+01 0.000000e+00
                                             0.00000e+00
Solved in 0 iterations and 0.00 seconds
```

Optimal objective -1.066666667e+01

worked x1 -3.33333 x2 7.33333 Obj: -10.6667

```
Standard Code:
from {\sf gurobipy} {\sf import} *
try:
   # Create my model
    m = Model('hw1a1b_standard')
    # Create my variables
    x = m.addVars(2, vtype= GRB.CONTINUOUS, name=['x1','x2'], lb
= 0.0)
    s = m.addVars(1, vtype= GRB.CONTINUOUS, name=['s'],lb = 0.0)
    # Set an objective function
    m.setObjective(x[0]+x[1], GRB.MINIMIZE)
    # Add regular constraints
    m.addConstr(-x[0]+x[1]+s[0] == 4)
    m.addConstr(-2*x[0]+5*x[1] == 30)
    # Optimize model
    m.optimize()
    print(' worked')
    for v in m.getVars():
        print('%s %g' % (v.varName, v.x))
    print('Obj: %g' % m.objVal)
except GurobiError as e:
    print('Error code ' + str(e.errno) + ": " + str(e))
except AttributeError:
    print('Encountered an attribute error')
```

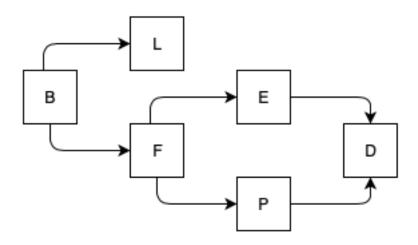
Standard Output:

```
Optimize a model with 2 rows, 3 columns and 5 nonzeros
Coefficient statistics:
 Matrix range
                   [1e+00, 5e+00]
 Objective range
                   [1e+00, 1e+00]
                   [0e+00, 0e+00]
 Bounds range
                   [4e+00, 3e+01]
 RHS range
Presolve removed 2 rows and 3 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration
             Objective
                             Primal Inf.
                                            Dual Inf.
                                                           Time
            1.0666667e+01 0.000000e+00
                                           0.000000e+00
Solved in 0 iterations and 0.00 seconds
Optimal objective 1.066666667e+01
worked
x1 3.33333
x2 7.33333
s[0] 0
Obj: 10.6667
```

Explanation:

The regular and standard form problems have the same absolute objective value output.

Problem 3: (10 points) Formulate a linear program that solves the problem. Explain your formulation. Note, there is no limit on the number of tasks that can be done in parallel. Solve the linear program using Gurobi and report the optimal solution.



Formulation:

$$\begin{aligned} & \text{minimize } S_D + d_D \\ & \text{subject to} \\ & S_D \geq S_E + D_E \\ & S_D \geq S_P + D_P \\ & S_E \geq S_F + D_F \\ & S_P \geq S_F + D_F \\ & S_P \geq S_F + D_B \\ & S_L \geq S_B + D_B \\ & S_L \geq S_B + D_B \end{aligned}$$

$$S_t \geq 0 \ \forall t \in \{B, F, E, P, D, L\}$$
 Where duration $S_t = \{3, 2, 3, 4, 1, 2\} \ \forall t \in \{B, F, E, P, D, L\}$

Formulation: I optimized for the end time of task d (start time of D + duration time of D) given that it is the last task that can be done, requires the completion of E and P, and is the last task that must be completed before the project is considered completed. If the duration data was random (or had some sort of variance), I would have created a more general objective function.

Problem 4: (10 points) For $m \in \{10,20,50,100,500,1000,10000\}$ and $n \in \{10,20,50,100,1000,10000\}$ generate matrices $A \in R^{m \times n}$ whose entries are uniformly random between [0,1]. Similarly generate $b \in R^m$ randomly with entries randomly between [0,1000]. Also generate a cost function $c \in R^n$ with entries randomly between [0, 1000].

(a) Formulate the linear program $\{\min c^T x : Ax \ge b, x \ge 0\}$ for the above random data. Solve 10 instances of the program for each pair of values of m and n with a time out of 2 minutes. Note the time taken and objective value for each run and average over the 10 runs for each pair of (m, n).

Continuous Code:

```
from gurobipy import *
import numpy as np
import csv

with open('problem_4_data.csv', mode='w') as problem_4_data:
    p4_writer = csv.writer(problem_4_data, delimiter=',', quotechar='"', quoting=csv.QUOTE_MINIMAL)
    p4_writer.writerow(["Rows", "Columns", "Avg Obj Val", "Avg RT", "Iterations"])
    #Define Proportions

# m=[10,20,50,100,500,1000,10000]
# n=[10,20,50,100,1000,10000]
"""
    m == 10000 and n == 10000 were not used due to time constraints
```

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m = [10, 20, 50, 100, 500, 1000]

n=[10,20,50,100,1000]

for r in m:

```
for c in n:
            outDict = {}
            for i in range(10):
                i += 1
                # Create A
                A = np.random.rand(r,c)*1000
                # Create b
                b = np.random.randint(0,1000,r)
                # Create cost function
                cost = np.random.randint(0,1000,c)
                # Create dictionary for output
                try:
                    # Create my model
                    m = Model('hw1p4a')
                    m.setParam('TimeLimit',
120, 0) # Sets time limit to 2 minutes
                    m.setParam('OutputFlag', 0) # Mutes output
                    # Create my variables
                    x = m.addVars(c, vtype= GRB.CONTINUOUS)
                    # Create objective function expression
                    expression = quicksum(np.transpose(cos\overline{t})
[j]*x[j] for j in range(c))
                    #Set objective function
                    m.setObjective(expression, GRB.MINIMIZE)
                    # Add regular constraints
                    m.addConstrs(
```

```
(quicksum(A[i,j] * x[j] for j in range(c
                            >= b[i]
                        for i in range(r)))
                    # Optimize model
                    m.optimize()
                    # Append results to output dictionary
                    outDict[i] = (m.objVal, m.runTime)
                except GurobiError as e:
                    print('Error code ' + str(e.errno) + ": " +
str(e))
                except AttributeError:
                    print('Encountered an attribute error')
            average_objVal = float(sum(v[0] for v in outDict.val
ues()))/float(len(outDict))
            average_runTime = float(sum(v[1] for v in outDict.va
lues()))/float(len(outDict))
            print(
                    "Rows: " + str(r) +
                    ", Cols: " + str(c) +
                    ", Avg Obj: " + str(average_objVal) +
                    ", Average RT: " + str(average_runTime) +
                    ", Iterations: " + str(10) + "\n"
            p4_writer.writerow([str(r), str(c), str(average_objV
al), str(average_runTime), str(10)])
```

Continuous Output:

Rows	Columns	RxC	Avg Obj Val	Arg RT	Herations
10	10	100	267.10200758122500	0.0004413604736028130	10
10	20	200	119.86294777630800	0.0006429744720458990	10
10	50	500	73.19295081167900	0.0010250091552734400	10
10	100	1000	37.270193382752400	0.0008771181106587380	10
10	1000	10000	1 3152489686091800	0.002842107309687700	10
20	10	200	480 598296002#1700	0.0003355264363696290	10
20	20	400	89.44367118888790	0.0003612041473388670	10
20	50	1000	93.53874777249660	0.0006245613398144530	10
20	100	2000	30.027006644827700	0.0008089303970336910	10
20	1000	20000	1.5094143547087100	0.003244495391845700	10
50	10	500	544.7916211823890	0 00042042732238769500	10
50	20	1000	285.17163027834700	0.0006036566329656060	10
50	50	2500	84.55353273960980	0.0010224103327612300	10
50	100	5000	63.11174726616620	0.0016416072345459000	10
50	1000	50000	2.306586420037393	0.01066208911896750	10
100	10	1000	477.6998653507580	0.0006019187927246090	10
100	20	2000	283.3250162197110	0.0007596300374438480	10
100	50	5000	125.37113620660000	0.0015504598317553700	10
100	100	10000	62.77988203973680	0.0032430171966552700	10
100	1000	100000	2.0940641683527300	0.032617616353442400	10
500	10	5000	744.6194010006380	0.0021213293375561500	10
500	20	10000	366.24953308778100	0.0031814813313891600	10
500	50	25000	189.06906246645000	0.006858539581298830	10
500	100	50000	88.2265267486503	0.01696019172668460	10
500	1000	500000	2.593984487962500	0.19888641334259000	10
1000	10	10000	£35.0283£05183093	0.003960800170896440	10
1000	20	20000	399.44947552835600	0.007132697105407720	10
1000	50	50000	115.25829962305500	0.016473793983459500	10
1000	100	100000	106.9468178118723	0.03492882251739500	10
1000	1000	1000000	0.5507803607110900	0.48083478332244900	10

(b) Update the formulation to insist the variables are integers. Repeat the experiment. Note the time taken and objective value for each run.

Integer Code:

```
from {\sf gurobipy} {\sf import} *
import numpy as np
import csv
with open('problem_4_data_integer.csv', mode='w') as problem_4_d
ata:
    p4_writer = csv.writer(problem_4_data, delimiter=',', quotec
har='"', quoting=csv.QUOTE_MINIMAL)
    p4 writer.writerow(["Rows", "Columns", "Avg Obj Val", "Avg
T", "Iterations"])
    #Define Proportions
    \# m = [10, 20, 50, 100, 500, 1000, 10000]
    # n=[10,20,50,100,1000,10000]
    m == 10000 and n == 10000 were not used due to time constrai
nts
    m = [10, 20, 50, 100, 500, 1000]
    n=[10,20,50,100,1000]
    for r in m:
        for c in n:
            outDict = {}
            for i in range(10):
                 i += 1
                 # Create A
                 A = np.random.rand(r,c)*1000
```

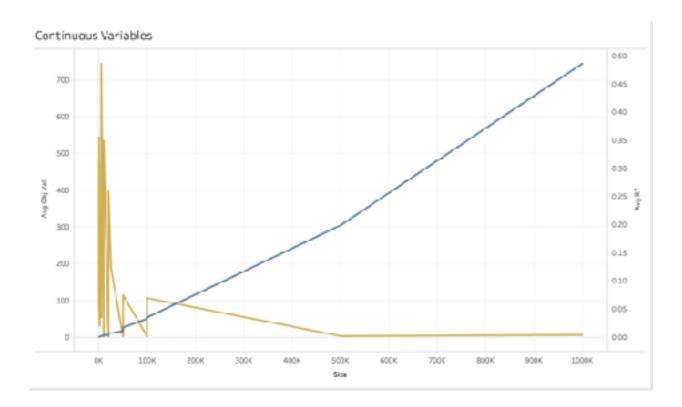
```
# Create b
                b = np.random.randint(0,1000,r)
                # Create cost function
                cost = np.random.randint(0,1000,c)
                # Create dictionary for output
                try:
                    # Create my model
                    m = Model('hw1p4a')
                    m.setParam('TimeLimit',
120, 0) # Sets time limit to 2 minutes
                    m.setParam('OutputFlag', 0) # Mutes output
                    # Create my variables
                    x = m.addVars(c, vtype= GRB.INTEGER)
                    # Create objective function expression
                    expression = quicksum(np.transpose(cost)
[j]*x[j] for j in range(c))
                    #Set objective function
                    m.setObjective(expression, GRB.MINIMIZE)
                    # Add regular constraints
                    m.addConstrs(
                        (quicksum(A[i,j] * x[j] for j in range(c
                            >= b[i]
                        for i in range(r)))
                    # Optimize model
                    m.optimize()
                    # Append results to output dictionary
                    outDict[i] = (m.objVal, m.runTime)
```

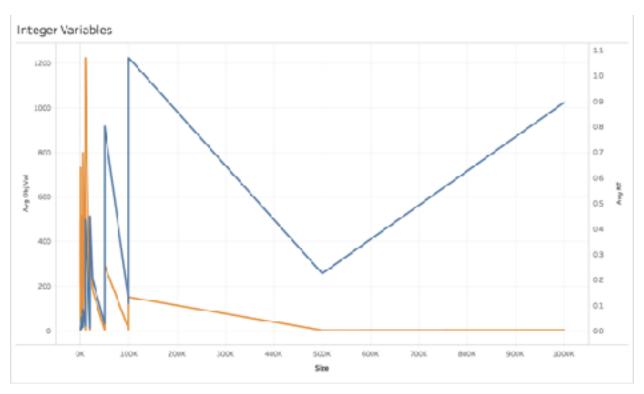
```
except GurobiError as e:
                    print('Error code ' + str(e.errno) + ": "
str(e))
                except AttributeError:
                    print('Encountered an attribute error')
            average_objVal = float(sum(v[0] for v in outDict.val
ues()))/float(len(outDict))
            average_runTime = float(sum(v[1] for v in outDict.va
lues()))/float(len(outDict))
            print(
                    "Rows: " + str(r) +
                    ", Cols: " + str(c) +
                    ", Avg Obj: " + str(average_objVal) +
                    ", Average RT: " + str(average_runTime) +
                    ", Iterations: " + str(10) + "\n"
            p4_writer.writerow([str(r), str(c), str(average_objV
al), str(average_runTime), str(10)])
```

Integer Output:

Rows	Columns	R : C	Avg Obj Val	Aug RT	Iterations
10	10	100	357.1	0.004089641571044920	10
10	20	200	208.3	0.0024032115936279300	10
10	50	500	69.3	0.0033963613821601600	10
10	100	1000	17.5	0.0021613359451293900	10
10	1000	10000	2.6	0.020534396171589800	10
20	10	200	499.3	0.0037323852279663100	10
20	20	400	238.5	0.005616283416748050	10
20	50	1000	84.5	0.005693316459655760	10
20	100	2000	45.5	0.00803515911102295	10
20	1000	20000	0.3	0.013233732967346200	10
50	10	500	732.6	0.01743350028991700	10
50	20	1000	350.8	0.01872096061706540	10
50	50	2500	123.1	0.017633015747070300	10
50	100	5000	35.1	0.014810967445373500	10
50	1000	50000	1.6	0.02877697944641110	10
100	10	1000	410.5	0.016840100288381100	10
100	20	2000	517.9	0.06486728760884260	10
100	90	9000	160.3	0.064823868031237800	10
100	100	10000	67.2	0.06275600360870860	10
100	1030	100000	4.8	0.10746071388683600	10
500	10	9000	799.0	0.0663332936831666	10
500	20	10000	824.9	0.26407437334633900	10
500	e0	25000	184.5	0.20476577281951900	10
500	100	60000	137.0	0.36586837768654700	10
500	1030	F00000	2.3	0.2264390001297000	10
1000	10	10000	1224.5	0.48673376063201900	10
1000	20	20000	508.7	0.4610218381881710	10
1000	эÜ	60000	298.6	0.8023282766342160	10
1000	100	100000	160.5	1.070122981071470	10
1000	1030	1000000	8.8	0.894/25/618/6831	10

(c) Plot the time and objective value as the y-axis and size (m + n) as the x-axis.





Problem 5: (10 points) A furniture manufacturing company makes two models of tables for libraries and other university facilities. Both models use the same table tops but model A has 4 short (18-inch) legs and model B has 4 longer ones (30 inches). It takes 0.10 labor hour to cut and shape a short leg from stock, 0.15 labor hour to do the same for a long leg and 0.50 labor hour to produce a tabletop. An additional 0.30 labor hour is needed to attach the set of leges for either model after all parts are available. Estimated profit is 30 for each model A sold and 45 for each model B sold. Plenty of top material is on hand but the company wants to decide how to use the available 5000 feet of leg stock and 800 labor hours to maximize profit assuming that everything made can be sold.

(a) Formulate a LP to choose the optimal plan. Assume that the number of tables and legs manufactured can take fractional values.

$$\begin{array}{c} \text{maximize } 30Q_A + 45Q_B\\ \text{subject to} \\ 72Q_A + 120Q_B \leq 60{,}000\\ (0.4+0.5+0.3)Q_A + (0.6+0.5+0.3)Q_B \leq 800\\ Q_A,Q_B \geq 0 \end{array}$$

(b) Solve the linear program using Gurobi.

Code:

```
from gurobipy import *

try:

    # Create my model
    m = Model('hw1p5a')

# Create my variables
```

```
q = m.addVars(2, vtype= GRB.CONTINUOUS, name=['qa','qb'])
   # Set an objective function
   m.set0bjective(30*q[0]+45*q[1], GRB.MAXIMIZE)
   # Add regular constraints
   m.addConstr(72*q[0] + 120*q[1] \le 60000)
   m.addConstr(1.2*q[0] + 1.4 * q[1] <= 800)
   # Add non-negativity constraints
   m.addConstrs((q[i] >= 0 for i in range(2)))
   # Optimize model
   m.optimize()
    for v in m.getVars():
        print('%s %g' % (v.varName, v.x))
    print('Obj: %g' % m.objVal)
except GurobiError as e:
   print('Error code ' + str(e.errno) + ": " + str(e))
except AttributeError:
    print('Encountered an attribute error')
```

Output:

```
Optimize a model with 4 rows, 2 columns and 6 nonzeros
Coefficient statistics:
 Matrix range [1e+00, 1e+02]
 Objective range [3e+01, 4e+01]
 Bounds range [0e+00, 0e+00]
 RHS range
                  [8e+02, 6e+04]
Presolve removed 2 rows and 0 columns
Presolve time: 0.00s
Presolved: 2 rows, 2 columns, 4 nonzeros
Iteration
            Objective
                           Primal Inf.
                                           Dual Inf.
                                                          Time
           2.5000000e+04
                           3.659990e+02
                                          0.000000e+00
                                                            0s
      0
           2.3333333e+04
                           0.000000e+00
                                          0.000000e+00
                                                            0s
Solved in 2 iterations and 0.00 seconds
Optimal objective 2.333333333e+04
qa 277.778
qb 333,333
Obj: 23333.3
```

(c) Can you justify the assumption that the variables can take fractional values?

Yes! You can partially build a table. If these constraints were for a calendar month (say, January), then it would be optimal to use as much of our time and stock resources to build as many tables as possible. Any table partially built could be finished in February.