BANA4095: Decision Models - Spring 2021 Linear Optimization, Part 2 (Python)



Dr. Charles R. Sox Associate Dean - Impact & Partnerships Professor of Operations and Business Analytics

Outline

- · Google Colab
- Optimization with Python
- · Google OR Tools
- Python Objects
- · Simple Sidneyville Example
- · Python LP Code Summary

Google Colaboratory (Colab)

- · Cloud environment for Jupyter Notebooks
- Runs on a remote server
- Requires a Google account
- · Save and open files in Google Drive
- · Convenient resource when you can't or don't want to run Jupyter on your local computer
- Access through a web browser recommend Google Chrome

https://colab.research.google.com

Optimization with Python

- Different optimization modeling packages available
- » Glop free open source, Google OR Tools we will use this one!

- » Gurobi commercial, free student/academic license
- » PuLP fee open source, simple, LP and MILP
- » Pyomo free open source, more complex optimization problems
- » Usually includes one or more solver algorithms
- Different optimization solver algorithms available
- » Open Source: CBC, GLPK, Glop, SCIP
- » Commercial: Gurobi, CPLEX, Xpress

Installing Google OR Tools

- · Must have Python and an Internet connection
- Within a Jupyter Notebook, run the command !pip install ortools
- Or within a Command Prompt / Terminal Window
 - » Select Run as Administrator option
 - » python -m pip install --upgrade --user ortools
- · Download may take a few minutes

5

Google OR Tools pywraplp Module

- · GLOP: Google Linear Optimization Package
- Import pywraplp package
 from ortools.linear_solver import pywraplp as glp
- For documentation details use the commands help() and dir()

6

New Python Concept – Object

- An object is a collection of related data and code that is used to manipulate that data
- · For example, a list is actually an object
- The coded functions associated with an object are called methods. We can use the dir() function to find the methods associated with an object. Object methods are invoked using 'dot notation'

```
mylist = ['Clifton', 'Calhoun', 'Ludlow']
dir(mylist)
mylist.append('Corry')
mylist.sort()
```

New Python Concept - Object

- Other objects we've already used: string, integer, float
- Objects can be nested, that is, an object can be composed of other objects with their own associated methods
- In Glop, an optimization model is a compound object that contains other objects:
 - » Variables
 - » Objective
 - » Constraints

8

Glop LP Optimization Objects LP Model Objective Constraint 9

Example - Sidneyville Desk Mfg.

- Allocation/Product Mix Problem
- Produces two types of desk
- Using three types of wood in every desk (measured in board feet, b.f.)

Туре	Profit/desk	
Rolltop	\$115	
Regular	\$90	

	Amount Used		Amount	
Wood	Rolltop	Regular	Available	
Pine	10	20	200	
Cedar	4	16	128	
Maple	15	10	220	

10

Sidneyville Linear Programming (LP) Formulation

```
max 115x_1 + 90x_2 Maximize Total Profit

s.t. 10x_1 + 20x_2 \le 200 Pine

4x_1 + 16x_2 \le 128 Cedar

15x_1 + 10x_2 \le 220 Maple

x_1, x_2 \ge 0 Non-negative
```

```
Glop Optimization Objects

Model Object

mymodel = glp.Solver('Sidneyville',glp.Solver.GLOP_LINEAR_PROGRAMMING)

Variable Objects

Roll_Top = mymodel.NumVar(0, mymodel.infinity(), 'Roll_Top Desks')

Regular = mymodel.NumVar(0, mymodel.infinity(), 'Regular Desks')

Objective Object

Profit = mymodel.Objective()

Profit.SetCoefficient(Roll_Top, 115)

Profit.SetCoefficient(Regular, 90)

Profit.SetMaximization()

Constraint Object

Pine = mymodel.Constraint(-mymodel.infinity(), 200)

Pine.SetCoefficient(Roll_Top, 10)

Pine.SetCoefficient(Regular, 20)
```

3

Creating an Optimization Model

Import pywraplp package

from ortools.linear_solver import pywraplp as glp

Create the optimization model object

```
mymodel = glp.Solver('Sidneyville', glp.Solver.GLOP LINEAR PROGRAMMING)
 model
                                                   model type
```

· Create the decision variables

```
Roll_Top = mymodel.NumVar(0, mymodel.infinity(), 'Roll_Top Desks')
 variable
                          lower
                                      upper
                                                             label
                          bound
```

```
Creating an Optimization Model
```

· Create the objective function

```
Profit = mymodel.Objective()
object
      Profit.SetMaximization() Set optimization direction
      Profit.SetCoefficient(Roll Top, 115)
                                         coefficient
                                 variable
```

Create the constraints

```
bound
constraint Pine = mymodel.Constraint(-mymodel.infinity(), 200)
```

upper

Pine.SetCoefficient(Roll Top, 10)

variable coefficient

Displaying the Model

- Download lptools.py file from Canvas and place a copy in the local file folder folder
- Import the lptools module import lptools as lpt
- Use the lpt.print_model()function to print a mathematical description of your model

Displaying a Solution

Solves the model "mymodel" and returns its solution status

status = mymodel.Solve()

· Other useful methods

Profit.Value()	
Roll_Top.solution_value()	
Pine.dual_value()	

Pine.dual_value()	
<pre>mymodel.ComputeConstraintA</pre>	Activities()

0	Optimal	
1	Feasible	
2	Infeasible	
3	Unbounded	
4	Abnormal	
5	Not Solved	

- Display the solution
 - » Use print() function with formatting commands to display the optimal value, decision variables, and dual/shadow prices

Python Optimization Code Steps

- Import Glop package
- Input parameters
- · Create model object
- · Create variable objects
- Create objective function and coefficients
- Create constraints and coefficients

- · Solve the problem
- Check solution status
- · Display optimal value
- Display optimal solution
- · Display constraint values

- 1

Important Glop Object Methods

Model Object

```
.NumVar(), .Objective(), .Constraint(), .variables(), .constraints(), .ComputeConstraintActivities(), .Clear()
```

Objective Object

```
.SetMaximization(), .SetMinimization(),
.SetCoefficient(), .Value()
```

Variable Objects

```
.name(), .lb(), .ub(), .solution_value(), .reduced_cost()
```

· Constraint Objects

.SetCoefficient(), .name(), .lb(), .ub(), .dual_value()

18

Simple Sidneyville LP Model

- Requires excessive lines of code
- · Time consuming to construct and maintain
- Should use Python data structures and loops to generalize the code and make it more efficient to construct and maintain

Summary

- Google Colab
- · Optimization with Python
- Google OR Tools
- Python Objects
- Simple Sidneyville Example
- Python LP Code Summary

20