



Introduction to Optimization Modeling in Python

PSE Summer School
Universitat Politècnica de Catalunya



Contents

- ▶ **Install PYOMO and solvers**
- ▶ **Introduction to Python**
- ▶ **PYOMO Components**
- ▶ **Case Studies**

<https://github.com/CAChemE/>

Install PYOMO and Solvers

PYOMO

- conda install -c conda-forge pyomo
- conda install -c conda-forge pyomo.extras
<http://www.pyomo.org/installation/>

SOLVERS

- glpk [LP, MILP] >> conda install -c conda-forge glpk
<http://ftp.gnu.org/gnu/glpk/>
- gurobi [LP, MILP] >> download and install. Free university license
<http://www.gurobi.com/>
- IPOPT [NLP] >> conda install -c conda-forge ipopt
<https://www.coin-or.org/download/binary/>
- SCIP [MINLP] >> download and add the solver installation to the path environment variable
<http://scip.zib.de/#download>

PYOMO Sources

Homepage:



<http://www.pyomo.org/>



<https://software.sandia.gov/trac/pyomo>

Book Reference

Springer Optimization and Its Applications 67

William E. Hart
Carl D. Laird
Jean-Paul Watson
David L. Woodruff
Gabriel A. Hackebeil
Bethany L. Nicholson
John D. Siirola

Pyomo —
Optimization
Modeling
in Python

Second Edition

Springer

PYOMO Sources

GitHub

<https://github.com/Pyomo>

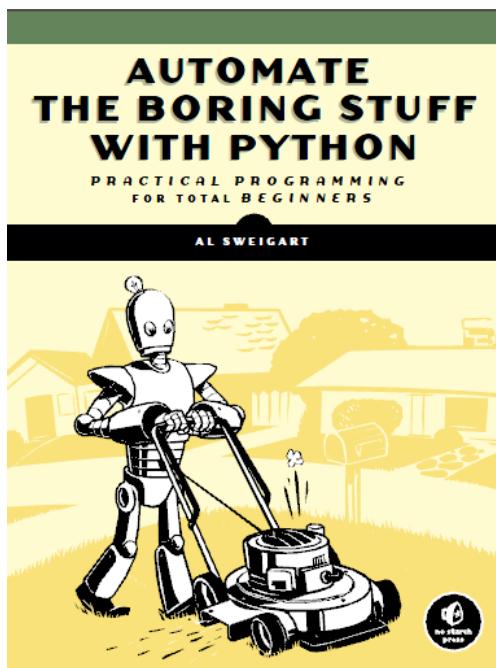
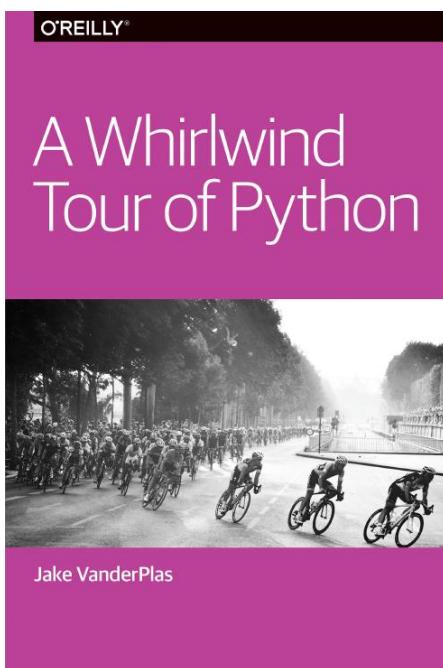
Help Forums

<https://groups.google.com/forum/#!forum/pyomo-forum>

Stack Overflow

<https://stackoverflow.com/questions/tagged/pyomo>

PYTHON Sources



pycse - Python3 Computations in Science and Engineering

John Kitchin
jkitchin@andrew.cmu.edu
<http://kitchingroup.cheme.cmu.edu>
Twitter: @johnkitchin

2015-04-25

python pycse - Python3 computations in science and engineering

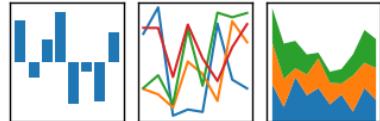
A Python logo icon followed by five small square plots showing various scientific and engineering data visualizations like heatmaps and line graphs.

PYTHON Sources



pandas

$$y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$$



matplotlib

SQLAlchemy



PYOMO Components

PYOMO Components

Example: Machinery Problem

A company manufacture four types of machinery. The factory is divided in three sections. The first section has available 960 h/week, the second 1110 h/week and the third 400 h/week. Each machinery unit requires the following time at each section

Plants	hours per machinery			Profit [units/machinery]
	Machining	Painting	Assembly	
Machinery 1	6	3	2	12
Machinery 2	4	3	1	8
Machinery 3	4	6	2	12
Machinery 4	8	9	1	17

Determine the number of units of machinery for each type that should be manufacture per week to **maximize the profit**.

PYOMO Components

Example: Machinery Problem

Nomenclature

- m → machinery type (set)
s → factory section (set)
 profit_m → profit per machinery type (parameter)
 b_s → time availability in each section per week (parameter)
 $T_{m,s}$ → time required for each machinery type in each section (parameter)
 x_m → number of units of machinery for each type (variable)

$$\begin{aligned} \max_x : & \sum_m \text{profit}_m x_m && \text{Objective function} \\ s.t. & \sum_m T_{m,s} x_m \leq b_i & \forall s & \text{Factory section time limit} \\ & x_m \in \mathbb{Z} \end{aligned}$$

Model Structure

```
from pyomo.environ import *

m = ConcreteModel()

M = m.M = Set(initialize = ['m1', 'm2', 'm3', 'm4'])
S = m.S = Set(initialize = ['s1', 's2', 's3'])

profit = {'m1':12, 'm2':8, 'm3':12, 'm4':17}
max_time = {'s1': 960, 's2': 1110, 's3': 400}
time_x_section = {
    ('m1','s1'): 6 , ('m1','s2'): 3 , ('m1','s3'): 2 ,
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x = m.x = Var( M, within = PositiveIntegers)

m.value = Objective(
    expr = sum( profit[i] * m.x[i] for i in M),
    sense = maximize )

def constraint_rule(m, j):
    return sum(time_x_section[i,j] * x[i] for i in M)
        <= max_time[j]
m.constraint = Constraint(S, rule = constraint_rule)

opt = SolverFactory('glpk').solve(m)
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Specify/import problem data

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Constraint functions declaration

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```

Solver call

Case Studies

Assignment Problem

In this problem, we have a number of people “p” and a number of tasks “t”. Each person has a suitability coefficient “SC”, which represents how effectively can a person “p” perform a task “t”. The objective is to **maximize the total suitability of the system**. For this example, the following data is presented:

PEOPLE: Pedro, Marta, Laura

TASKS: Accountant, Sell Manager, Human Resources

SUITABILITY COEFFICIENTS:

Person	SC Accountant	SC Sell Manager	SC Human Resources
Pedro	11	5	2
Marta	15	12	8
Laura	3	1	10

Assignment Problem

Mathematical Model

$$\max \left(\sum_{p,t} C_{p,t} y_{p,t} \right)$$

Maximize the suitability of the assignment

s.t.

$$\sum_t y_{p,t} = 1 \quad \forall p$$

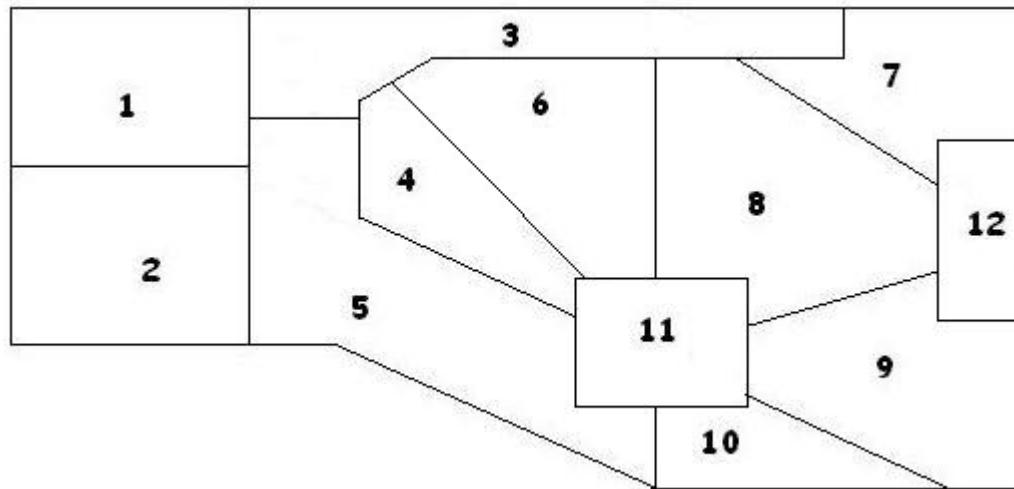
Each person can only perform one job

$$\sum_p y_{p,t} = 1 \quad \forall t$$

Each job must be performed by one and only one person

Set Covering Problem

In this problem, we have a number of zones in which we may or may not install awesome new firefighter stations. However, the mayor of the city is a bit stingy, and wants us to install the absolute minimum number of stations possible. Having the following map of the zone:



And considering that a single station can only provide service to the zones in its immediate neighborhood, **what stations should be built?**

Set Covering Problem

Mathematical Model

$$\min\left(\sum_i y_i\right)$$

Minimize the number of stations

s.t.

$$\sum_i C_{i',i} y_i \geq 1 \quad \forall i' \quad \text{Service constraint}$$

Knap-Sack Problem

In this problem, we are adventurous thieves. We want to loot all of the treasures that we can before the guards arrive. Since it will not be possible to come back to loot whatever we leave behind, we must ensure that we maximize the benefit of what we steal. Our horse can handle up until 2500 g of weight (It's a tiny pony) and a volume of 2000 cm³. Considering the loot table, what should we carry out there to sell?

Item	Market Price	Volume (cm ³)	Unit Weight (g)	Units available
Chest	50	1000	2000	1
Ring	5	2	20	10
Necklace	3	10	300	1
Mirror	20	500	1000	1
Bracelet	16	15	300	15
Ruby	5	3	75	1
Parfum	1	100	100	1
Diamond	30	5	50	1
Gold goblet	12	250	500	1
Spice	40	100	100	1

Knap-Sack Problem

Mathematical Model

$$\max\left(\sum_i MP_i n_i\right) \quad \text{Maximize Profit}$$

s.t.

$$\sum_i V_i n_i \leq 2000 \quad \text{Volume constraint}$$

$$\sum_i W_i n_i \leq 2500 \quad \text{Weight constraint}$$

$$n_i \leq N_i \quad \forall i \quad \text{Amount constraint}$$

Sudoku problem

Nomenclature

$r \rightarrow$ rows | $c \rightarrow$ columns | $k \rightarrow$ value

$y_{r,c,k} \rightarrow$ binary variable. $y_{r,c,k} = 1$ means cell [r, c] is assigned number k

Every position in the Sudoku is filled

$$\sum_k y_{r,c,k} = 1 \quad \forall r, c$$

Cells in the same column must be assigned distinct numbers

$$\sum_r y_{r,c,k} = 1 \quad \forall c, k$$

Cells in the same row must be assigned distinct numbers

$$\sum_c y_{r,c,k} = 1 \quad \forall r, k$$

	c1	c2	c3	c4	c5	c6	c7	c8	c9
r1	5	3			7				
r2	6			1	9	5			
r3		9	8					6	
r4	8				6				3
r5	4			8					1
r6	7				2				6
r7		6					2	8	
r8				4	1				5
r9					8			7	9

Cells in the same 3x3 grid must be assigned distinct numbers

$$\sum_{r=3p-2}^{3p} \sum_{c=3q-2}^{3q} y_{r,c,k} = 1 \quad \forall k, p, q = 1, 2, 3$$

Strip packing 2D problem

Nomenclature

$i \rightarrow$ rectangles, $i = \{1, 2, \dots, n\}$

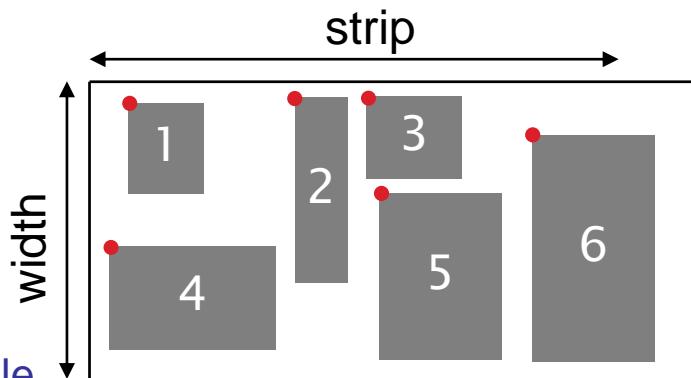
$l_t \rightarrow$ length of the strip

$(x_i, y_i) \rightarrow$ rectangle coordinates

$L_i, H_i \rightarrow$ Length and height of rectangle i

$W \rightarrow$ Width of the strip

$UB_i \rightarrow$ Upper bound for the x-coordinate of every rectangle



GDP Formulation

Source: Sawaya & Grossmann (2005),
<https://doi.org/10.1016/j.compchemeng.2005.04.004>

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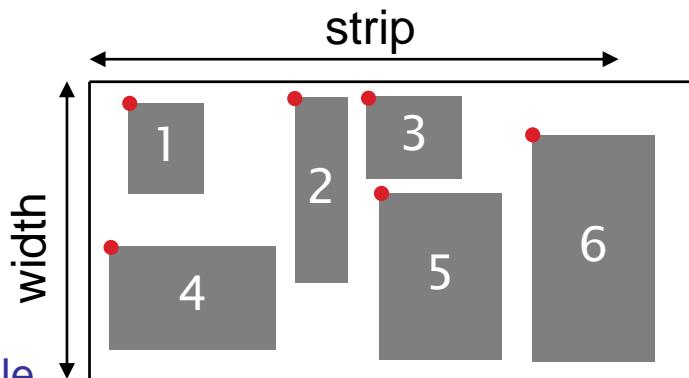
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GDP Formulation

$$\min l_t$$

$$s.t \quad l_t \geq x_i + L_i \quad \forall i \in N$$

$$Y_{i,j}^1 \vee Y_{i,j}^2 \vee Y_{i,j}^3 \vee Y_{i,j}^4 \quad \forall i, j \in N, i < j$$

$$x_i \leq UB - L_i \quad \forall i \in N$$

$$H_i \leq y_i \leq W \quad \forall i \in N$$

$$l_t, x_i, y_i \in \mathbb{R}$$

$$Y_{i,j} \in \{True, False\}$$

Source: Sawaya & Grossmann (2005),
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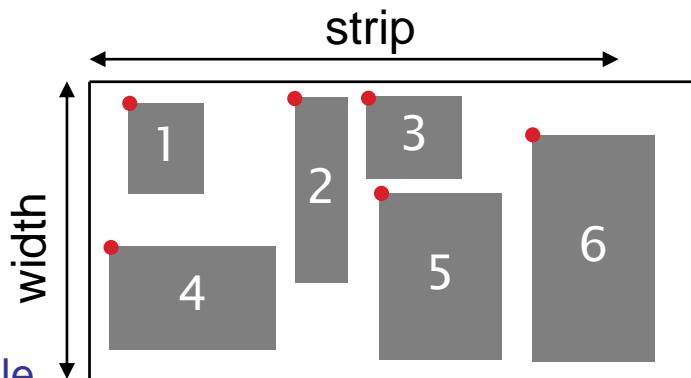
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MILP Formulation [Big-M]

$$\begin{aligned}
 & \min \quad l_t \\
 \text{s.t} \quad & l_t \geq x_i + L_i \quad \forall i \in N \\
 & y_j - H_j \geq y_i - M_{ij}^1 \quad 1 - w_{ij}^1 \quad \forall i, j \in N, i < j \\
 & x_i + L_i \leq x_j + M_{ij}^2 \quad 1 - w_{ij}^2 \quad \forall i, j \in N, i < j \\
 & y_i - H_i \geq y_j - M_{ij}^3 \quad 1 - w_{ij}^3 \quad \forall i, j \in N, i < j \\
 & x_j + L_j \leq x_i + M_{ij}^4 \quad 1 - w_{ij}^4 \quad \forall i, j \in N, i < j \\
 & \sum_{d \in D} w_{ij}^d = 1 \quad \forall i, j \in N, i < j \\
 & x_i \leq UB - L_i \quad \forall i \in N \\
 & H_i \leq y_i \leq W \quad \forall i \in N \\
 & l_t, x_i, y_i \in \mathbb{R}, \quad w_{i,j} \in [0, 1]
 \end{aligned}$$



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