

GUROBI
OPTIMIZATION

Python II: Advanced Algebraic Modeling with Python and Gurobi

- This is the second in our series on optimization modeling in Python
 - You *can* build models in other popular programming languages
 - This series focuses on special features for Python that help for optimization models
- Since this is an “advanced” presentation, we assume you are familiar with:
 - Python
 - Optimization modeling
- If this is new to you, check the Resources section on www.gurobi.com
 - View the recorded presentation “Python I: Introduction to modeling with Python”

Agenda

- Get started using a simple example of workforce scheduling
- Understand general principles of optimization modeling
- Learn powerful modeling features in Python and the Gurobi interface

Getting started

Workforce scheduling example

Optimization model principles: Decisions

- Decisions: What to do
 - Optimization variables
- Examples
 - Workforce: do you assign a worker to a shift
 - Others
 - Select
 - Assign
 - Make
 - Go
 - Buy / Sell
- Types of decisions
 - Yes / No
 - Quantities – how much to do

Optimization model principles: Quantities

- Quantities: Numerical values used in relationships
 - Coefficients in expressions – linear, quadratic, logical, etc.
- Examples
 - Workforce: cost for a worker in a shift
 - Others
 - Cost, price
 - Distance, space
 - Time
 - Materials required or used
 - Supply, demand
 - Capacity

Optimization model principles: Constraints

- Constraints: Requirements or limits
 - Equality or inequality relationships
- Examples
 - Workforce: assign enough workers to meet the requirements for each shift
 - Others
 - Demand
 - Equipment
 - Supplies
 - Space
 - Workers
 - Money
 - Time

Optimization model principles: Objective



- Objective: Goal to achieve
 - Mathematical function to optimize
- Examples
 - Workforce
 - Initial: Minimize labor cost
 - Later: Minimize shortfall in shifts
 - Others
 - Maximize: profit, utilization, balance
 - Minimize: costs, time, waste

Optimization model principles: Indexes

- Index: Specifies a key value
 - Identifies an element of an array
- Examples
 - Workforce
 - Workers
 - Shifts
 - Availability: a list that specifies a shift when a worker is available
 - Others
 - Product models
 - Workers
 - Equipment – machines, trucks, ...
 - Locations
 - Customers or suppliers
 - Supplies
 - Time – hour, day, week, month, quarter, year

Indexes are key to optimization models

- Describe individual variables, numerical quantities, constraints
- Use of indexes
 - Iterate over variables or constraints – for-all statements
 - Compute sums over variables and numerical quantities
- An efficient model depends on proper handling of indexes
 - For an efficient model, setup should require much less time and memory than solving

- A list is a built-in data type that
 - Can store items – int, float, string, tuples (pairs, ...), other objects, ...
 - Is ordered
 - Efficiently supports iteration and random access
 - Can be modified – append, delete, join, ...
- Lists are a good way to create optimization indexes
 - Note: Unlike a mathematical set, lists can contain repeated elements

- Set-builder notation for creating a list

- From another iterator

```
Squares = [i**2 for i in range(6)] # [0,1,4,9,16,25]
```

- From multiple iterators

```
Pairs = [(i,j) for j in range(4) for i in range(j)]  
# [(0,1), (0,2), (1,2), (0,3), (1,3), (2,3)]
```

- With a filter (if condition)

```
Nonsquares = [i for i in range(10) if i not in Squares]  
# [2,3,5,6,7,8]
```

- List comprehension syntax is similar to optimization notation

- You can use for-loops, but they are typically more wordy

```
Pairs = []  
for j in range(4):  
    for i in range(j):  
        Pairs.append((i,j))
```

Python generator expressions

- Similar syntax as list comprehension, without the intermediate list

- List comprehension

```
Squares = [i**2 for i in range(6)] # [0,1,4,9,16,25]
```

- Generator expression

```
SumSquares = sum(i**2 for i in range(6)) # 55
```

- The Gurobi interface takes advantage of lists, list comprehension and generator expressions

- More details in a moment

- A Gurobi class to store lists of tuples

```
Cities = ['A', 'B', 'C', 'D']
```

```
Routes = tuplelist([('A', 'B'), ('A', 'C'), ('B', 'C'), ('B', 'D'), ('C', 'D')])
```

- What makes it special: select statement for efficient filtering

```
for c in Cities:                                # [('A', 'B'), ('A', 'C')]
    print(Routes.select(c, '*'))                # [('B', 'C'), ('B', 'D')]
                                                # [('C', 'D')]
                                                # []
```

- The tuplelist is indexed to make `select()` efficient

- A Gurobi dictionary class, where the keys are tuples
 - The list of keys is a tuplelist, so it provides `select()`, `sum()` and `prod()` operators
- Used for decision variables and constraints
 - Examples in a moment

Indexed variables: `Model.addVars()`

- Using integers

```
x = m.addVars(2, 3, name="x")  
# x[0,0], x[0,1], x[0,2], x[1,0], x[1,1], x[1,2]
```

- Using lists of scalars

```
y = m.addVars(Cities, Cities, name="y")  
# y[A,A], y[A,B], y[A,C], y[A,D], y[B,A], y[B,B], y[B,C], y[B,D]  
# y[C,A], y[C,B], y[C,C], y[C,D], y[D,A], y[D,B], y[D,C], y[D,D]
```

- Using a tuplelist

```
z = m.addVars(Routes, name="z")  
# z[A,B], z[A,C], z[B,C], z[B,D], z[C,D]
```

- Using a generator expression (!)

```
w = m.addVars((i for i in range(5) if i != 2), name="w")  
# w[0], w[1], w[3], w[4]
```

More about Model.addVars()

- Automatically takes the cross-product of multiple indexes

```
x = m.addVars(2, 3, name="x")  
y = m.addVars(Cities, Cities, name="y")
```

- Generates names automatically, using the subscripts
- Requires lists of scalars or lists of tuples

Indexed constraints: Model.addConstrs()

- Uses a generator expression

```
x = m.addVars(Routes, name="x")  
y = m.addVars(Routes, name="y")  
m.addConstrs((x[i,j]+y[i,j] <= 2 for i,j in Routes), name="capacity")
```

- Linear and quadratic expressions are used in constraints and the objective
 - Basic (binary) mathematical operators (+ − × ÷)
 - Aggregate sum operator (\sum)
 - Used alone as well as in products (dot product)
- Indexes are key to the aggregate sum operator
 - Gurobi Python interface has 2 versions: full and simple
 - Why two?
 - The simple syntax is easier
 - There are some things you cannot do with the simple syntax

Aggregate sum: Full

- Use generator expression inside a quicksum function

```
obj = quicksum(cost[i,j]*x[i,j] for i,j in arcs)
```

- quicksum works just like Python's sum function, but it is more efficient for optimization models

Aggregate sum: Simple

- A tupledict of variables has a `sum()` function, using the same syntax as `tuplelist.select()`:

```
x = m.addVars(3, 4, vtype=GRB.BINARY, name="x")  
m.addConstrs(x.sum(i, '*') <= 1 for i in range(3))
```

- This generates the constraints:

```
x[0,0] + x[0,1] + x[0,2] + x[0,3] <= 1  
x[1,0] + x[1,1] + x[1,2] + x[1,3] <= 1  
x[2,0] + x[2,1] + x[2,2] + x[2,3] <= 1
```

Dot product

- A tupledict of variables has a `prod()` function to compute the dot product

- If `cost` is a dictionary, then the following are equivalent:

```
obj = quicksum(cost[i,j]*x[i,j] for i,j in arcs)
```

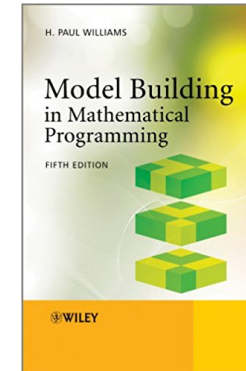
```
obj = x.prod(cost)
```


Putting it all together

Network flow example

Resources to learn more

- Code Examples - <http://www.gurobi.com/resources/examples/example-models-overview>
 - Functional (code) examples
 - Modeling examples
- Book:
 - Model Building in Mathematical Programming by HP Williams
- Gurobi Python documentation - <http://www.gurobi.com/documentation/>
- Python – www.python.org



Thank you for joining us



- If you haven't already done so, please register at www.gurobi.com, and then visit <http://www.gurobi.com/get-anaconda> to try Gurobi and Python for yourself.
- The next Python webinar will focus on optimization and heuristics and take place on April 19th and again on April 20th.
 - You can learn more at <http://www.gurobi.com/company/events/webinars-python>.
- For questions about pricing, please contact sales@gurobi.com or sales@gurobi.de.
- A recording of the webinar, including the slides, will be available in roughly one week.