Superpowered Optimization in Python With Gurobi and Anaconda

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- Optimization Support Engineer at Gurobi Optimization
- Ph.D. in Industrial and Systems Engineering, Georgia Tech
- Expert in optimization modeling and software development
- Over a decade of experience implementing decision support systems







Michael C. Grant, Ph.D.

- Senior Solution Architect at Continuum Analytics
- Ph.D. in Electrical Engineering, Stanford University
- Subject matter expert in nonlinear convex optimization
- Developer of the CVX modeling framework





Overview

- The Proposition
- Introducing Gurobi Optimization
- About Continuum Analytics & Anaconda
- The Gurobi Python module
- A simple MIP model using Gurobi & Spyder
- Portfolio optimization using Gurobi, Jupyter, Pandas, & Bokeh
- Wrap up / Call to action







- The Gurobi Python module provides the intuitive feel of a modeling language inside a powerful programming language
- The Anaconda platform delivers a full application development stack as well as a great interactive environment for data analysis and modeling
- Bringing developers and analysts/modelers together using Python means collaboration has never been easier

The Gurobi Optimizer











- Algorithms for continuous optimization (LP, QP, QCP)
 - Simplex, Barrier
- Algorithms for discrete optimization (MIP, MIQP, MIQCP)
 - Parallel branch-and-bound
- Other algorithms
 - Presolve, parameter tuning, irreducible inconsistent subsystems...
- Command-line interface, full-featured interactive shell
- Programming interfaces
 - C, C++, Java, .NET, Python, MATLAB, R



Gurobi Model Components



- Decision variables
- Objective function
 - minimize $\mathbf{x}^{\mathsf{T}}\mathbf{Q}\mathbf{x} + \mathbf{c}^{\mathsf{T}}\mathbf{x} + \alpha$
- Constraints

$$- Ax = b$$

- $1 \le x \le u$

- some x_h integral

- some x_i lie within second order cones

 $- \mathbf{x}^{\mathsf{T}} \mathbf{Q}_{j} \mathbf{x} + \mathbf{q}_{j}^{\mathsf{T}} \mathbf{x} \leq \beta_{j}$

- some x_k in SOS

(linear constraints)

(bound constraints)

(integrality constraints)

(cone constraints)

(quadratic constraints)

(special ordered set constraints)



Example: Mixed Integer Program



- Decision variables
- Objective function
 - minimize $\mathbf{x}^{\mathsf{T}}\mathbf{Q}\mathbf{x} + \mathbf{c}^{\mathsf{T}}\mathbf{x} + \alpha$
- Constraints

$$-Ax = b$$

$$1 \le x \le u$$

- some x_h integral
- some x_i lie within second order cones

$$-\mathbf{x}^{\mathsf{T}}\mathbf{Q}_{i}\mathbf{x}+\mathbf{q}_{i}^{\mathsf{T}}\mathbf{x}\leq\beta_{i}$$

- some x_{ν} in SOS

(linear constraints)

(bound constraints)

(integrality constraints)

(cone constraints)

(quadratic constraints)

(special ordered set constraints)



Benchmarks – Open Source MIP



- Mittelmann MIPLIB2010 tests, P=1 (>1X means Gurobi wins):
 - http://plato.la.asu.edu/bench.html

	January 2012	January 2013	January 2015	November 2015	% Solved
CBC	10X	13X	17X	26X	55%
SCIP	6X	7X	7X	9X	80%
GLPK	22X	27X	-	-	1%
LPSOLVE	19X	24X	-	-	6%

GLPK and LPSOLVE are not currently tested in the public benchmarks







- Performance leader
 - Average 2X improvement with each major release
- Consistent track record of innovation
 - First cloud offering, client-server support, distributed algorithms, ...
- Outstanding support
 - Direct access to Ph.D.-level optimization experts
- No surprises
 - Flexible licensing and transparent pricing
- Strong academic offering
 - Free full-featured academic licenses, Take Gurobi with You program

About Continuum Analytics & Anaconda









Continuum Analytics

- Founded in 2012 by Travis Oliphant and Peter Wang
- Key contributors to the Python open source ecosystem
 - Travis: primary developer, NumPy; founding contributor, SciPy
 - PyTables, Pandas, Jupyter/IPython, Matplotib
 - New projects: Blaze, Numba, Conda
- Commercial efforts:
 - Enterprise software & support
 - Training: Python for Science, Python for Finance
 - Consulting: custom software development





Anaconda for Data Science

Empowering Everyone on the Team

Data Scientist

- Advanced analytics with Python & R
- Simplified library management
- Easily share data science notebooks & packages

Developer

- Support for common APIs & data formats
- · Common language with data scientists
- Python extensibility with C, C++, etc.

Ops

- · Validated source of up-to-date packages including indemnification
- Agile Enterprise Package Management
- · Supported across platforms

Data Engineer

- Powerful & efficient libraries for data transformations
- Robust processing for noisy dirty data
- Support for common APIs & data formats

Business Analyst

- Collaborative interactive analytics with notebooks
- Rich browser based visualizations
- Powerful MS Excel integration

Computational Scientist

- Rich set of advanced analytics
- Trusted & production ready libraries for numerics
- Simplified scale up & scale out on clusters & GPUs





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Why Python?

GUROBI OPTIMIZATION

- Designed for readability
- Functional, object-oriented, declarative
- Great glue language for legacy C/Fortran
- Rich standard library
- Incredible community library
- Great online resources
- Great name

"The second-best language for everything"

— Peter Wang, CTO



The Anaconda Python Distribution



- Windows, Mac, Linux; 32/64-bit
- Python 2.x, 3.x your choice (though do pick 3.x!)
- ~150 packages in the default installation
- conda package and environment management
 - ~350 packages available with a simple conda install
 - Create isolated "silos" for multiple projects
- Like R? Anaconda has that too







conda package management

nstall pandas numpy=1.9 gurobi

Update conda update gurobi

List conda list

Search conda search gurobi

Remove conda remove matplotlib







Create

conda create -n lp project anaconda gurobi

Select

source activate lp_project (Linux/Mac)
activate lp_project (Windows)

Return to default

source deactivate (Linux/Mac) deactivate (Windows)

Remove

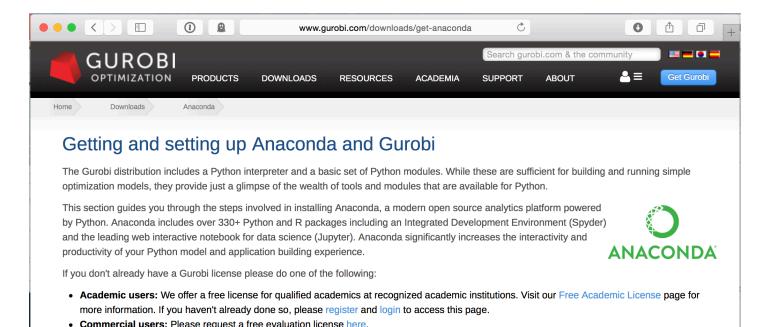
conda remove -n lp_project --all



Obtaining Anaconda

From Continuum: https://www.continuum.io/downloads

From Gurobi: http://www.gurobi.com/downloads/get-anaconda



The Gurobi Python Module









- Building Models with Gurobi
- Option 1: use a modeling language
 - Very easy to build models
 - Optimization modeling constructs built into language
 - Attractive choice for non-programmers
- Option 2: use a full programming language
 - Much more powerful and flexible development environment
 - Complete access to solver functionality
 - Richer set of language features
 - Natural choice when deploying and/or integrating with applications



Gurobi Python Environment



- High-level optimization modeling constructs embedded in Python
- Design goals:
 - Require minimal programming skills to get started
 - Bring "feel" of a modeling language to the Python interface
 - Allow for code that is easy to write and maintain
 - Maintain unified design across all of our interfaces
 - Remain lightweight and efficient compared to solver alone
 - Support all solver and programming needs



Essential Gurobi Python Constructs



- Objects represent model components (Model, Var, Constr)
- Overloaded operators
 - Arithmetic (+, -, ×, \div), constraints (≤, =, ≥)
- Aggregate sum operator (quicksum)
- Python provides the rest for representing data, indices and subscripts
 - Lists, tuples, dictionaries, loops, generator expressions, ...

Ex:
$$x_i + y_i \le 5, \forall i \in I \iff \text{for i in I:}$$

m.addConstr(x[i] + y[i] <= 5)



Building a Gurobi Model: Six Steps



m = Model()

Add variables

x = m.addVar(...)

Commit changes

m.update()

Set objective

m.setObjective(...)

Add constraints

m.addConstr(...)

Optimize

m.optimize()



Installing the Gurobi Python Module



Add the channel

conda config --add channels gurobi

Install Gurobi

conda install gurobi

Retrieve license

Test license

>>> import gurobipy

Test token server

gurobi_cl --tokens

A Simple MIP Model using the Spyder IDE







Spyder

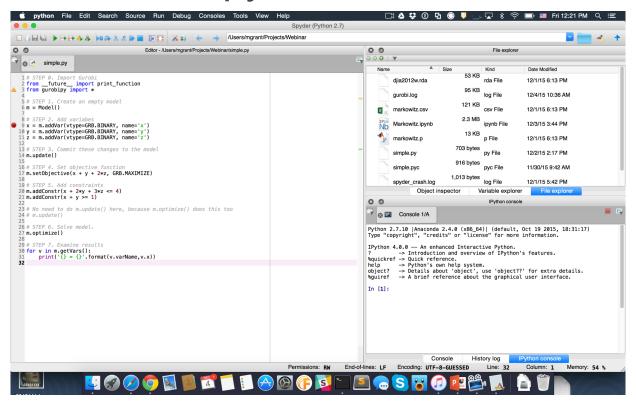
- Scientific Python Development EnviRonment
- Familiar integrated development environment (IDE) for users of MATLAB, Visual Studio, etc.
- Built-in editor, file explorer, documentation browser
- Graphical interface for debugging (ipdb)
- IPython console for interactive code execution







Spyder Demo



Portfolio Optimization in Jupyter







Modern Portfolio Theory In One Slide GURO

- First offered by Harry Markowitz in 1952
- Recognized by a Nobel Memorial Prize in Economics
- Associates with each stock an expected return and a risk or volatility measure, and a matrix of correlations between pairs of stocks
- The goal is to study the tradeoff between risk and return
 - Minimize risk for a given target (expected) return
 - Maximize return given an upper bound on risk
- The most basic models are quadratic programs (QPs)



(okay, two slides)



minimize
$$x^T \Sigma x$$
 minimize $\sum_{i=1}^n \sum_{j=1}^n \sigma_{ij} x_i x_j$
subject to $\vec{1}^T x = 1$ subject to $\sum_{i=1}^n x_i = 1$
 $\rho^T x = \gamma$ $\sum_{i=1}^n \rho_i x_i = \gamma$
 $x \ge 0$ $x_i \ge 0$ $i = 1, 2, \dots, n$

 x_i amount of investment allocated to stock i ρ_i expected return for stock i σ_{ii} variance of the performance for stock i $\sigma_{ij}/\sqrt{\sigma_{ii}\sigma_{jj}}$ correlation between stocks i and j







- A web application for interactive data analysis
- Create notebooks containing live code, text, equations, and visualizations
- A great way to prototype new algorithms, document experiments, share reproducible results
- Originally for Python, now there are kernels for over 50 different languages





Pandas



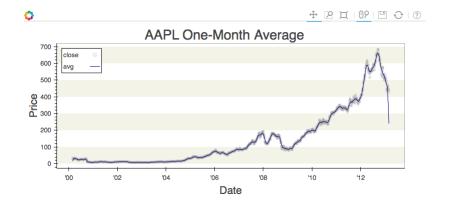
- Fast and efficient DataFrame and Series objects
- Read/write CSV, HDF5, Excel, SQL, plain text
- Missing data handling
- Slicing, fancy indexing, subsetting
- Merges, joins
- Split-apply-combine operations (groupby)



Bokeh

Interactive visualization for web browsers

- High performance with streaming data and big data
- In-browser interactivity on otherwise static pages
- Bindings for Python, Scala, Julia, R







Jupyter Demo

https://notebooks.anaconda.org/mcg/portfolio

```
1
                         (1) A
                                                                                                         O A O
localhost:8888/notebooks/Markowitz.ipvnb
      Upyter Markowitz Last Checkpoint: 21 hours ago (autosaved)
                                                                                                          Python 2 O

    Cell Toolbar: None

                                       C Markdown
                   for (symb,c) in zip(esyms,{'green','red','blue','purple'}):
                       fig.line(tx, growth[symb], legend=symb, color=c, line width=2)
                       fig.text([tx[-1]], [growth[symb][-1]], [symb], text_font_size='10px', text_baseline='middle
                   # Equal allocation
                   eq growth = growth.mean(axis=1)
                   fig.line(tx, eq growth, color='maroon', legend='equal allocation', line width=2)
                   # All the other stocks, and the text labels
                   for symb in syms:
                       if symb not in esyms:
                           fig.line(tx, growth[symb], color='maroon', alpha=0.25, name=symb)
                           fig.text([tx[-1]], [growth[symb][-1]], [symb], text font size='10px', text baseline='mic
                   # Expand the y-axis to give us a good view
                   fig.v range = Rangeld(growth.values.min()/1.25,1.25*growth.values.max())
                   fig.legend.orientation='bottom_left'
                   show(fig)
                                                                          4 2 I I C
                    10^1
```

Closing thoughts











Analysts benefit from Jupyter:

- Interactive, iterative development
- Notebooks combine code, results, visualizations, documentation

Developers benefit from Spyder:

- Traditional IDE organization
- Graphical debugging

Both benefit from straightforward collaboration:

Code from notebooks readily inserted into standard Python modules



Wrap up

- The Gurobi Python interface brings the expressive power of a modeling language into the powerful Python ecosystem
- Spyder offers a traditional IDE for Python development, including fullfeatured editing and graphical debugging
- IPython and the Jupyter notebook system provide a powerful paradigm for interactive development, collaboration, and sharing
- The Anaconda platform delivers a suite of powerful open-source tools and connects to world-class commercial software like Gurobi
- Need training, consulting, enterprise support? Continuum can help





Download Gurobi and/or Anaconda and try them for yourself:

www.gurobi.com/get-anaconda

Email: sales@gurobi.com

Twitter: @Gurobi

Sales: sales@continuum.io

Twitter: @ContinuumIO



