Using Python to Create Robotic Simulations for Planetary Exploration

Pycon 2010, Atlanta, Georgia

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Background



- Jet Propulsion Laboratory
 - NASA Center focusing on robotic planetary missions
 - Mobility and Robotic Systems (http://robotics.jpl.nasa.gov)
 - Dartslab creates simulations for robotic vehicles
 - Rovers, Airships, spacecraft in space (eg, Cassini), spacecraft entering a planet's atmosphere

http://dartslab.jpl.nasa.gov





MER rover in Victoria crater



A Roams Simulation of Opportunity's Descent into the Victoria Crater

DARTS Lab Jet Propulsion Laboratory August 2007



Purposes of vehicle simulations

- Develop and test vehicle control algorithms
 - Accurate physics, camera imagery
- Mission planning (or analysis of mission planning)
- Vehicle stand-in simulations for developing/testing flight software
 - High fidelity physics
 - Faster than real-time
- Analysis (plotting, multiple simulations for performance changes to parameter variations)
- Create and test new vehicles simulations
 - Develop and test vehicle concepts

Software Architecture Issues

- Modular code for wide range of applications
 - Many modules for multibody dynamics, vehicle physics, hardware device models, terrain modeling, camera modeling, vehicle control, ephemeris, domain specific models, etc
 - Mix and match modules for specific problems
- Agile environment to build simulation models quickly
 - Script-based language really helps speed customization
- Rapidly evolving code
 - New problem areas
 - Improving common code modules
 - Need to mix and match modules without recompiling everything

Software Architecture Highlights

- All Dartslab software is C++ & Python and is organized as independent modules & libraries.
- Use **Python** as glue to configure different applications at run-time
- Use C++ library backend for "heavylifting', and Python for rapid prototyping – combines performance with flexibility.
- Includes support modules for visualization, large-scale Monte Carlo parametric simulations, data logging, introspection, checkpointing etc.



How do we use Python?

- Python interfaces to C++ code using SWIG
- Runtime module loading (via SWIG)
- Connecting C++ code at runtime
- Parameter data management
- User-defined callback routines
- Regression testing

C++ & Python Roles within a module

C++

- Computationally expensive algorithms (eg. multibody dynamics)
- Time critical software
- Foundational classes
- Mature and stable code
- Math libraries

Python

- User interface
- Command line options
- Configuration scripts
- Templates
- Regression test scripts
- GUIs
- Loading parameters
- IPC
- Matlab, Mathematics I/F

Goal:

Migrate most low-level code / libraries to C++



Why Python?

- Scripting language (like Tcl, Matlab), but with rich object oriented language and namespace support
- Python OO implementation schema can be made close to C++ OO software design for deployment
- Very expressive software constructs as well as meeting computational/algorithmic needs
- Easy to import functionality from third party C/C++/Fortran libraries as Python extensions
- Ideal for prototyping no compilation needed
- Scales well with software complexity (unlike Tcl or Perl)
- Supports named arguments
- Facilitates "live" implementation of requirements as well as validation experiments
- Comes with "batteries included" many extension modules
- Supports command line help, inspection & introspection
- Doxygen can process Python code to generate documentation
- Supports doctests combines documentation with testing

Developing Software...

Donald Knuth's strategy for software development:

- 1. Get it working 🗖
- 2. Get it right
- 3. Then optimize

Python has been a very effective language for making these transitions

Tony Hoare/Donald Knuth: "Premature optimization is the root of all evil."

Why Prototype with Python?

- To flesh out and mature issues related to
 - Architecture
 - Design
 - Requirements
 - Usage
- Refactoring existing code
- Working directly in C++ is not a problem for incremental or modest extensions (if design is clear)
- Large scale extensions, significant refactoring, or new module development can be more expensive and time consuming to do directly in C++
 - C++ complexity
 - Makefiles
 - Declarations
 - Compiling & linking
 - Limited access to mid-level TPS modules, eg. Sockets, XMLRPC, XML etc. that straddle multiple domains

Python connects C++ code

- Using Python to glue together C++ code at runtime
 - Although low-level classes are C++, we use Python classes to interact with C++ objects. Once connected all interactions are at full C++ speeds.
 - Use Python data (dictionaries) to configure and set up low-level C++ objects. Once set up, the objects interact with C++ mechanisms; therefore they are faster.
 - Reduces need for monolithic complied programs
 - Lazy loading of C++ modules/libraries

Python for regression testing

- Using Python for regression testing
 - Use doctests for testing (not PyUnit yet ⁽²⁾)
 - Large regression test suite
 - Run regression tests hourly, daily, and with major releases
 - Results reporting through Django website



Python for regression testing

- Reporting through internal Django website
 - Virtualenv
 - "flot" Javascript library for plotting
- Identifies when regtests fail
- Helps maintain large codebase despite rapid code evolution



Conclusions

- Good experience with Python
- An important tool in the Dartslab