A Short Course on Python and Julia

National University of Singapore Workshop

John Stachurski and Matthew McKay

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Assumptions

• Programmers but unfamiliar with Python / Julia

Aims

- Background, overview and comparisons
- Help with installation
- Lower fixed costs to getting started
- Provide resources for further study

Thanks

- **1. NUS**
- 2. Sponsors supporting quantecon.org







See

- http://quantecon.org/
- http://www.numfocus.org/

Team

Lecturer: John Stachurski

- Dynamic programming, asset pricing, Markov process theory
- Computational methods (mainly theory)

Will be assisted by Matt McKay

Pythonista and computational development economist

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Resources

Workshop homepage:

• https://github.com/QuantEcon/NUS_workshop_2016

Other

- Cheat sheets: http://cheatsheets.quantecon.org/
- Lectures: http://lectures.quantecon.org/
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Software options

- 1. Install Anaconda Python (Python only)
- 2. Install Julia + Anaconda Python
- 3. Use our server http://workshop.quantecon.org:8000/

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Testing, testing

Try starting

- Julia REPL (REPL = Read Eval Print Loop)
- Python and IPython REPLs

Use menus or terminal

terminal in UNIX/macOS and cmd in Windows

Overview of Scientific Computing

Tasks

- Solve numerical problems
- Produce figures and graphs
- Manipulate data
- Explore (simulate, plot, visualize, etc.)

And sometimes we need speed

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The Need for Speed

Maximum speed:

- Optimal use of hardware
- High level of control over calculations / logic

First best = assembly / machine code

Individual instructions at the CPU level

Example here sums 1+2

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Now imagine a heterogeneous agent model with 5 state variables...

And of course you need to optimize for specific hardware

- pipelining
- cache hierarchies
- branch prediction
- coprocessors
- etc.

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Lesson: There's a trade off between

- · machine speed
- coding efficiency

Low level languages give us fine grained control

High level languages give us

- abstraction
- automation of some tasks
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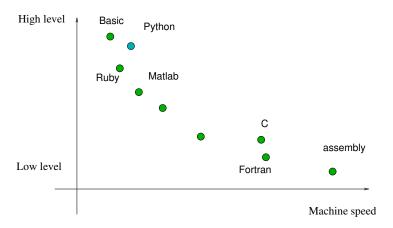
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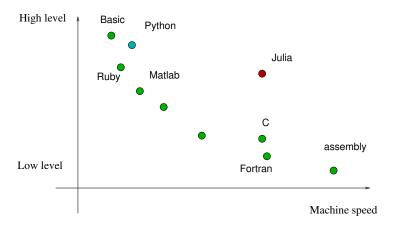
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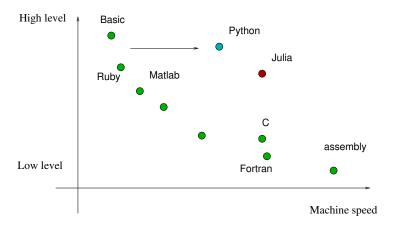
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Although the curve is starting to shift...





A horse race

• See fast loops

Overview of Python and Julia

- History
- Pros and cons
- Which to choose?

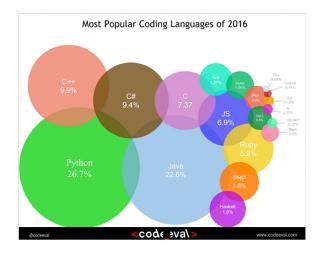
What's Python?

Modern, high level, open source, general purpose programming language

Used extensively by

- Tech firms (YouTube, Dropbox, Reddit, etc., etc.)
- Hedge funds and finance industry
- Scientists (academia, NASA, CERN, etc.)
- etc., etc.

Very popular in "data science" / machine learning



Strength 1: Intelligent, modern design

- Often used to teach first courses in comp sci
 - MIT
 - Stanford
 - Chicago, etc.

Strength 2: Simple, readable syntax

```
for name in name_list:
if name not in list_of_users:
    print("Your name is not found"
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Other pros

- Great libraries
- Friendly community
- High productivity

Some negatives for scientific work

- Need scientific libraries
- Need some tricks to get speed

What's Julia?

Modern, high level, open source, scientific programming language

Strengths:

- High productivity...
- and high performance!

Negatives

- Still under development
- The "rabbit hole" of advanced features (plus or minus?)

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Julia vs Python

Happy Julia user:

- Focused on scientific programming
- Write your own algorithms
- Need optimization / high performance

Happy Python user:

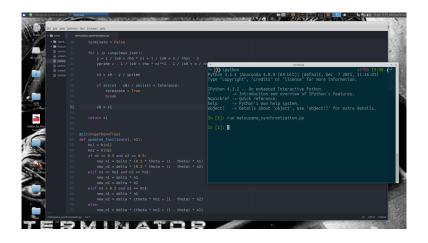
- Care about stability and high productivity
- Diverse coding needs
- Use a lot of data / empirics

Interacting with Python / Julia

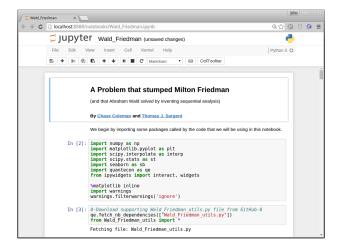
Options

- 1. The REPLs
- 2. Text editors (e.g., Atom or Sublime) plus the REPL
- 3. IDEs like Spyder and Juno
- 4. Jupyter notebooks

Atom + REPL



Jupyter notebooks



Jupyter Notebooks

Let's focus on Jupyter notebooks

- A browser based front end to Python, Julia, R, etc.
- Allows for rich text, graphics, etc.
- Easy to run remotely on servers / in cloud

References and examples:

- Ref: http://quant-econ.net/py/getting_started.html
- Examples: http://notebooks.quantecon.org/