

# 2017 QuantEcon Workshops

Economic Modeling with Python and Julia

Introduction

September 2017

# Team

- Chase Coleman – NYU
- Victoria Gregory – NYU
- Matthew McKay – QuantEcon
- John Stachurski – ANU
- Natasha Watkins – QuantEcon

**Thanks to the Alfred P. Sloan Foundation ;-)**



# Workshop Timeline

1. **9:30 am – 1 pm**: Python
2. **1 pm – 2 pm**: Lunch
3. **2 pm – 4:30 pm**: Julia

# Morning Timeline

1. **9:30-10:30** Introduction and First Steps
  - John Stachurski
2. 10:30-10:45 break
3. **10:45-11:45** Data Analysis with Python
  - Natasha Watkins
4. 11:45-12:00 break
5. **12:00-13:00** Advanced Data Analysis with Python
  - Matt McKay

# Afternoon Timeline

1. **14:00-15:00** Introduction to Julia with Applications
  - Chase Coleman
2. 15:00-15:30 break
3. **15:30-14:30** Dynamic Programming with Julia
  - Victoria Gregory

# Aims / Outcomes / Expectations

Aims != learn Python + Julia + libraries + etc.

Aims =

- Give an overview
- Show some examples
- Discuss / argue
- Resources for further study

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# Workshop Resources

Cheatsheets, downloads, etc. — see

[https:](https://quantecon.org/2017-phd-workshops-on-computational-methods)

[//quantecon.org/2017-phd-workshops-on-computational-methods](https://quantecon.org/2017-phd-workshops-on-computational-methods)

Look for Resources

Download workshop files from the GitHub repo

- via **git** or the **Download** button

# Downloads / Installation / Troubleshooting

## Install Python + Scientific Libs

- Install Anaconda from <https://www.anaconda.com/downloads>
- Not plain vanilla Python

# Jupyter notebooks

A browser based interface to Python / Julia / R / etc.

Step 1: Open a terminal

- on Windows, use Anaconda Command Prompt

Step 2: type `jupyter notebook`

# Cloud-based server

`workshop.quantecon.org:8000`

Password = **economics**

## Topics:

- opening a notebook
- executing code
- edit / command mode
- installing `quantecon`
- getting help
- introspection
- math and rich text

# Background — Open Source

## Proprietary

- MATLAB
- STATA, etc.

## Open Source

- Python
- Julia
- R

closed and stable vs open and fast moving

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# Background — Language Types

Low level languages give us fine grained control

## Example. $1 + 1$ in assembly

---

```
pushq    %rbp
movq     %rsp, %rbp
movl     $1, -12(%rbp)
movl     $1, -8(%rbp)
movl     -12(%rbp), %edx
movl     -8(%rbp), %eax
addl     %edx, %eax
movl     %eax, -4(%rbp)
movl     -4(%rbp), %eax
popq     %rbp
```

---

High level languages give us abstraction, automation, etc.

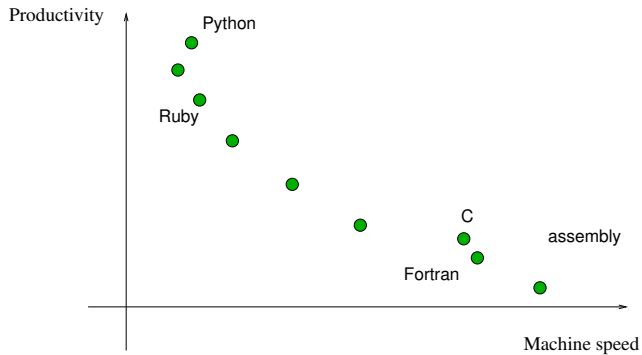
## Example. Reading from a file in Python

---

```
data_file = open("data.txt")
for line in data_file:
    print(line.capitalize())
data_file.close()
```

---

# Trade-Offs

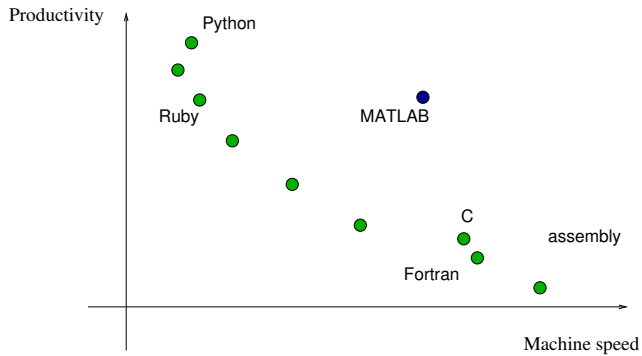


# But what about scientific computing?

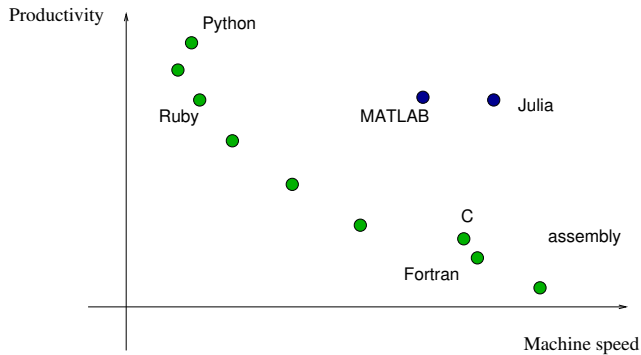
## Requirements

- Productive — easy to read, write, debug, explore
- Fast computations

# Trade-Offs

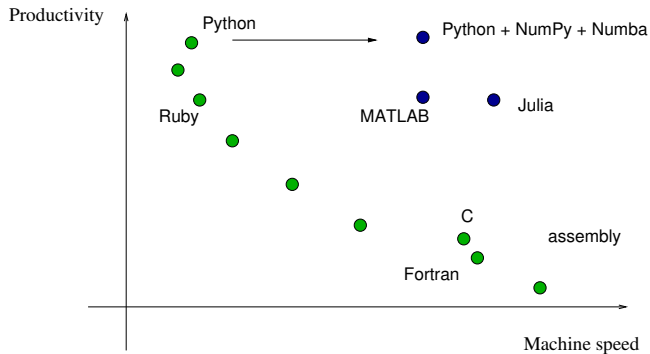


# Trade-Offs





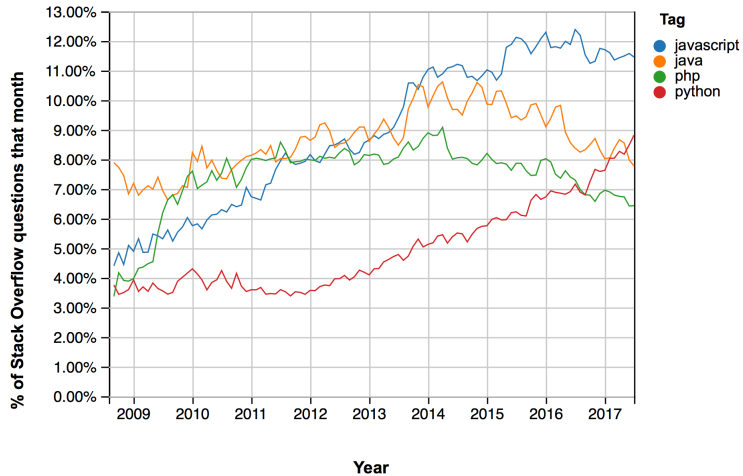
# Trade-Offs



# Python for Productivity

From local infrastructures to cloud-based systems to building websites to interfacing with SQL databases, Python has nearly limitless applications. Despite its wide-ranging impact, it remains gloriously clean and easy to learn.

– *mashable.com*



# Python for High Performance Computing

- See `John/numba.ipynb`

# Exercises

- See `John/plots.ipynb`