RSE-QuantEcon Computational Economics Workshop

An introduction to computational methods for economics and finance

10:00am - 12:00am February 16th 2022

Personel

- Thomas J. Sargent (NYU)
- John Stachurski (RSE)
- VC?

Topics

- Introduction to scientific computing
- Option pricing with Python
- Discussion of high performance computing
- Dynamic programming with Python

Assumptions:

- econ/computer/maths/stats literate
- programming not required

Aims:

- Discuss options
- Review trends
- Learn techniques

Resources

• https://github.com/QuantEcon/rse_comp_econ_2023

What are the major trends in scientific computing?

- what's driving them?
- how can we benefit?

Trend 1: Proprietary \rightarrow Open Source

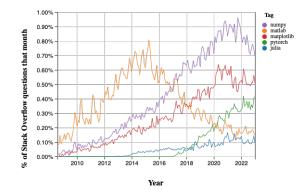
Proprietary

- Excel
- MATLAB, Mathematica
- STATA, Eviews, SPSS.

Open Source / Open Standard

- Python
- Julia
- R

closed and stable vs open and fast moving



Which Language?

Trend 2: Low Level \rightarrow High Level

Low level

- C/C++
- Fortran
- Assembly

High level

- Python
- Javascript
- PHP

Low level languages give us control

- control CPU
- control memory

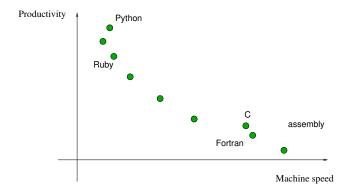
High level languages give us

- abstraction
- automation
- flexibility, etc.

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

```
Example. 1 + 1 in C/C++
```

```
#include <stdio.h>
int main() {
    int sum = 1 + 1;
    printf("1 + 1 = %d\n", sum);
    return 0;
}
```

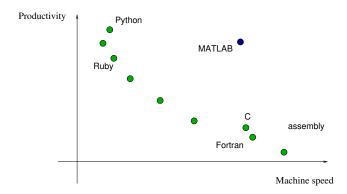


Which Language?

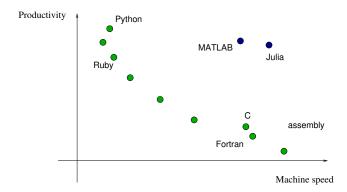
Requirements:

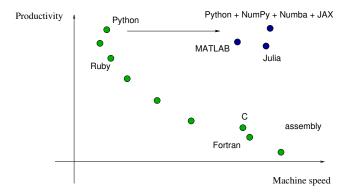
- <u>Productive</u> easy to read, write, debug, explore
- Fast computations

Trade-offs:



Which Language?





Which Language?

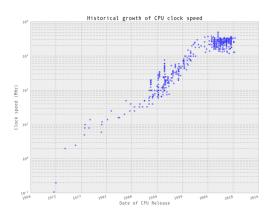
Example. What platforms/languages does OpenAl use?

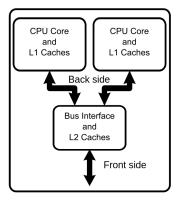
In order (according to repo stats):

- 1. Python
- 2. C/C++
- Javascript
- 4. Jupyter notebooks
- 5. Ruby

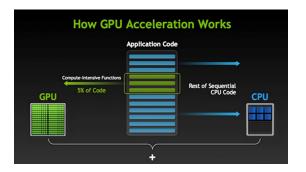
Trend 3: Parallelization

CPU frequency (clock speed) growth is slowing





Source: Wikipedia



Applications: machine learning, deep learning, etc.

Support for Parallelization

While scientific computing environments best support parallelization?

- Most have some support
- but which make it easy to harness its power?

Current winner:

Google JAX (Python library)

Which Language

How about R?

- Specialized to statistics
- Huge range of estimation routines
- Popular in academia
- Loosing some ground to Python (AI, machine learning)

Pros:

- Fast and elegant
- Many scientific routines
- Julia is written in Julia

Cons:

Low rates of investment in some important libraries

- Easy to learn, well designed
- Massive scientific ecosystem
- Heavily supported by big players
- Strong support for parallel computing
- Huge demand for tech-savvy Python programmers

Accessing Python

Option 1: Via a service (remote option)

• https://colab.research.google.com

Option 2: Local install (Python + scientific libs)

- Install Anaconda from https://www.anaconda.com/
 - Select latest version
- Not plain vanilla Python

How to Interact with Python?

Many options:

- write with VS Code / Emacs / Vim
- run with base Python, IPython, etc.

Or do both with Jupyter notebooks / Jupyter lab

for simplicity we focus only on the last option

Jupyter Notebooks

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

Search for jupyter notebook

Useful for:

- getting started
- exploring ideas

Working with Notebooks

- Entry and execution
- Markdown
- Getting help
- Copy paste
- Edit and command mode

RA Work

We and other academics are looking for talented RAs

Ideal skills

- git and GitHub
- Python
- scientific libraries
- maths
- stats
- economics