

# ECON-GA 1025 Macroeconomic Theory I

## Lecture 1

John Stachurski

Fall Semester 2018

# Introduction

The first half of ECON-GA 1025 – Macroeconomic Theory I

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# Contact hours

## Lectures:

- Times: Monday & Wednesday, 9:30–11:30
- Location: Room 517, 19 West 4th

## Recitation:

- Times: Friday 12:30–2:30
- Location: Room 517

# Resources

## Course notes:

- *Lectures in Quantitative Economics: Theory and Foundations* by John Stachurski and Thomas J. Sargent
- Permanently available at [https://github.com/jstac/nyu\\_macro\\_fall\\_2018](https://github.com/jstac/nyu_macro_fall_2018)
- Will change! Don't print!
- Related: <https://lectures.quantecon.org/>

## Supplementary reading:

- *Recursive Macroeconomic Theory* by Lars Ljungqvist and Thomas J. Sargent, MIT Press, fourth edition, 2018, chapters 1-7
  - Due out Sept 11 (according to MIT Press)
- *Recursive Methods in Dynamic Economics* by Nancy Stokey and Robert E. Lucas, Harvard University Press, 1989

## Other favorites

- *Analysis for Applied Mathematics* by Ward Cheney, Springer Science, 2013
- *Introduction to Real Analysis* by Robert Bartle and Donald Sherbert, Wiley, 2011

# Assessment

Assessment = assignments + exam

## Assignments

- Compulsory but not graded
- Must be of reasonable quality
- Work together but submit alone!
- All assignments independently written
- Posted each Sunday, due following Friday before recitation

Exam is on Oct 22nd, more details later...

# Topics

- Deterministic dynamics
- Linear stochastic models
- Markov chains
- Search problems
- LQ and discrete decision problems
- Optimal savings and consumption
- The theory of dynamic programming

# Mathematics

Two strands

- Analysis
- Probability theory

We will use analysis for solving equations and optimization problems

What is/are the solution/solutions to these equations?

1.  $x = ax + b$

2.  $x = x + 1$

3.  $x^2 = 1$

Now let  $x$  be  $n \times 1$  and  $A$  be  $n \times n$

When does this **vector equation** have a solution ?

$$Ax = b$$

When does this vector equation in  $\mathbb{R}^n$  have a unique solution?

$$x = Ax + b$$

When does the **method of successive approximations** converge?

1. pick any  $x_0 \in \mathbb{R}^n$
2. set  $x_{n+1} = Ax_n + b$  for  $n = 0, 1, \dots$

Now let's make it a bit curly:

$$x = (b + (Ax)^{1/\gamma})^\gamma, \quad \gamma > 0$$

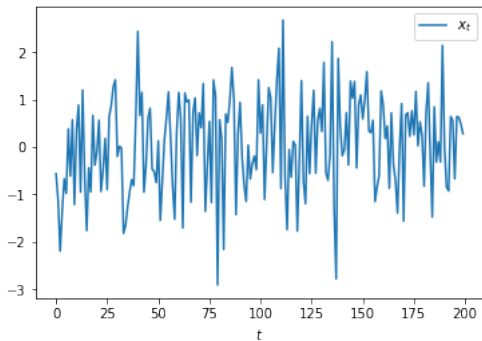
When does this have a solution?

Is it unique?

How would we compute it?

# Probability

This sequence is IID



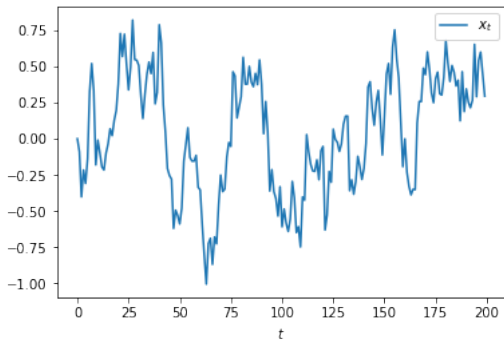
Does it follow that, for some function  $h$ , we have

$$\frac{1}{n} \sum_{t=1}^n h(x_t) \rightarrow \mathbb{E} h(x_t) ? \quad (1)$$

If so this is good:

- Left hand side is data
- Right hand side is model
- Now we can compare. . .

This sequence is **not** IID



Does it follow that, for some function  $h$ , we have

$$\frac{1}{n} \sum_{t=1}^n h(x_t) \rightarrow \mathbb{E} h(x_t) \quad ? \quad (2)$$

# Programming

Most of the weekly assignments will require programming

Acceptable languages

- Python
- MATLAB

You should try both

# Programming Background

A common classification:

- **low** level languages (assembly, C, Fortran)
- **high** level languages (Python, Ruby, Haskell)

**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()
```

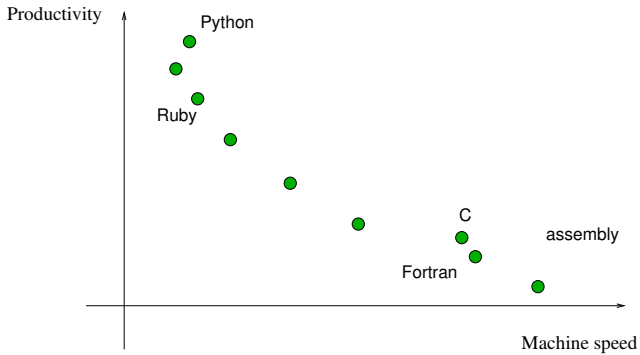
Jane Street on readability:

*There is no faster way for a trading firm to destroy itself than to deploy a piece of trading software that makes a bad decision over and over in a tight loop.*

*Part of Jane Street's reaction to these technological risks was to put a very strong focus on building software that was easily understood—software that was readable.*

– Yaron Minsky, Jane Street

# Trade-Offs

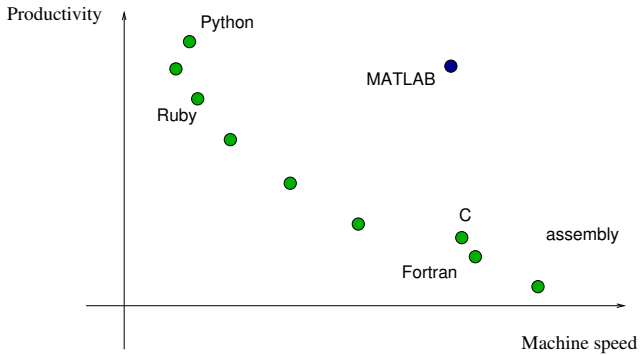


# But what about scientific computing?

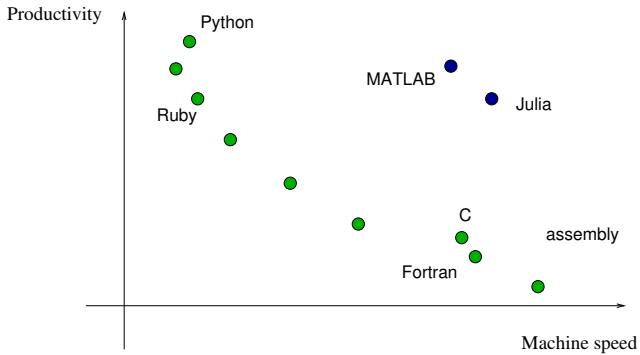
## Requirements

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

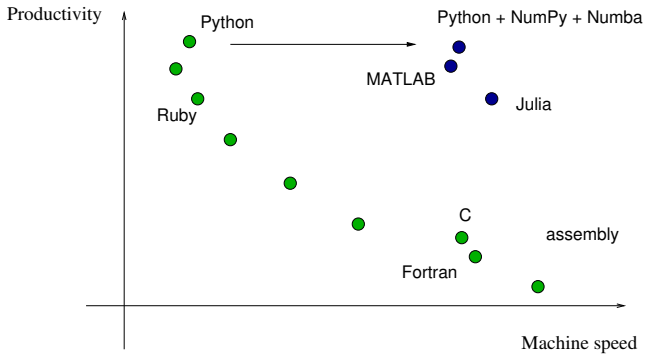
# Trade-Offs



# Trade-Offs



# Trade-Offs

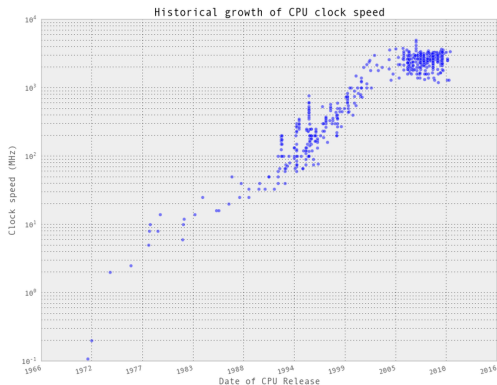


# Key Takeaways

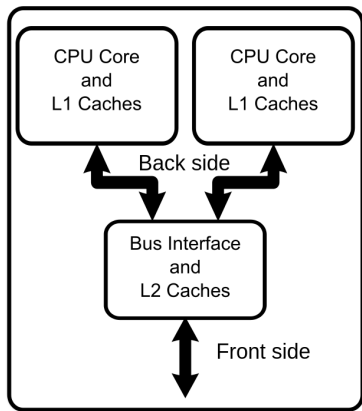
- Don't write in C / C++ / Fortran, no matter what your professor says
- JIT compilation is changing scientific computing
- Same with parallelization
- New algorithms, new techniques — and opportunities

# Programming Background — Hardware

CPU frequency (clock speed) growth is slowing



Chip makers have responded by developing multi-core processors



Source: Wikipedia

Exploiting multiple cores / threads is nontrivial

Sometimes we need to redesign algorithms

Sometimes we can use tools that automate exploitation of multiple cores

# Why Recursive Methods?

This course is an introduction to **recursive** methods for economic analysis

Example.

- Recursive Macroeconomic Theory by LL and TJS
- Recursive Methods in Dynamic Economics by NS and REL

Recursive methods are used to solve high dimensional optimization and equilibrium problems

Breaks the problem down into smaller steps

Helps tackle the **curse of dimensionality**

**Example.** A typical problem from undergraduate choice theory:

Choose consumption at time 0 and 1 to solve

$$u(c_0) + \beta u(c_1) \tag{3}$$

subject to

$$c_1 \leq R(y_0 - c_0) \tag{4}$$

If  $u$  is concave, strictly increasing and differentiable, then the unique solution is found by taking the  $c_0$  that satisfies

$$u'(c_0) = \beta R u'(R(y_0 - c_0)) \tag{5}$$

In general, undergraduate style optimization problems are relatively easy

- All functions are differentiable
- Few choice variables (low dimensional)
- Concave (for max) or convex (for min)
- First order / tangency conditions relatively simple

But PhD macro / PhD research problems are harder...

Possibilities:

- High dimensions
- Can't take derivatives
- No analytical solution for FOCs
- Neither concave nor convex — local maxima and minima

Many interesting research problems have these features

**Example.** A typical problem from graduate macroeconomic theory:

Choose consumption at time  $t = 0, 1, \dots$  to solve

$$\mathbb{E} \sum_{t=0}^{\infty} \beta^t u(c_t), \quad (6)$$

subject to

$$w_{t+1} = R_t(w_t - c_t) + y_t \quad (7)$$

An infinite dimensional problem because we must choose  $c_0, c_1, \dots$

And stochastic!

# Can Computers Save Us?

For any function we can always try brute force optimization

Here's an example for the following function

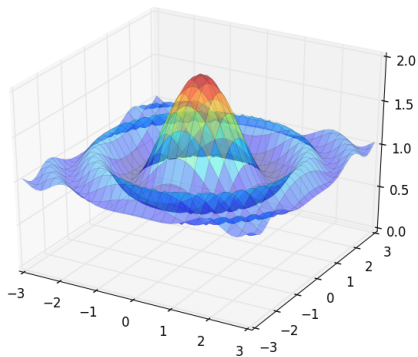


Figure: The function to maximize

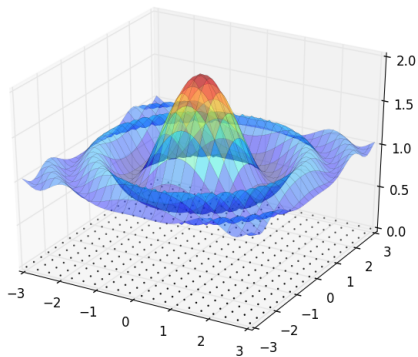


Figure: Grid of points to evaluate the function at

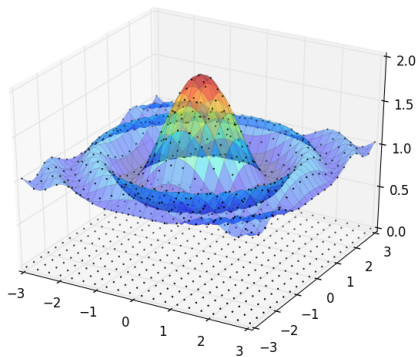


Figure: Evaluations

Grid size =  $20 \times 20 = 400$

## Outcomes

- Number of function evaluations = 400
- Time taken = almost zero
- Maximal value recorded = 1.951
- True maximum = 2

Not bad and we can easily do better

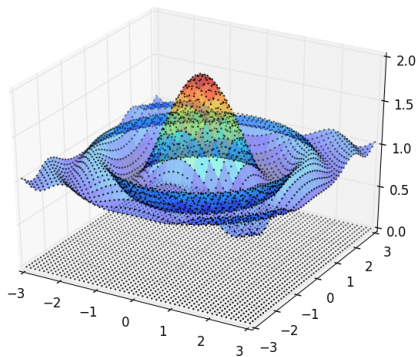


Figure:  $50^2 = 2500$  evaluations

- Number of function evaluations =  $50^2$
- Time taken =  $400 \mu s$
- Maximal value recorded = 1.992
- True maximum = 2

So why even study optimization?

The problem is mainly with larger numbers of choice variables

- 3 vars:  $\max_{x_1, x_2, x_3} f(x_1, x_2, x_3)$
- 4 vars:  $\max_{x_1, x_2, x_3, x_4} f(x_1, x_2, x_3, x_4)$
- ...

If we have 50 grid points per variable and

- 2 variables then evaluations =  $50^2 = 2500$
- 3 variables then evaluations =  $50^3 = 125,000$
- 4 variables then evaluations =  $50^4 = 6,250,000$
- 5 variables then evaluations =  $50^5 = 312,500,000$
- ...

**Example.** Recent study: Optimal placement of drinks across vending machines in Tokyo

Approximate dimensions of problem:

- Number of choices for each variable = 2
- Number of choice variables = 1000

Hence number of possibilities =  $2^{1000}$

How big is that?

---

```
In [10]: 2**1000
```

```
Out[10]:
```

```
107150860718626732094842504906000181056140481170  
553360744375038837035105112493612249319837881569  
585812759467291755314682518714528569231404359845  
775746985748039345677748242309854210746050623711  
418779541821530464749835819412673987675591655439  
460770629145711964776865421676604298316526243868  
37205668069376
```

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Let's say my machine can evaluate about 1 billion possibilities per second

How long would that take?

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```
In [16]: (2**1000 / 10**9) / 31556926 # In years
```

```
Out[16]:
```

```
339547840365144349278007955863635707280678989995  
899349462539661933596146571733926965255861364854  
060286985707326991591901311029244639453805988092  
045933072657455119924381235072941549332310199388  
301571394569707026437986448403352049168514244509  
939816790601568621661265174170019913588941596
```

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What about high performance computing?

- more powerful hardware
- faster CPUs
- GPUs
- vector processors
- cloud computing
- massively parallel supercomputers
- ...

Let's say speed up is  $10^{12}$  (wildly optimistic)

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```
In [19]: (2**1000 / 10**(9 + 12)) / 31556926
```

```
Out[19]:
```

```
3395478403651443492780079558636357072806789899958  
9934946253966193359614657173392696525586136485406  
0286985707326991591901311029244639453805988092045  
9330726574551199243812350729415493323101993883015  
7139456970702643798644840335204916851424450993981  
6790601568621661265174170019
```

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For comparison:

---

```
In [20]: 5 * 10**9 # Expected lifespan of sun
```

```
Out[20]: 5000000000
```

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Message: There are serious limits to computation

What's required is clever analysis

Exploit what information we have

- without information (oracle) we're stuck
- with information / structure we can do clever things

# Getting Started with Python

See [https://lectures.quantecon.org/py/getting\\_started.html](https://lectures.quantecon.org/py/getting_started.html)

Jupyter notebooks

- How to use
- Markdown
- LaTeX
- Getting help

# Homework

Go to <https://lectures.quantecon.org/py/index.html>

Study the following lectures **before** Friday's recitation

- Setting up your Python Environment
- An Introductory Example
- Python Essentials
- OOP
- NumPy
- Matplotlib
- SciPy

Bring your laptop on Friday