#### Complex Systems 530: Computer Modeling of Complex Systems

Lecture 1: Introduction and overview - 1/9/20

## Welcome!

- Today
  - Class logistics & introductions
  - Why model? How to choose what kind of model to use?
  - Agent-based models (ABMs)
  - ABM exploration

# Getting to know you

- Math background?
- Coding background?
- Modeling background?
- Previous complex systems courses?

## Getting to know you

- Name
- Department, stage/year (e.g. Magical studies, master's student, 1st year)
- Math & coding background/level of comfort
- Why you're interested in this course
- Anything else you'd like me to know (optional)

## Course Info

- Instructor: Prof. Marisa Eisenberg
- Office Hours Tuesdays 2-4 pm or by appointment
- Course website & syllabus
  <u>https://epimath.github.io/cscs-530-materials/</u>
  (Go through this together & talk about the plan for the class)
- Course Structure
  - Homework/Labs & Project
- Course Goals & Philosophy

## Course overview

- Focus of course: how to design, build, simulate, visualize, analyze, document, and compare agent-based models
- Along the way, we will also explore:
  - A bit of philosophy of modeling/complex systems
  - Intro/intermediate programming
  - Some probability, statistics
  - Cellular automata
  - Some network (graph) theory & data analysis
  - A bit of collective behavior, pattern formation, & game theory
  - Models in social, biological, and physical systems

## Course overview

- We will not focus on equation-based models (e.g. ODE's, PDE's)—these are also super useful! But there are a lot of existing courses covering these topics (e.g. CSCS 511)
  - (That said, if math is your jam, we will have some of that too!)
  - If you are curious whether your question is better addressed with ABMs vs. other kinds of models, we will discuss some & can talk more as we go

## Goals for the course

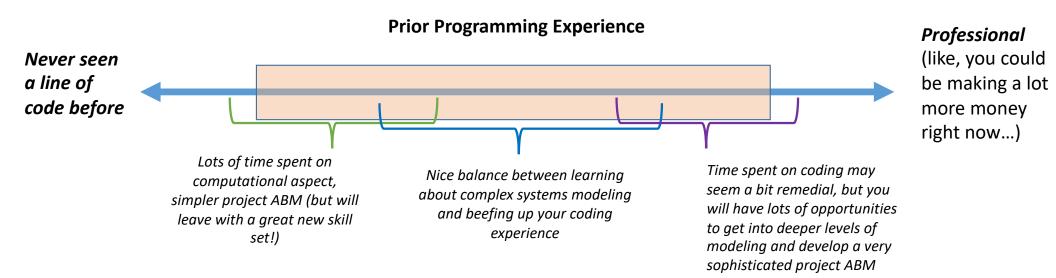
- Learn about models and model-building—how and why to build models for a given problem
- Become a sophisticated consumer of articles involving ABM and gain a solid grounding in some of the key models in the field
- Develop familiarity and competencies in two programming languages, NetLogo and Python
- Leave this class with an ABM you have designed, built, and analyzed that will (hopefully) be useful to you in your future research!

# Languages for Coding

- Example code and HW files will be mostly in NetLogo & Python (3?)
  - See website for more information on how to install
  - Anaconda
- However, for much of the class you can code in whatever language is comfy for you (e.g. R, MATLAB) (within reason...)

# Prior Background

- A background in complex systems is not required, but hopefully you have a strong interest!
- Some basic programming and basic math/stats background is **strongly** encouraged.

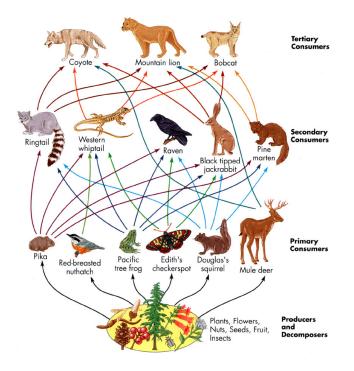


## Laptops

- A laptop is required for this class
- If you don't have access to a personal laptop, we can make arrangements through LSA to get you one for the semester

#### Introduction

#### What is a complex system?

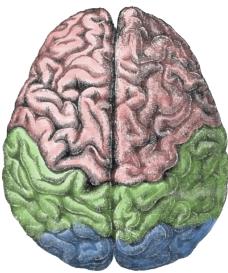


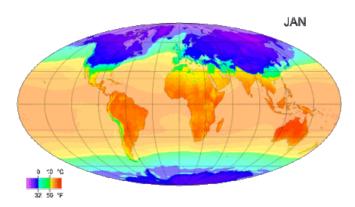




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#### What is a complex system?

- A collection of components that interact, often in nonlinear ways
- Sayama: "Complex systems are networks made of a number of components that interact with each other, typically in a nonlinear fashion. Complex systems may arise and evolve through selforganization, such that they are neither completely regular nor completely random, permitting the development of emergent behavior at macroscopic scales."

## Interactive Example!

1





## Interactive example

- 1) Mark a point at random within the triangle (can really be anywhere but this is simpler)
- 2) Randomly pick a vertex
- 3) Mark your next point: the point halfway between your current point and the vertex you chose
- Repeat steps 2) and 3)

#### Interactive example

• What do you think will happen?

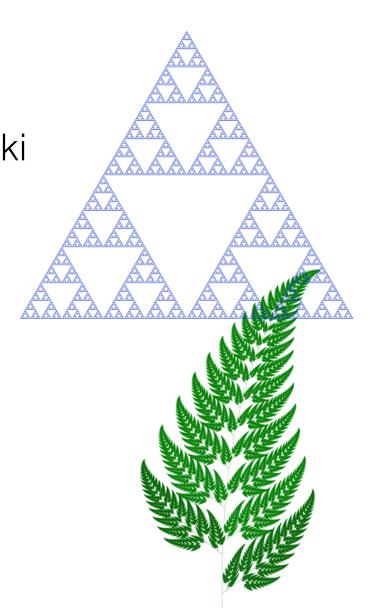
- **Try it out!** Everyone plot 10-15 points on your transparency starting at a random point.
  - Use the ruler to measure carefully!
- Once you're done, we'll combine results

## What did we find?

- Discuss
- Might be tough to see, let's code this process and see what happens with a lot more points!
- (Run code here)

## The chaos game!

- Generates a fractal! This is called the Sierpiński gasket or Sierpiński triangle
- Fractals are objects that exhibit self-similarity (ferns, trees, snowflakes, rivers, & more)
- Simple rules can generate surprising emergent patterns!

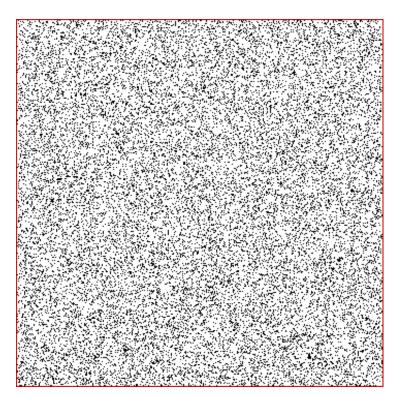


## The chaos game

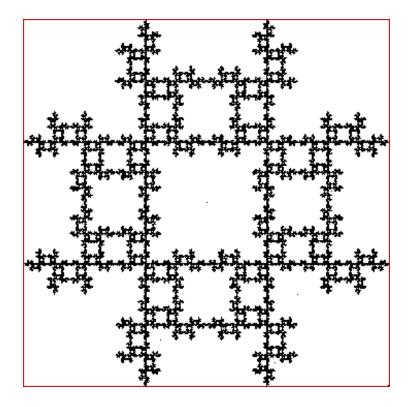
- The Sierpinski gasket is an attractor for this process
- Why does this work? A little intuition—think about transformations—discuss
- We will do more with this later
- Does it work with other shapes?

## The chaos game

#### Try it with a square?



Use a square but don't choose the same vertex twice



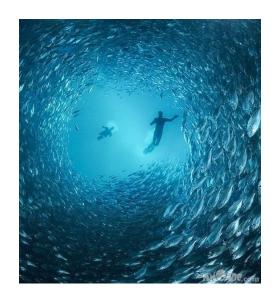
Many other variations!

# Complex Systems

- Emergent behavior
- Self-organization
- Adaptive interactions
- "Fat-Tail" Behavior
- Chaos
- Nonlinearity, tipping points, etc.

## Example: flocking!

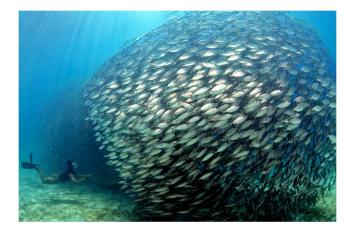
- Flock of geese/school of fish
  - Forms a large, organized pattern
  - But no 'group mind' or leader
  - Birds follow local rules



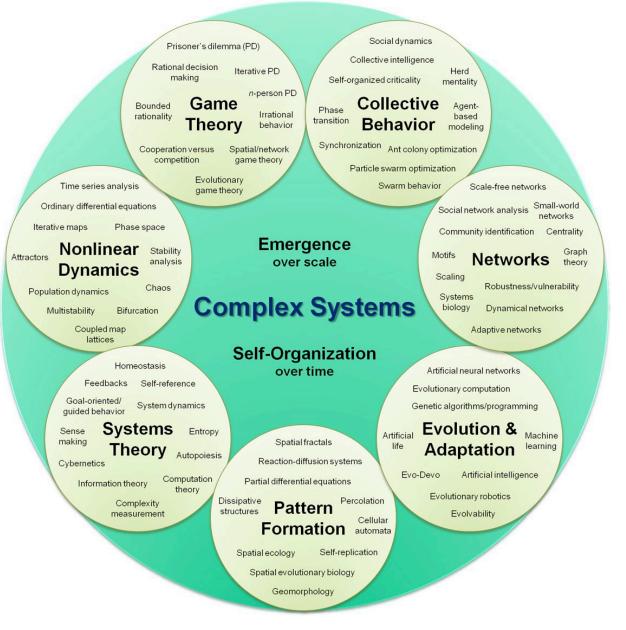
• Result is emergent, organized behavior







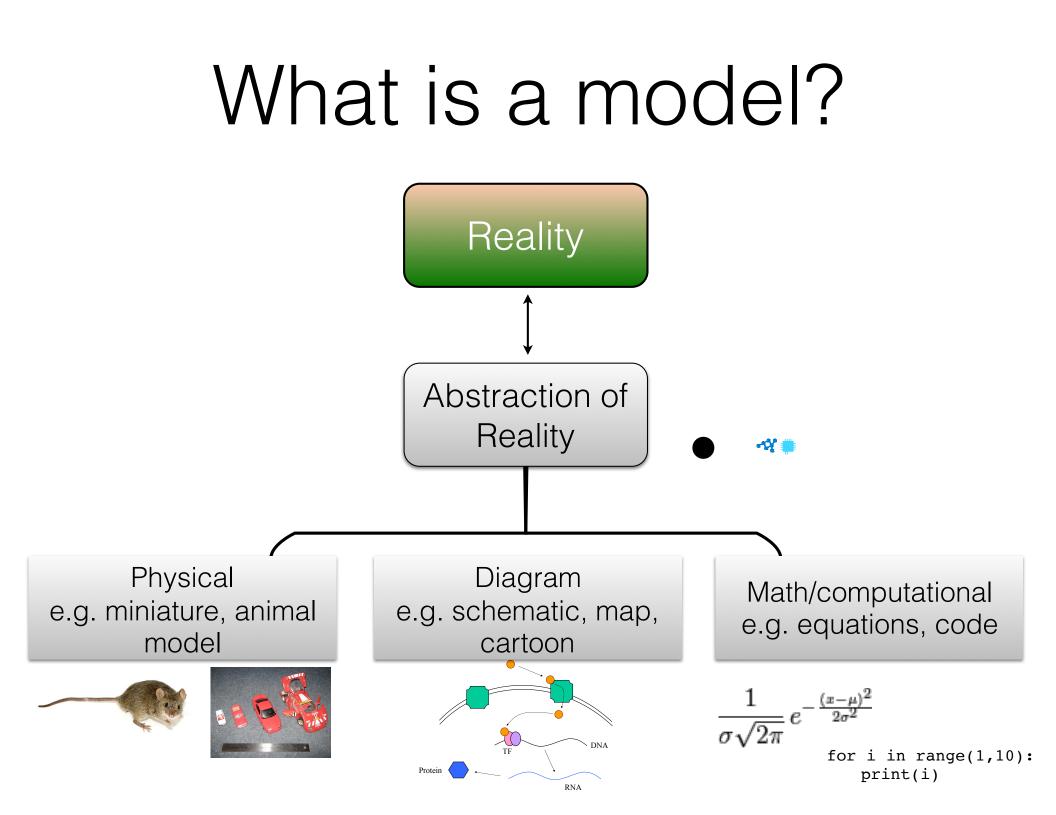
#### **Complex Systems Science**



Introduction to the Modeling and Analysis of Complex Systems by Hiroki Sayama

## What is a model?

• In a few words, what does a model mean to you?



## Models are abstractions

- They are **simplifications**—and so they are **wrong**!
- But they are also often **useful**
- Abstraction is both their strength & weakness
- The trick is, how much to simplify? What details to keep, which ones to remove?
- Even "simple" systems can be complex

• Discuss

- You already model all the time! All of us build mental models, pictures, etc. of the world and the systems we're interested in.
- Really, we are asking: why build explicit models, where we lay out the mechanisms, assumptions, parameters, etc. as equations/code?
  - If we make our models explicit, we can test, interrogate, explore them—can make sure they are self-consistent, explore more complex scenarios, etc.

- Understand mechanisms, causality
- Help to guide & test control, treatment, intervention strategies,
- Test alternative scenarios (counterfactuals)
- Forecast/predict

From Epstein 2008:

- Explain
- Guide data collection
- Illuminate core dynamics
- Suggest dynamical analogies
- Discover new questions
- Promote a scientific habit of mind
- Bound (bracket) outcomes to plausible ranges
- Illuminate core uncertainties
- Offer crisis options in nearreal time

- Illuminate core dynamics
- Demonstrate tradeoffs / suggest efficiencies
- Challenge the robustness of prevailing theories
- Expose prevailing wisdom as incompatible with data
- Train practitioners
- Discipline the policy dialogue
- Educate the general public
- Reveal the apparently simple (complex) to be complex (simple)

Known	Known
knowns	unknowns
Unknown	Unknown
knowns	unknowns

Known	Known
knowns	unknowns
Unknown	Unknown
knowns	unknowns

## Words of caution

- How to know if you have the right mechanism?
- How much do our assumptions and simplifications affect our outcome?
- May not be possible to, for example, predict certain things from the data/understanding that we have! (e.g. early epidemic curve)
- Importance of understanding/testing assumptions, uncertainty quantification, model comparison, etc.