

# Simulation: The basics

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# Roadmap

- ▶ Motivation
- ▶ Concepts
- ▶ Computation
- ▶ Organizational research
- ▶ Frontiers

# Motivation

# What is simulation?

## For our purposes. . .

- ▶ Simulation is a research approach that uses computers to imitate some complex social or organizational system.
- ▶ Researchers have been using simulation to model social behavior for many years (and not always using computers).
  - ▶ residential segregation (Schelling, 1978)
  - ▶ organizational decision making (Cohen et al., 1972)
  - ▶ cooperation (Axelrod, 1984)

## We'll focus on two uses. . .

- ▶ **Understanding the real world**—using simulation to better understand the nature and structure of observed data.
- ▶ **Creating your own world**—using simulation to better understand whether and how small sets of rules can generate complex phenomena.

# Concepts

# Using simulation to understand real world data

- ▶ Researchers often use simulation to understand the nature and structure of real data.
- ▶ When we observe some complex phenomenon in the social world , our inclination is to look for some social process as an explanation.
- ▶ However, even randomly generated data can exhibit complex structures.
- ▶ We may attribute the distribution of word frequencies to the economy of word use (c.f., Zipf) but random text also follows such distributions (c.f., Manning & Schütze, 1999).
- ▶ Within this context, simulation is useful for generating some “null” model against which we can compare our observed data.
- ▶ We saw an example of this in the Bearman et al. (2004) paper.

**Broadly, work that uses this approach often has a “theory testing” orientation.**

# Using simulation to understand real world data

## Challenges (and some solutions)

### **Getting the null model right**

- ▶ You don't need to search long to find a random model from which your real data deviate.
- ▶ But that doesn't mean that your data were generated by a meaningful social process.
- ▶ There are all sorts of constraints in the real world that contribute to observed outcomes.
- ▶ Ideally, you want your null model to incorporate the constraints that are not of interest.

# Using simulation to create your own worlds

- ▶ Researchers also use simulation to generate their own worlds.
- ▶ Within this literature, agent-based modeling is probably the most well-known stream.
- ▶ The basic idea is to define a small set of simple micro-level rules governing the behavior of “agents” and seeing whether these lead to complex macro level phenomenon.
- ▶ For example, Schelling (1978) shows that small in-group preferences can lead to significant residential segregation.
- ▶ We saw an example of this in the Hernandez and Menon (2017) paper.
- ▶ This approach is particularly valuable when complex interdependencies or adaptive behaviors are of interest that would be difficult to model empirically.

**Broadly, work that uses this approach often has a “theory development” orientation.**



# Using simulation to create your own worlds

## Challenges (and some solutions)

### Keeping things simple

- ▶ When you are creating your own world, there are no constraints on what you can do.
- ▶ There is a tendency to make models match too many aspects of the “real world,” thereby making the model too complex.
- ▶ Complex models are difficult to evaluate because it is unclear what features are driving observed outcomes.
- ▶ Also, if your model is complex enough, you can more or less get any result you want.
- ▶ One common solution to this challenge is to “tie your hands” as a researcher a bit.
- ▶ By limiting the complexity of the model, it's probably easier for you to be wrong (and more compelling when you're right).

# Using simulation to create your own worlds

## Challenges (and some solutions)

### Connecting to real world data

- ▶ Another nice solution is to tie your simulation to real data.
- ▶ You might show some interesting emergent phenomenon that results from some micro mechanisms, but does that happen at realistic values of your parameters?
- ▶ If you can use empirical data to inform your model (e.g., values of your “hyper-parameters”) your results are likely to be more compelling.
- ▶ We saw an example of this in the Guimera et al. (2005) paper.

# Computation

# What's the connection to computational social science?

## **At a surface level. . .**

- ▶ Computation is pretty much the de facto (though not the only) approach to simulation.
- ▶ Because they typically entail many interactions and interdependencies, simulations can be very computationally intensive.

## **At a deeper level. . .**

- ▶ Some people would argue that simulation IS computational social science.
- ▶ Simulation really embodies the inductive ethos of computational social science.
- ▶ Given simple rules, what kinds of complex phenomena can emerge?

# Organizational research

# Simulation and organizational research

- ▶ Researchers have been using simulation to study organizations for a long time.
- ▶ However, it has been more widely used in some subfields than others.

## Learning

- ▶ search processes (Levinthal, 1997)
- ▶ exploration and exploitation (March, 1991)
- ▶ decision making (Cohen et al., 2002)

## Networks

- ▶ diffusion processes (Lazer & Friedman, 2007)
- ▶ evolution (Tatarynowicz et al., 2016)

## Structure

- ▶ formal organization (Clement & Puranam, 2018)
- ▶ emergence of culture (Centola & Baronchelli, 2015)

Frontiers

# Where is simulation going next?

My \$0.02

## Understanding “big data”

- ▶ The massive increase in availability of “big data” has created a lot of opportunities for research.
- ▶ But there’s also a lot of noise in big data sets—patterns that seem meaningful but that may not be.
- ▶ Simulation offers a way to compare these patterns to null models.

## Multi-method research

- ▶ Along similar lines, as data has become more available, there is also increasing pressure to measure things directly (rather than purely simulating) when we can.
- ▶ This creates new opportunities for multi method research designs, that couple empirical data analysis with simulation (e.g., for unpacking mechanisms or exploring implications).



# Appendix