



Simulation in the social sciences

Introduction

- What is simulation?
- The uses of simulation?
- History of simulation
- The special features of social simulation
- The stages of simulation-based research



Simulation in the social sciences

■ References

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Models

- There is some “real world” phenomenon which we are interested in. Example, the persistence of inflation in Turkey.
- If the “real world” phenomenon (“target”) is very complex, create a model which is simpler to study.
- A model is a simplification – smaller, less detailed, less complex of a system.
- If the model and the target are sufficiently similar, the conclusions drawn about the model will also apply to the target.



Models

- Model aeroplanes
- Statistical/econometric models
- Natural science models (DNA models)
- Analytical/mathematical models
- Computational models
 - CGE models, simulation models, etc.



Why simulate?

- Verbal descriptions
 - Easy, allow gaps and inconsistencies
- Mathematical descriptions
 - Difficult, precise, rigorous, homogeneous
- Computational models
 - Rigorous, allow complexity, allow experiments



Simulation as a paradigm

- Process analysis
 - Easy, allow not just at one moment in time
- Abstraction
 - Not descriptive
- Macro *and* micro
- Experimental



The uses of simulation

- Understanding
 - MOSES, the Model of the Swedish Economic System
- Prediction
 - DYNAMOD (National Center for Social and Economic Modelling)
- Tool
 - Expert systems to simulate the expertise of professionals

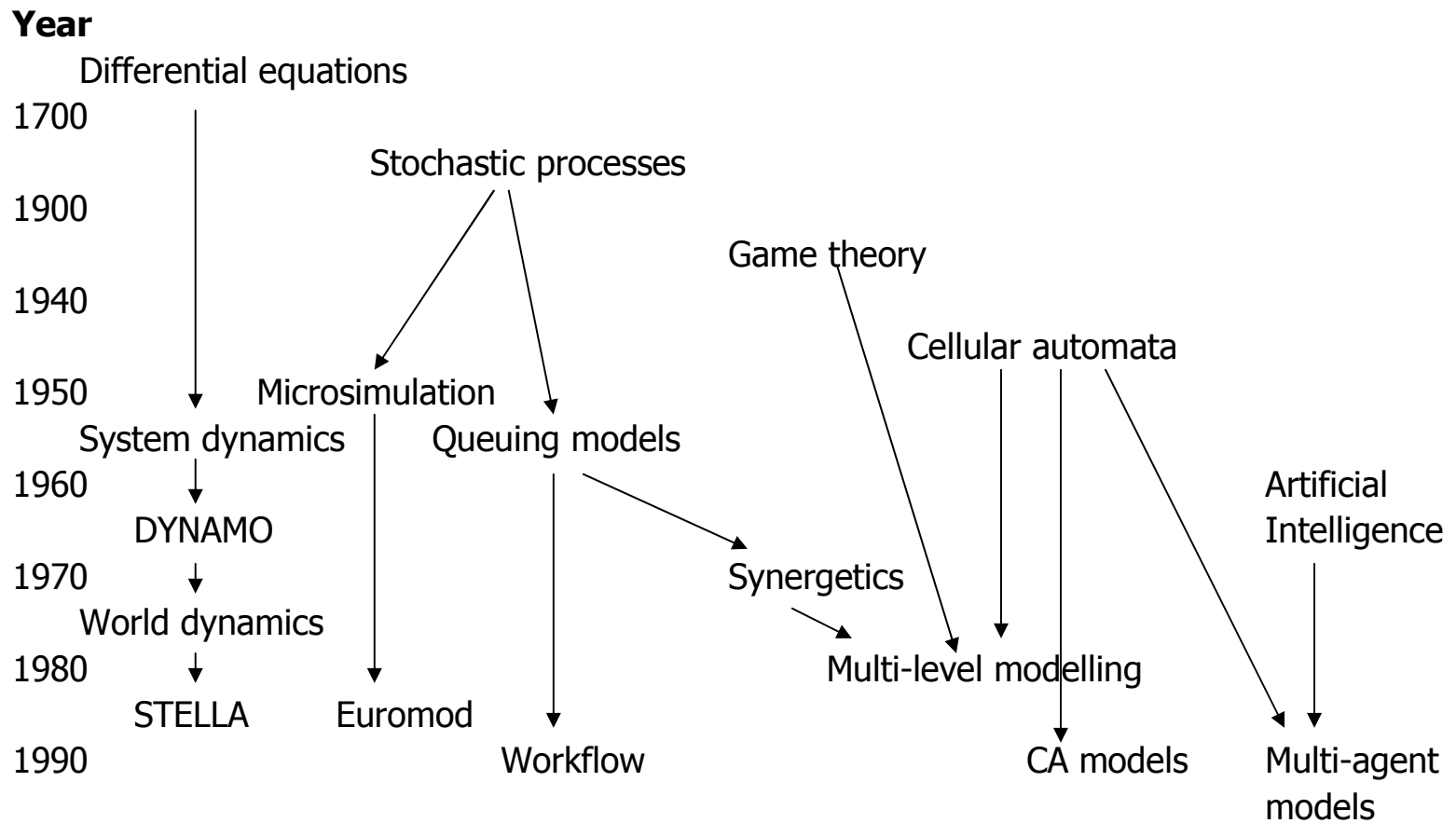


The uses of simulation

- Training
 - Flight simulator
- Entertainment
 - SimCity
- Formalisation
 - MANTA, formalization of theories on self-organization
- Discovery
 - Discover the consequences of theories in the “artificial society”, study the effectiveness of policies



History





Social simulation: Special features

- Non-linearity
- Path-dependence
- Complexity
- Emergence
- Self-organization
- Bounded rationality



Non-linearity

- Linear models
 - Effect on a dependent variable is proportional to the sum of a set of independent variables
- Non-linear
 - Chaotic, sensitive to initial conditions
 - The only generally effective way of exploring non-linear behavior is to simulate it by building a model and then running the simulation.



Path dependence

- Where the system evolves depends on its history: trajectories are important



Complexity

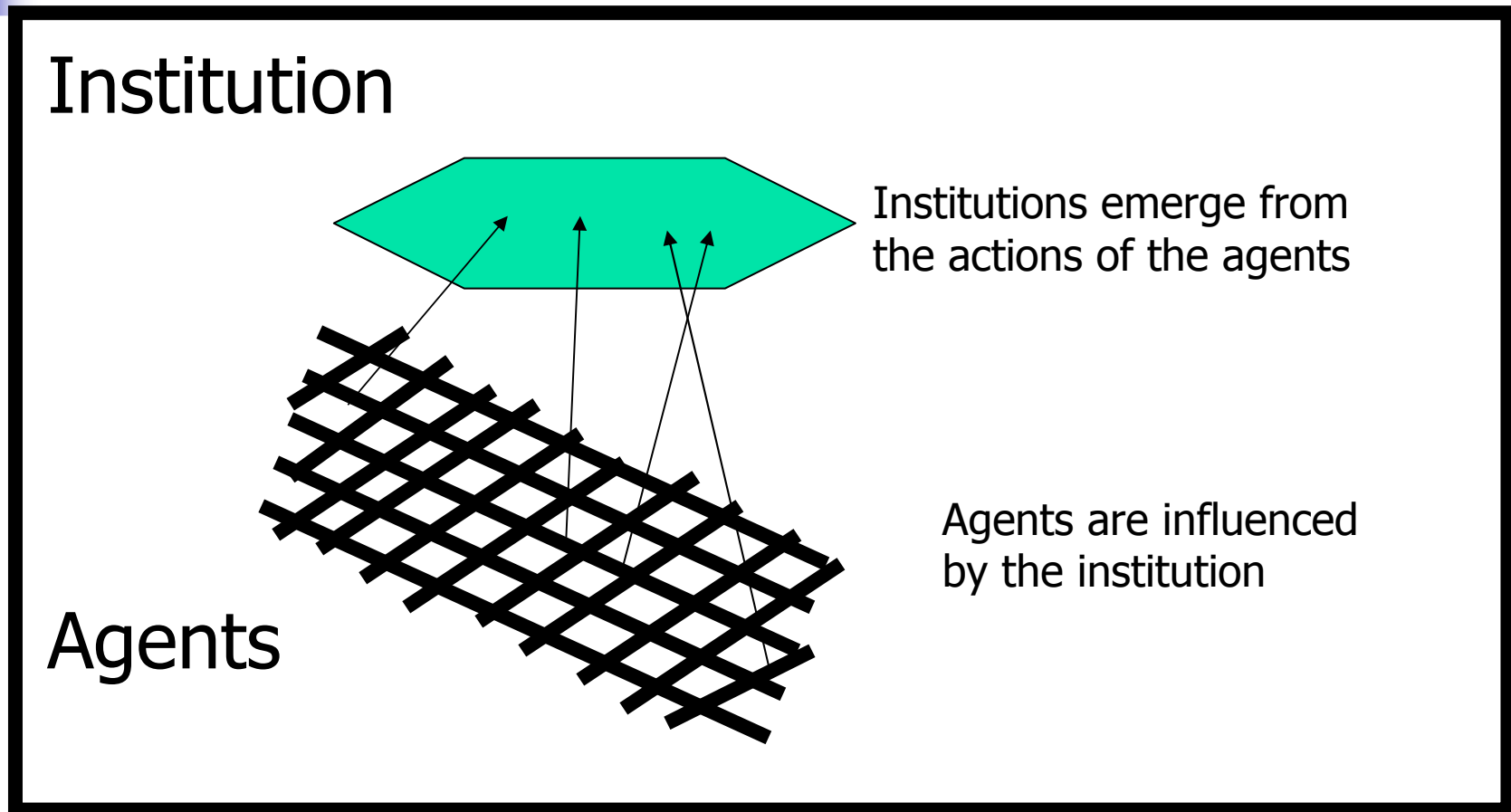
- Simple behavior and simple rules can yield very complex organizations
 - Ants
 - Human organizations



Emergence

- A phenomenon is emergent if it requires new categories to describe it that are not required to describe the behavior of the underlying components
- Example: Temperature is an emergent property of the motion of atoms. An individual atom has no temperature, but a collection of them does.

Second-order emergence





Self-organization

- General organization that appears without central planning from the actions of individual agents
- An emergent property



Bounded rationality

- Herbert Simon (1954): Bounded rationality implies
 - Local information
 - Limited capacity to process information



Classification of simulation models

	Number of levels	Interaction between agents	Complexity of agents	Number of agents
System dynamics	1	No	Low	1
Microsimulation	2	No	High	Many
Multi-level simulation	2	No	Low	Many
Cellular automata	2	Yes	Low	Many
Distributed AI	2+	Yes	High	Few
Evolutionary models	2+	Yes	High	Many



Limits to simulation

- Repeatability and comparability
- Generalisability
- Prediction
 - Qualitative
 - Quantitative
- Data availability
- Validation



Stages of simulation-based research

- Pre-modelling
 - The “puzzle”
 - The objectives
- Modelling
 - Designing
 - Building
 - Verification
 - Calibration
 - Validation
 - Sensitivity analysis
- Reporting



The “puzzle”

- Identify a “puzzle”: A question whose answer is not known and which will be the aim of the research to resolve.



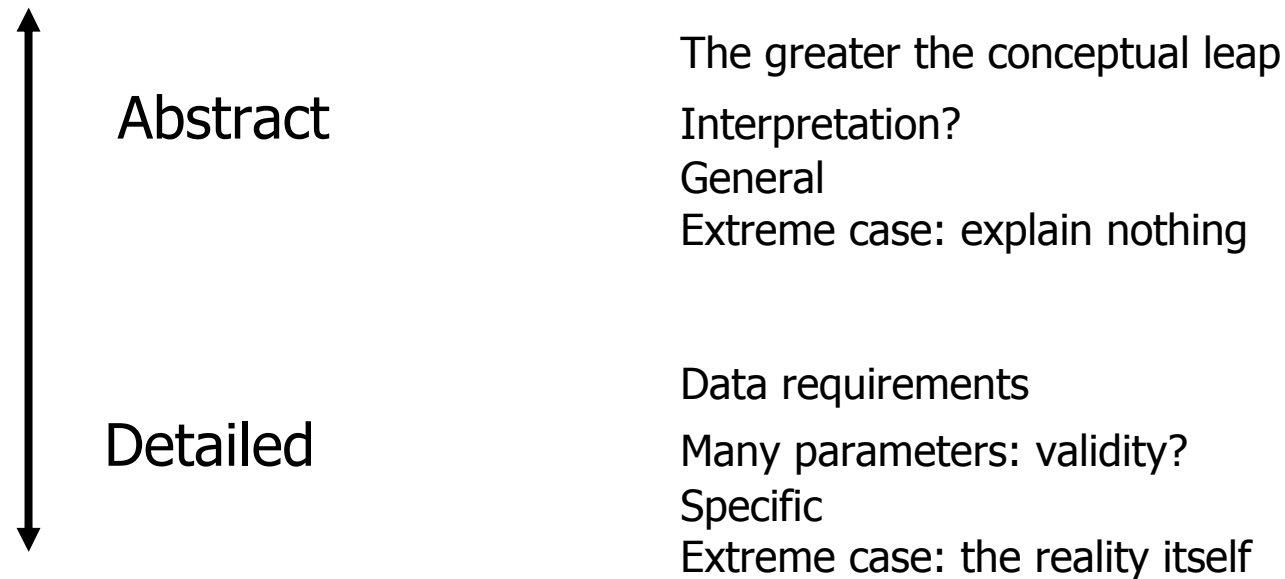
The objective

- What is the objective of using a simulation model?
 - Prediction or understanding?
 - Qualitative or quantitative prediction?
 - Etc.



Design

Every model is a simplification: Decide what needs to be left out and what needs to be included.





Building

- Select a computer language/platform
 - Interpreted languages (Java, Lisp, APL, etc.) vs. compiled languages (C, C++, Pascal, etc.)
 - Package programs
- Structure
 - Agents, interactions, initial conditions, etc.



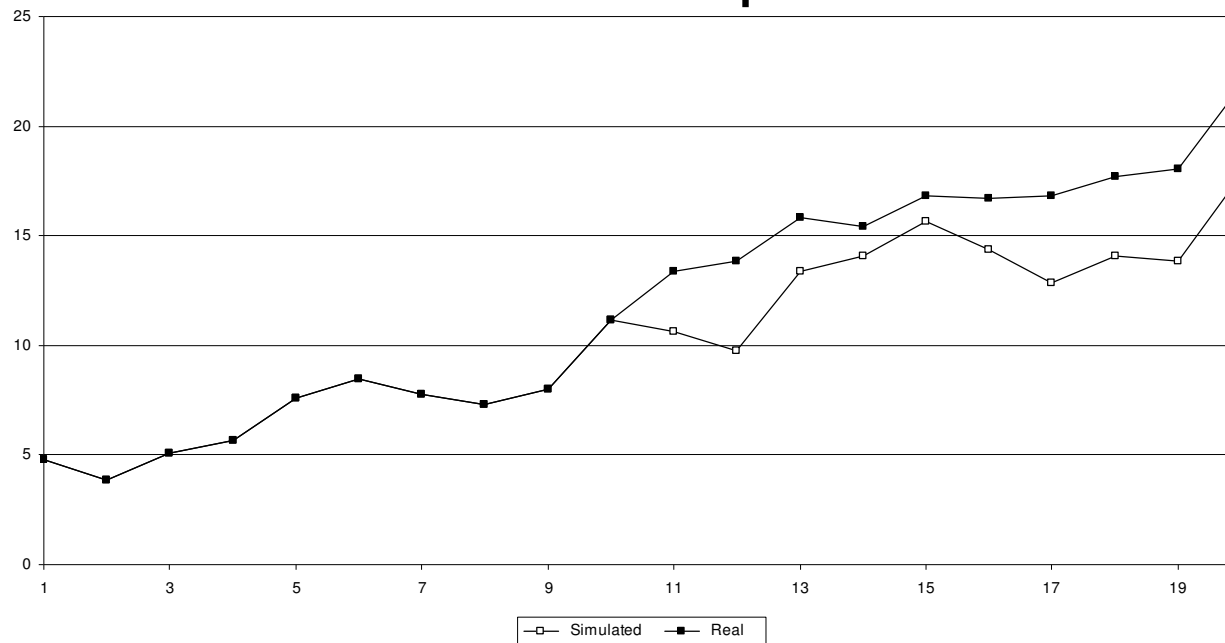
Verification

- Check that the the program does what it was planned to do.
- A difficult and time consuming, but very important task
- Use a modular structure
- Plan “test cases”



Calibration

- Select parameter values such that the simulation data will be as close as possible to the real data.





Validation

- Check if the simulation is a good model of the target.
- Compare the output of the simulation with the data collected from the target
 - Initial conditions
 - Parameter values
 - Stochastic processes
 - Data problems
- Are long run simulations “reasonable”?
- Are “emergent” behaviors/structures are “reasonable”?



Sensitivity analysis

Check if the simulation results are **sensitive** to

- Random effects
- Changes in initial conditions
- Changes in parameter values
- Changes in behavioral specifications



Reporting

- The reader should be able to understand the social science aspects of the research
- The reader should be able to replicate the simulation