Computations, Mechanisms, and Socio-Ecological Systems

A meta-theoretical appraisal of ABM

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Agent-based modeling in Ecological Economics - A useful tool or just a fancy gadget? ESCP Europe Business School 20.05.2016

- I. Motivation of a theoretical assessment of ABM
- II. An epistemological framework for ABM
- III. When ABM is an essential tool
- IV. Critical challenges for ABM to be useful
- V. Summary and outlook



Motivation: The structure of economic research

- Judging the adequateness a *method* is a question of economic *methodology*
- The arguments put forward rely on a particular epistemology
- Discussing the value of a particular
 method benefits greatly from an
 explicit epistemological framework
 - We leave ontology out for now
 - Here: proposal of such a framework to facilitate discussions later

Methodological options

- Economic methodology: which method is adequate for the problem at hand? What are the theoretical implications of different methods?
- What are the main alternatives to ABM when it comes to the study of macroeconomies and socio-ecological systems?

Inductive verbal reasoning

- Theory expressed verbally
- Focus on parametric estimations
- Observation Theorizing Testing

Explorative data-mining

- "Unprejudiced" empirics
- Focus on machine learning and nonparametric econometrics

Applied General Equilibrium Models

- In part. DSGE & CGE models
- Strong relation between mathematical form and empirical investigation
- Estimation/calibration of the parameters suggested by the model
 - From which method can we expect most insight?
 - This requires an epistemological framework
 - Otherwise, it is not clear how ,insight' is to be understood

Further development of Mäki's Models as Isolations and Surrogate Systems (MISS, Mäki 2009)



- Important distinction between *surrogates* and *substitutes* (Mäki 2009)
 - This is where *ontology* necessarily enters the stage
- What is the aim of scientific investigation in such a framework?
 - I. Isolation of the effects of particular mechanisms (i.e. inference of mechanisms)
 - II. Predicting the behavior of the SUI (i.e. inference of facts)...
 - ...based on an mechanistic understanding (c.f. Deaton 2010, Grüne-Yanoff 2016)
 - III. Investigating the interaction effects among different mechanisms



	Verbal theorizing	AGE	ABM
Flexibility of complexity reduction function	++	0	+
Transparency of model exploration	0	++	+

- Further different w.r.t. link to empirical work
 - AGE and ABM allow direct consideration
 - Verbal models use empirics indirectly



- Considering the flexibility on the side of the assumptions (,representation'), and the rigor of the model exploration, ABM are somehow the *golden (?) middle* between analytical GE models and verbal theorizing
- The combination of a flexible mean of complexity reduction and a 100% deductive model corpus represents a *distinctive feature* of ABM compared with its alternatives
 - 100% deductive if numerical algorithms understood as logical rule of inference, as common in econometrics and AGE literature
- Additional feature: affinity to *constructive* mathematical reasoning (c.f. the *generative social sciences* of Epstein (2007))



• Whether an ultimate identification of mechanisms is possible remains unclear...

- ...but perfect prediction is certainly not feasible (c.f. Laplace's demon)
- For progress, both representation and and resemblance part of models must advance
- Calls for realism on several ontological levels at once
- May even suggest moving beyond Occam's razor as a discrimination tool
 - Alternative: *deepness of an explanation* (Bunge 1997, Gräbner & Kapeller 2016)

ABM vs. GET models: Taking the Lucas critique seriously

- Reference point for DSGE and CGE: stagflation in the 70s, partly due to misleading policy advice based on the *Cowles Commission approach* (see e.g. Moneta 2005)
- Lucas critique (1976): necessity for sound micro foundations of economic analysis and consideration of expectations
 - Methodological consequence: Use of *rational* expectations (RE)
- Focus on RE reduces the degrees of freedom of the complexity reduction function
 - No real micro calibration of assessment of intermediate results, e.g. Euler equation
 - Features such as true uncertainty and the corresponding heuristics (Gigerenzer et al.) are excluded from analysis
 - Cannot provide surrogate for situations in which such features are *essential*
- Due to focus on representative agents prohibits the study of particular mechanisms, including aggregation mechanisms (c.f. SMD results, see also Fagiolo & Roventini (2012))
- ABM as the ,better' answer to the Lucas critique?!

ABM vs. verbal models: On consistency and surprise

- All verbal model do reduces the complexity of the real world just as as every formal model
- They provide a very flexible form of complexity reduction, but:
 - The deductions within verbal models are potentially fuzzy and may entail (unrecognizable) creative steps
 - Investigation of the combination of mechanisms can be difficult
 - In contrast to verbal models, ABM can cause *surprise* (Roos 2016)
 - Without formalization the link between data and theory is less stringent
- Verbal models usually use the language ,English'
 - Good to describe some things...but bad when it comes to other
 - There are examples on which it is easier to talk in the language of ABM

The benefits of an algorithmic language: computations

- ABM provide a different language than real analysis or English
 - Different set of vocabulary, e.g. that of a *computation*
- Directly relate to the problem of the *complexity* of decision problems
 - Every economic agent faces decision problems
 - Since an optimal computation of these problems is not feasible, agents develop *heuristics*
 - This includes the computation (or: estimation) of rational expectations
 - What if the way humans use heuristics carried explanatory power?
- Essence of Newell & Simon (1972) was that humans think *algorithmically*
 - Vantage point for a modern account of heuristics (e.g. Gigerenzer & Gaissmaier 2011)
 - Assumption of rationality eliminates any reasonable consideration of computations

ABM and socio-ecological systems

- What are *essential properties* of socio-ecological systems? What are the corresponding methodological demands?
 - 1) Heterogeneity among the social agents, other-regarding preferences, relevance of local institutions and communication (Ostrom et al. 1992)
 - → Way people make their decision becomes important (c.f. computations)
 - 2) Coupling of social and bio-physical dynamics, complex decision problems
 - Resulting true uncertainty for the decision making agents and application of heuristics
 - 3) Spatial settings matter, e.g. infrastructure, access to resources, land use, effect on behavior in dilemma situations
 - Necessity to account for interaction topology via explicit networks
- In many empirically informed cases, ABMs will be necessary (see also Janssen & Ostrom 2006)

Summary: the theoretical pros of ABM

- Compared with verbal theorizing and analytical GET
 - ...golden (?) middle between adequate representation of the system investigated...
 - ...and transparent model exploration
- Allows a *mechanism-based* explanation of reality
- Provides a language that is sometimes richer than English (example: computations)
- Allows a reasonable coupling of theoretical and empirical work
 - Important distinction to machine learning or non-parametric econometrics

Critical challenges for ABM

- There is a number of potential methodological shortcomings of ABM
 - I. Intransparency
 - ➡ What is the cause for a particular model dynamics?
 - ➡ Need for adequate model *verification*
 - II. Overparametrization and ad-hoc specification
 - ➡ Why choosing exactly this kind of decision making?
 - ➡ Need for adequate model validation
 - III.Communication of the model
 - Journal space is scarce and (most) economists have a fetish for particular equations...
 - ➡ Need for ,standard procedures' (?)
- The flexibility of ABM is often called a major weakness: potentially ad-hoc
- Here: stress the potential for adequate validation on different ontological levels

An ongoing challenge: validating ABM

- Exceptional property of ABM: flexibility but concreteness of the complexity reduction function and mechanisms in the model
 - Flexibility on various ontological level as distinctive feature of this method
 - "Dis"advantage: forces us to be explicit
- Potential answer: Empirical validation...
 - ...on various ontological levels
 - This is distinctive for ABM
- Status quo: Important progress for the macro calibration of ABM:
 - Quantitative approaches include Werker-Brenner approach (Brenner & Werker 2007), history-friendly modeling (Malerba et al. 2016), ...
 - On individual level: dominance of qualitative and anecdotal evidence
- To exploit the full potential of ABM we require tools for the simultaneous validation of micro and macro level of the ABM
 - Promising suggestions: Guerini & Moneta (2016)

Summary and outlook

- Is ABM necessary for the problem at hand?
 - The answer to this question always refers to an epistemological framework
- There are situations in which ABM are an indispensable tool to understand certain systems because...
 - 1. ... the essential properties of the system under investigation are such that they cannot be captured in an analytical model
 - 2. ... the algorithmic language of ABM allows a realist approach to phenomena otherwise only understood through narrow instrumental approaches
 - 3. ... the implications of mechanisms or their particular combination cannot be conjectured verbally
- Full potential of (macro) ABM requires progress with validation procedures
- But also proof-of-concept ABM have a number of desirable epistemological properties
 - Close relationship to empirical results from within and outside economics
- ABM is not a cure-all but many of its weaknesses also bear potential

Thank you for listening!

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