# Multi-Scale Resolution of Human Social Systems: A Synergistic Paradigm for Simulating Minds and Society

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#### 1 Introduction

Recently, we put forth an initial sketch of what we call the *Resolution Thesis*[2]. The thesis holds that 1) models of cognition will be improved given constraints from the structure and dynamics of the social systems in which they are supposed to be embedded, and 2) the resolution of social simulations of agents will be improved given constraints from cognitive first principles<sup>1</sup>. This thesis reflects a variety of motivations, the most obvious being the observation that there is little overlap between the cognitive sciences and the generative social science approach[\*REF EPSTEIN, 2008], both of which rely heavily on computer simulation to understand aspects of human systems, albeit different aspects of human systems with respect to scale. The former focuses almost exclusively on the mind as a scientific object of study for which the lion's share of simulation efforts reflect methods that represent a generalizable conception and the latter emphasizes multiple aspects of social systems, the mind being only one of these aspects. For this latter case, there is little relation to the vast experimental evidence of how the mind operates or what are the central theoretical entities that compose the mind. To a large degree, the resolution thesis is a recognition that an interdependence between cognitive science and generative social science has yet to be leveraged for the purposes of improving our understanding of both.

The Resolution Thesis can be understood from multiple perspectives. From the generative social science perspective, the Resolution Thesis means that the representations of agents in social simulations should be informed closely by cognitive science and relevant neurophysiological considerations. This runs somewhat counter to the principled adherence to simplification of the internal processing of simulated agents found in this literature, one that, in fact, served to show that complex social dynamics can be driven by simple behavioral rules of

<sup>&</sup>lt;sup>1</sup>Cognition considered as theoretical models of human perception, thought and action that includes, broadly, explanations of emotion, motivation, and affect in addition to more tradtional domains of cognitive psychology and cognitive science; one could arguably use the term psychological first principles as equivalent

agents. However, more recently, there are efforts in the generative social sciences that acknowledged that closer ties to the psychological and neurophysiological underpinnings of human behavior may yield benefit. Epstein's neurocognitive approach is a notable effort in this vein[REF, Epstein 2014]; there are other related approaches (e.g. REF 19-22 from BRiMS). These efforts notwithstanding, there remains a large gap between them and the implementation of models from cognitive science and psychology, not necessarily in principle, but in practice. It is worth noting that there are some threads in cognitive science that align with this perspective of the *Resolution Thesis*. In particular, Ron Sun's push for multi-agent systems based on cognitive agents and using cogitive first principles as the basis for the social sciences[4]; other work in this vein exists[?]. The relatively new field of computational social psychology is clearly relevant as well [5] as well as the computational organizational theory approach[3].

From the view of cognitive science, the thesis means that patterns of organization (e.g., information flow on the internet [REF; FROM SOCIALSIM], clustering of behaviors in a community[REF CHRISTAKIS]) at the social and organizational level should inform cognitive models. In other words, when possible, these patterns should be included as convergent evidence for a theory or model of cognition. At first consideration, however, this notion might seem hard to fathom. The implications of cognition for the structure and dynamics of social systems are little understood from the cognitive science perspective, and so without an explanatory scheme that links facets of the cognitive model to aspects of social organization, how do we

(Anderson's Relevance Thesis, a rare exception, reasons about how cognition may have implications for social organization.) THe work above (sun, etc.) points in this direction also, but is not quite driven by providing resolutionn to the cognitive models, but about simulating social systems from cognitive first principles. To be clear this is only one component of the thesis.

But that may be precisely the point—it is natural, once of the disposition to think about how cognition scales to social systems, to think about using the degree to which it scales accurately as part of the convergent evidence for having confidence in the model.

A third and more general view is that the Resolution Thesis is about human systems for which the distinction between neural, cognitive and social levels of scale should be considered counter-productive. An understanding of any of these levels of scale is dependent, to some degree, on an understanding of the others. In effect, the notion of convergent evidence as originating, in part, from other levels of scale, applies to all levels of scale. The implication is that we should leverage information across scale in an iterative and synergistic way, if not simultaneously.

The Resolution Thesis, despite sounding both reasonable and practical at face value, is opposed by argument from both the cognitive sciences and generative social sciences. Simon's notion of nearly decomposable systems—that the temporal dynamics of adjacent levels of scale, in most systems, are little correlated—suggests that we can understand well the dynamics at each level of scale independently of the others (\*REF, SIM 1962; see P. Anderson, 1972,

for similiar argument in phyical systems). The KISS principle (keep it simple, stupid) used heavily in the generative social sciences, is clearly akin to Simon's notion, and is bolstered by its early wins in understanding the behavior of social systems using simple, non-cognitive agents (REF 15-17 from BRIMS). In cognitive science, Newell, in considering the time scale of human behavior, suggested that the social band (>  $10^4$  seconds, representing social systems and organizational behavior) is characterized to be weak in strength in the sense that it may not provide computations in a systematic way (relative to lower temporal bands, e.g., cognitive and neural processes).

These counter arguments notwith standing, our working assumption is that the degree to which the  $Resolution\ Thesis$  is useful is an empirical issue. The state of the art in technology, computing and the their tight coupling to the current social mileau affords, we think, the testing of the Resolution Thesis. To this end, we've developed the  $Reciprocal\ Constraints\ Paradigm\ (henceforth\ RCP)$ , a methodological approach that might lay a foundation for fruitful social simulation and an understanding of the implications for theory and models across levels of scale.

# 2 The Reciprocal Constraints Paradigm

The value of the *RCP* does not lie in precise formal presecription, but in providing a scaffold for building understanding of social systems, and, possibly providing a deep understanding of the interdependencies among levels of scale. Further, the *RCP* was conceived when considering multi-agent systems and cognitive science where human behavior, in both cases, is implemented in a simulation enviorment in which agents are represented algorithmically. This is not meant to preclude other methods and approaches (e.g., analytically tractable economic games), but just to provide the context for the development of the ideas presented here.

Figure 1 shows the four primary components of the RCP: a cognitive system, a social system, and the constraints between levels of scale. We assume that social systems and cognitive systems are derived from and exhibit an abstract set of first principles & properties, called S and C, respectively (e.g., theoretial entities, experimental paradigms, patterns in empirical data in respect to the discipines that address a particular level of scale). Further, defining S and C will depend on the social system or cognitive system of interest. The notion of constraint simply refers to the use of information from S & C to inform one another in a principled way. An upward-constraint refers to information from C informing S; downward-constraints reverse this relation.

A primary example in C is the set of allowable algorithms A such that  $a \in C$  with grounding in empirical work in cognitive science and psychology. Primary examples in S are the social structures, channels of information, and dynamics that characterize a social system, much of which is formalized using graph theory/network science. It is important to emphasize that within S are notions regarding the behavior of agents, not only social structures.

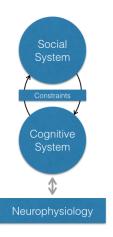


Figure 1: The four components of the RCP are a cognitive system, a social system, and the constraints on each one from the other. In the RCP social systems and cognitive systems are assumed to be derived from and exhibit an abstract set of first principles & properties, called S and C, respectively. Also captured here is the potential for integrating neurophysiolgical considerations into the cognitive system when appropriate; these may prove as essential for some social systems (the grey two-headed arrow indicates this potential). The notion of constraint refers to the use of information from S & C to inform one another in a principled way.

A central assumption in the RCP is that cognitive systems and definitions of agent behaviors in social systems are meant to represent human information-processing capacities that can be described as mathematical functions. [?, ?]<sup>2</sup>. Thus, in C we can define a cognitive system as  $\psi_{ct}: I_{ct} \to \psi_{ct}(i)$  where  $I_{ct}$  is the set of allowable inputs and  $\psi_{ct}(i)$  is the output; in S we have a corresponding agent definition as  $\phi_{at}: I_{at} \to \phi_{at}(i)$  where  $I_{at}$  is the set of allowable inputs and  $\phi_{at}(i)$  is the output for an agent<sup>3</sup>.

## 2.1 Applying the Reciprocal Constraints Paradigm

In practice there are multiple approaches available for application of the RCT, but what unites them is the study of a human social phenomena, either defined at one level of scale or at multiple levels of scale. Naturally, the first step is to identify a social phenomena of interest, a task that is inherently tied to one's perspective. If the perspective is largely in C, then the focus would most likely be on understanding the psychological processes, representations, etc. in relation to social systems, but informed in some way by S. Another perspective, largely in S, would dictate a concern with the social structures and dynamics of the social system (multiple humans interacting) with some degree of constraint from C. These two perspectives are both what we call single-scale approaches to the RCP. Of course, one could take a multi-scale perspective that treats C and S simultanteously. This perspective likely depends on a simulation approach that captures aspects of C and S in one runnable system.

These two approaches are different, primarily, in terms of the directness of the constraints. Single-scale approaches use information that exists prior to application (in any form it can get it); multi-scale generates simulated information at each level of scale, which, also exists prior to application.[NOT SURE IF WE KEEP THIS IDEA]

We will address the obvious issues and difficulties in applying the RCP after providing a description of some potential methods for application. The goal in this section is simply to express what it might look like to apply the RCP.

#### 2.1.1 Single-Scale Approaches

It is conceivable that a fruitful application of the RCP might occur at one level of scale. Let us first consider the cognitive level of scale. There exists a large literature on neural and cognitive approaches to understanding human social behavior and social psychology<sup>4</sup> with the requisite broad range of methods and theoretical orientations, so defining a social phenomena of interest is natural at the cognitive level. One potential approach, then, for implementing the RCP would be to map general properties of cognitive systems with properties of social systems. For example, a central finding in cognitive science and psychology

<sup>&</sup>lt;sup>2</sup>This is equivalent to Marr's computational level; we will use Marr's computational, algorithmic, and implementation levels of description[?, ?] throughout this paper.

<sup>&</sup>lt;sup>3</sup>Social and cognitive systems may define parameters regarding variability among a set of agents; this is not reflected here.

<sup>&</sup>lt;sup>4</sup>Division 8 of the American Psychological Association is dedicated to social psychology

is that learning mechanisms are sensitive to the order in which information is presented to the system (e.g., a human or computational model of a human). Further, there is a large set of social phenomena that imply some type of learning mechanism (attitude formation [REF ORR]; impression formation [REF MONROE]). Thus, we can conclude, reasonably, that order matters for some social phenomena at the psychological level. The application of the RCT, in this case, means inferring what a distribution of the temporal order of inputs would be given some  $s \in S$ . (This distribution, almost axiomatically, seems dependent upon some propoerties of S if S contains graph G where V(G) and E(G) are the agents and information channels, respectively.) Off-hand, there seem to be candidate examples from the literature regarding human socio-technical networks(e.g., [REF BEARMAN; REF BARABASI WEB]), but what this means precisely would depend on the social phenomena of interest (e.g., early language development may depend on a different  $G \in S$  compared to racial stereotypes).

At the social level of scale, an obvious approach is an analysis of the degree to which  $\phi_{at}$  compares to any  $\psi_{ct} \in C$ . For the case in which  $\phi_{at}$  and  $\psi_{ct}$  are formally well defined, this might be relatively straigtforward<sup>5</sup>. But, there will certainly be cases for which this is not true. For example, imagine that  $\psi_{ct}$  is only defined in terms of an experimental paradigm, theoretical apparatus and the interpetation of data resulting from application of the experimental paradigm<sup>6</sup>. One appraoch for comparison in this case would be to explore the behavior of  $\phi_{at}$  by simulating an experimental paradigm that is isomphorphic to that used to define  $\psi_{ct}$  (assuming  $\phi_{at}$  is defined algorithmically). In essence, this is like running an psychological experiment on artificial agents, an approach that is akin to simulation in the psychological sciences. The output of these experiments could be compared to the patterning and dynamics of human performance in C.

#### 2.1.2 Multi-Scale Approach

The multi-scale approach is simple in principle: build a simulation platform that simultaneosly captures essential aspects of both S and C (e.g., an agent-based model of cognitive agents). The upward-constraints would mean a type of substitution of  $\psi_{ct}$  for  $\phi_{at}$  (substitution might be one approach; for others this might be more akin to a focus or commitment to  $\psi_{ct}$ ). The downward-constraints, generated by some measure of how well the simulated social system matches S as defined by the pheonomena of interest, would serve as a signal that would suggest modifications to S, C or both.

The apparent simplicity of the multi-scale approach is misleading. The details will include some non-negotiable decision points; free parameters will abound. The primary issue is, naturally, to decide where to begin? Does one start with S and import aspects of C or the reverse? One might also consider

<sup>&</sup>lt;sup>5</sup>Potential methods for such a comparison would, idealy, focus not only comparison of input/output functions but also the nature of the runnable algorithm

<sup>&</sup>lt;sup>6</sup>Some might argue that under these conditions, the notion of a functional mapping  $\psi_{ct}$  is nonsensical

starting from scratch and bringing together properties from S and C in a new way. Let's explore these options via a concrete example which will, we hope, surface some of the key decision points.

Imagine we're interested in the patterning of obesity by race/ethnicity, a phenomena of interest with key social and policy aspects (e.g., cultural attractors, social and shared-environmental influence, spatial patterning co-occuring with racial segregation and residential mobility) and, in fact, a pre-existing set of social simulations from which to draw[REF review paper on simulation of obesity. Assume we adopt one of the existing simulation approaches, none of which incorporate much in C. Given this case, one approach would be to substitute  $\psi_{ct}$  for  $\phi_{at}$  while keeping the other aspects identical to the original simulation. To do this, however, because both  $\psi_{ct}$  for  $\phi_{at}$  are input-output functions, there is a minimum requirement to find a reasonable substitution such that both  $I_{ct}$  and  $\psi_{ct}(i)$  are similar to both  $I_{at}$  and  $\phi_{at}(i)$ . This is non-trivial in the sense that there is a limited but variable set of candidate  $\psi_{ct}$  to consider. The field of obesity in the health behavior tradition (argueably a co-opting of social psycholgy first principles) has several candidates, each reasonably considered as different (e.g.,  $\{\psi_{ct_i}, \psi_{ct_i}, \psi_{ct_k}, ...\}$ ). This analysis may be difficult or may require that not all aspects of  $\phi_{at}$  can be captured by substitution. This last issue notwithstanding, as long as the inputs  $I_{ct}$  can be defined in the context of the properties of S (and even if ouputs don't match well), we could still have a runnable simulation.

Next, we must consider the downward-constraints in this scenario. Imagine that along with a pre-existing social simulation from which to co-opt, come empirical data, judged of adequate quality, that could be used to compute an objective function w.r.t. the simulations accuracy. This signal, then, would serve as the downward-constraint, albeit in a relatively ill-structed way taken alone. However, in this scenario, one could imagine systematic variations in the objective function w.r.t. properties of  $\psi_{ct}$  surfaced through sensitivity analysis/Monte-Carlo methods (or another method of search through the space defined by  $\psi_{ct}$ . This leads to another decision point, how do we judge the validity, if you will, of the space spanned in  $\psi_{ct}$ , given this scenario. Surely, some would rightly argue that not all parameters in  $\psi_{ct}$  should be free to vary on theoretical grounds ?. Further, unless the properties in S captured in the simulation completely reproduce the empirical data (within a reasonable degree of tolerance), one needs to consider, given the objective function, whether to vary some components in the simulation that represent something in S. In other words, in the above scenario it is probably not good practice to only focus on changing C, so decisons must be made in this regard.

The above scenario is but one approach, one that is fixing S and importing C. But what if we fix C and generate S instead? What does this look like? We will stay with the obesity example from above, but change the scenario such that we don't know about any simulation approaches that represent mainly S. Instead, imagine that S is composed of a set of population-level empirical studies, some of which include information on social networks and the built environment, and some theoretical statements about peer-influence on social networks.

Furthermore, similar to the scenario above, there is a limited set of candidate  $\psi_{ct}$  to consider  $(\{\psi_{ct_i}, \psi_{ct_i}, \psi_{ct_k}, ...\})$  in the health behavior tradition and, likely, other aspects of C that may not be represented in them (e.g., categorization, learning, and attractor dynamics). Further, only one of these candidates<sup>7</sup>,  $\psi_{ct_s}$ , is in a computational formalism; we decide that this will be our starting point. Analysis of  $\psi_{ct_i}$  reveals that it captures the learning and on-the-fly formation of attitudes towards specific health behaviors (considered a precursor to behavior); it is composed in a general manner such that it applies to virtually any health behavior, and; it is empirically grounded using traditional health behavior theory measurment techniques (in one particular behavioral domain). All of this is useful, but some key components are missing. In particular, the model is mute with respect to the generation of social strucuture and dynamics (e.g., decision making for initiating/disolving relationships; social influence mechanisms in terms of how others' behavior or attitudes can potentially serve as the input to the system). Thus, at minimum, some decision points arise about network structure and the mechanism of socialinfluence—e.g., one could implement a static network topology from human empirical data; in terms of social influence, we might assume that what is spreading are attitudes and that exposure to others' attitudes can serve as input to an agent and, further there is a knowable function between attitude and behaviors relating to obesity, e.g., energy balance behaviors relating to caloric intake and use). (As an aside, we've implemented models of this sort [?].) At this point, we could build out further social structure and dyamics in relation to the problem of interest, racial/ethnic distributions in obesity, and thus would require further decision points. However, at some point, we need to run the simulation and determine how to apply the downward-constraint. We explore this next.

Once the simulation is runnable, the downward-constraint could operate, as describe above, by computing an objective function w.r.t. the accuracy of the simulation compared to extant empirical data. Here, the issues are mainly the same as in the fixing S and importing C case described above., with an important, subtle difference. In this case, although, depending on the quality and provenance of the data sources (to mean, roughly, which level of scale do they originate from), there will be decision points in terms of whether to modify C or S, the focus should, at least initially, be on C. In the limiting case, imagine that the only data we have is from C, e.g., some behavioral studies that reveal something about the learning and formation of attitudes about obesity related behaviors and the relation between attitudes and the behaviors. All of these studies, further, were conducted at the individual-level and reveal little about social structure and dynamics. In this case, the objective function would be computed with respect to how well  $\psi_{ct_i}$  captures the variation in these data alone. In this case, it is possible that aspect of both C and S could be modified in the simulation to optimize the objective function. What is interesting about this limiting case is that C may still depend on S. Getting the S right would

<sup>&</sup>lt;sup>7</sup>There do exist a small handful of computationally implemented health behavior theories, see[?] for a review

be essential.

These two examples are largely hypothetical. The value, we hope, is to provide some sketches of what the RCP might look like in practice. Clearly, there are many issues that are raised, even by these sketches, let alone by the general notion of the RCP and the  $Resolution\ Thesis$ . We address some of the obvious of these issues next.

## 2.2 Issues in Applying the Reciprocal Constraints Paradigm

Both single- and multi-scale approaches to the RCP are fraught with difficulty in application.

- 2. FREE PARAMETER PROBLEM AND Acountable modeling, mere parameter optimization?:
  - 3. What if S is dependent on C?
- 4. Many competing cog models for any phenomena ... Issues of the right cogntive model, lots of cog models....
  - 5. Downward constraints in multi-scale approach....what precisely is changed
- 6. quality of the data...some are quite veridical on some dimensions, but may be misleading on others...e.g., tweets are tmeporal and veridical in terms of content, but inference about meaning, motivation and intention may not be possible. data issues in general...the case when there is little data or there is a patchwork of data.

The second kind is to abstract some details from full-fledged cognitive models, but adhere to a principle of *Accountable Modeling*??.

Have data at both levels or just one...re

# 3 An Example of RCT Implemented

We now offer an examle of a multi-scale approach to the RCP. It is a highly stylized example that focus much on C using the cognitive architecture ACT-R with some properties of a specific social domain, online social coding in GitHub. In particular, [describe simulation sketch here].

It will be useful to introduce ACT-R before continuing to the simulations. ACT-R is a specific approach to modeling human cognition that invokes the idea of a cognitive architecture, a computational instantiation of the structures, mechanisms and representations of the mind that are coherent with respect to another notion, that of a unified theory of cognition[1]. A cognitive model that is developed using a cognitive architecture has a potential a constrains operations (if you will it constraints  $\psi_{ct}$  to limits of the human mind w.r.t. a unified theory of cognition standpoint. Cognitive models based on architectures are structured algorithms of  $\psi_{ct}$ , and, thus, not strictly normative; individual differences are captured naturally in cognitive architectures (e.g., working memory; expertise via domain knowledge). The generation of quantitative predictions given hypothetical situational domains provides rigor in estimating the quality of model in terms of empirical data.

ACT-R is composed of several dedicated modules(e.g., procedural and declarative memory, working memory, percetipn and action) that operate in parallel but asychronously vai capacity-limited buffer interfaces. Each module is composed of independent mechanisms, to include symbolic information processing structures that are combined with equations for the representation f specific regularities and phenomena (e.g., reinforcement learning, power law of practice, power law of forgetting). Further, ACT-R adapts to environmental structure via a number of learning mechanisms. In short, it provides an account of human information processing capability and limitations (e.g., inherent biases) precicely because of the coupling of computational mechanisms and limits on capacity (e.g., working memory, attention).

Using a cognitive architecture, versus any cognitive model, characterizes well what agents are supposed to do in the generative social science approach—adaptive behaviors that are coupled to the social and physical environment. ACT-R, in particular, has modeled a broad range of human behaviors to include language, complex decision making, experimental psychology paradigms, and dynamic task environments (see the ACT-R web site[?]). Of particular relevance for this paper are some efforts using ACT-R as the basis for agents in multi-agent social simulations—e.g, iterated two-player games, bot adversarial and cooperative [?]

In fact

# 3.1 Importing Neurophysiology

Base this on Stocco's paragraph from BRiMs and use it in the context of the ACT-R sim we did above. and

Also, add the paragraph from BRiMS on how neuro to cog is many to one and cog to social is one to one.

#### 4 Discussion

Why not just Deep Learning...?

#### 4.1 Moving Forward

From the single level of scale approach, One idea is to catalog social phenomena captured by psychological models and do a mapping of C to S.  $^8$ .

Downward constraints on neurophysiology? From neuro to cog to social to cog to neuro? Is this where it is headed? Stocco has a paragraph in BRiMS on the cog to neuro, use it.

 $<sup>^8\</sup>mathrm{In}$  the limit, the social features and the cognitive features have co-evolved on an evolutionary time scale

# References

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