

This paper, **The Kuhnian Paradigm**, is a programmatic paper describing Kuhn's philosophy of science as a genuinely new paradigm for philosophy of science. As such it provides the frame for much of the rest of my work in philosophy of science, most importantly:

1) **An agent-based model of Kuhn's Structure of Scientific Revolutions**, where I show how possibly the interactions of individually rational scientists can result in an aggregate pattern of normal science, crisis and revolution. Published in the Boston Studies in the Philosophy and History of Science (eds. Alisa Bokulich and William Devlin)

<http://www.springer.com/philosophy/epistemology+and+philosophy+of+science/book/978-3-319-13382-9>

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2) **A Unified Model of the Division of Cognitive Labor**, a unification of Kitcher and Kuhn's ideas on the division of cognitive labor. This paper won the Philosophy of Science Graduate Student Essay award of the Philosophy of Science Association. Published in the journal Philosophy of Science.

<http://www.jstor.org/discover/10.1086/676670?uid=3737592&uid=2134&uid=2&uid=70&uid=4&sid=21104891646057>

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3) **A Comparison of Two Models of Scientific Progress**, where I build an agent-based model to make explicit Kuhn's (intriguing but rather vague and implicit) model of scientific progress and contrast its implications with the traditional linear model of progress. Published in Studies in History and Philosophy of Science.

<http://www.sciencedirect.com/science/article/pii/S0039368114000211>

The penultimate version of this paper can be downloaded for free at:

[https://www.academia.edu/4917138/A\\_Comparison\\_of\\_Two\\_Models\\_of\\_Scientific\\_Change](https://www.academia.edu/4917138/A_Comparison_of_Two_Models_of_Scientific_Change)

I strongly believe novel developments such as agent-based modeling and scientometric data make it possible to "upgrade" Kuhn's vision. The goal of this upgrade is to close the gap between the appeal of Kuhn's ideas for practicing scientists and the difficulties philosophers faced when trying to analyze them. The result is a new kind of philosophy of science that finds a better balance between philosophical rigor and societal relevance.

7

## 3 The Kuhnian Paradigm

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5  
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7 Perhaps the current situation in philosophy of science is  
8 similar to that of Ancient astronomy. In Ancient astron-  
9 omy, the Aristotelian view based on static, perfect circles  
10 explained the movement of heavenly bodies but was not  
11 empirically adequate. The Ptolemaean view, on the other  
12 hand, allowed good predictions but did not explain them.  
13 Similarly sociological accounts of science are more  
14 empirically adequate but usually fail to explain why sci-  
15 ence works, while philosophical explanations of the  
16 workings of science tend to depart significantly from actual  
17 scientific practice. In an attempt at his own Copernican  
18 revolution, Thomas Kuhn tried to do both: adequately  
19 capture the challenges faced by practicing scientists with-  
20 out losing normative force. This resulted in an immensely  
21 popular book, the *Structure of Scientific Revolutions* (SSR  
22 1962). The vision of an explanatory and empirically ade-  
23 quate general philosophy of science showed the potential to  
24 broaden the scope of philosophy of science to questions  
25 such as why disciplines tend to cluster in schools, why  
26 communities are reluctant to embrace novel frameworks  
27 and what drives scientific innovation. These questions are  
28 highly relevant for practicing scientists but have tradi-  
29 tionally received little attention in philosophy of science.  
30 SSR's success makes it all the more frustrating that  
31 50 years on few philosophers of science would call them-  
32 selves "Kuhnians". Kuhn's view was never used as a  
33 starting point for broad-scale philosophical research but  
34 has itself remained the subject of scholarly debate.  
35 Whereas SSR thanks its immense popularity to the ques-  
36 tions it raised, the paradigm was never sufficiently articu-  
37 lated to provide (apparently much-anticipated) answers. I

claim this gap can now be closed. In this paper I explain 38  
why the gap was there in the first place, why it can now be 39  
bridged and what lies on the other end. In the first section I 40  
argue that the Kuhnian paradigm was not sufficiently 41  
articulated because Kuhn was one of the first to describe an 42  
instance of the interdisciplinary family of models that came 43  
to be known as "complex systems". In the second section I 44  
argue that recent developments provide powerful new tools 45  
for better articulating the Kuhnian paradigm. Kuhn is often 46  
credited for undermining the logical empiricist research 47  
consensus in philosophy of science in his time, but because 48  
of a lack of articulation the Kuhnian paradigm failed to 49  
provide an alternative research agenda, leaving the field in 50  
a state of fragmentation. 50 years on, these new tools might 51  
turn it into the genuine paradigm that Kuhn intended it to 52  
be, including both a descriptive and a normative research 53  
agenda. This program for a *systemic philosophy of science* 54  
is laid out in the third section. 55

### 1 Science as a Complex System 56

Already in SSR's opening pages Kuhn makes clear his rev- 57  
olutionary aspirations. Standard philosophy of science at the 58  
time restricted itself to the context of justification, as a result 59  
of which "it will, therefore, never be a permissible objection 60  
to an epistemological construction that actual thinking does 61  
not conform to it" (Reichenbach 1938, 6). Against this, Kuhn 62  
envisions a more empirical "new image of science" (SSR, 3) 63  
that was not confined to the finished products of science as 64  
represented in scientific textbooks, but based on the histori- 65  
ography of science. However Kuhn's recourse to case- 66  
studies led to disappointment, most explicitly in his essay 67  
*The Trouble with the Historical Philosophy of Science* in 68  
which he calls them "misleading" (Kuhn 2000, 111) because 69

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70 their study only deepens the problems they suggest rather  
71 than solving them. Instead, he came to realize that “many of  
72 the most central conclusions we drew from the historical  
73 record can be derived instead from first principles” (ibid.  
74 112), and these principles “are necessary characteristics of  
75 any developmental or evolutionary process” (ibid. 119).<sup>1</sup>  
76 Here I argue that Kuhn probably realized that his account of  
77 science was an instance of a family of models that came to be  
78 called *complex systems*. This would explain Kuhn’s two  
79 claims above; a complex system is an *evolving* system that  
80 can generate complex structure from *simple* rules. And the  
81 fact that research on complex systems was still in its infancy  
82 during most of Kuhn’s lifetime would explain why Kuhn  
83 failed to articulate his account to the extent that it could be  
84 operationalized.

85 A complex system consists of many components, none of  
86 which needs to interact with all others. What makes it  
87 “complex” instead of merely “complicated” is that the rules  
88 governing these localized interactions change through time.  
89 For example a car is complicated. It consists of many locally  
90 interacting components, but the interaction between those  
91 components is stable and can be centrally controlled by the  
92 driver. Studying their parts is sufficient to understand how  
93 the car works. Traffic, on the other hand, is complex. Drivers  
94 adjust their behavior as a result of previous interactions  
95 (endogenous change), as a result of events outside the system  
96 (exogenous change) and usually both (the endogenous  
97 reinforcement of exogenous chance events). No single driver  
98 controls the system, yet these complex interactions can give  
99 rise to simple, stable patterns such as traffic jams. The system  
100 is thus capable of *self-organization*, a process that can occur  
101 at various levels and thus create a *hierarchy* within the sys-  
102 tem (e.g. food chains). Such patterns are called “emergent”  
103 because they cannot be understood as the sum of their parts.  
104 Rather they must be explained by reference to the *interaction*  
105 of their parts through *time*.

106 Kuhn’s account of science can be interpreted as an  
107 attempt at describing science as a complex system. Kuhn  
108 describes how science emerges from the localized *inter-*  
109 *actions* of scientists through *time*. Exogenous chance  
110 events such as new instruments, new data and personal  
111 idiosyncracies cause change, and this change can in turn  
112 lead to further (endogenous) changes through “a *feedback*  
113 *loop through which theory change affects the values which*  
114 *led to that change*” (Kuhn 1977, 336). Such a feedback  
115 loop allows for the endogenous reinforcement of chance  
116 events, making possible critical events such as “scientific  
117 revolutions” and self-organization such as the emergence  
118 of relatively homogenous networks of scientists who share

conceptual, theoretical, instrumental and methodological 119  
commitments, viz. “paradigms” (SSR, 42). Because of 120  
self-reinforcement “Small changes, however, can have 121  
large-scale effects” (Kuhn and Thomas 1990, 12), as a 122  
result of which very different paths can emerge under 123  
similar initial conditions, allowing incommensurability: 124  
“even men who, being in the same or in closely related 125  
fields, begin by studying many of the same books and 126  
achievements may acquire rather different paradigms in the 127  
course of professional specialization” (SSR, 49). The 128  
interaction between these emergent groupings results in 129  
aggregate patterns that, unlike the interactions of their 130  
components, might be simple and stable. In Kuhn’s case: a 131  
cycle of normal science, crisis and revolution. 132

133 Although the development of the idea of Kuhn’s account  
134 of science as a complex system is material for a paper in its  
135 own right, this brief sketch already indicates its potential to  
136 unify many different elements of his account. In addition I  
137 list here two other indications that suggest this interpreta-  
138 tion is correct. First, the seemingly disparate analogies that  
139 Kuhn used throughout his work, such as institutional  
140 dynamics, political revolutions, biological evolution, eco-  
141 system dynamics and cognitive dynamics, have turned out  
142 to be prime examples of what we now call “complex  
143 systems” (Newman 2011). For lack of a theoretical account  
144 to describe science as a complex system, Kuhn seems to  
145 have taken recourse to analogical descriptions using well-  
146 known properties of other such systems. Second, Kuhn was  
147 a Harvard condensed matter physicist (PhD 1949). Con-  
148 densed matter systems such as magnets, crystals, glasses  
149 and superconductors are among the earliest known exam-  
150 ples of complex systems and these were exactly the kind of  
151 systems that Kuhn had been working on for almost a  
152 decade before turning philosopher.

## 2 New Tools for the Articulation of the Kuhnian 153 Paradigm 154

I have been arguing that the lack of articulation of the 155  
Kuhnian paradigm prevented it from being operationalized 156  
and that this lack of articulation is not due to the questions 157  
it asked but to the answers it provided. Philosophers were 158  
not satisfied with the answers Kuhn provided on all levels: 159  
input, output and their connection. 160

1. The core concepts allow for too many alternative 161  
interpretations (Masterman and Margaret 1970), 162
2. no clear mechanism for rationality and progress is laid 163  
out (Sharrock and Read 2003), 164
3. empirical support is indirect and inconclusive: partic- 165  
ular historical case studies to support claims about 166  
general patterns in science (Kuhn 2000, 109). 167

1FL01 <sup>1</sup> Kuhn had started developing this view in a book project that was  
1FL02 never finished due to his untimely death in 1996 and for legal reasons  
1FL03 the material is inaccessible to this day.

168 In the previous section I have argued that these short-  
 169 comings stem from the fact that Kuhn tried to describe  
 170 science as a complex system while this field was still in its  
 171 infancy at the time. In this section I will argue that this  
 172 situation has now changed. Describing science as a com-  
 173 plex adaptive system requires hypotheses about how pos-  
 174 sibly its emergent features can be produced from the  
 175 localized interactions of its components, and the right data  
 176 to test them. The onset of Big Data, the resulting surge in  
 177 network theory, and the increased computational power to  
 178 analyze that data and carry out simulations, are recent and  
 179 substantial developments that were largely unavailable  
 180 during Kuhn's lifetime. Developments outside philosophy  
 181 have previously shown the ability to have an enormous  
 182 impact on our philosophical understanding. The following  
 183 breakthroughs related to complex systems could be for the  
 184 Kuhnian paradigm what the breakthroughs in logic were  
 185 for logic empiricism.

## 186 2.1 Network Theory

187 The exemplar of a complex adaptive system is a *network*, a  
 188 structure composed of many interconnected components or  
 189 nodes that interact locally through links with their neigh-  
 190 bors. Through time these networks evolve by events such  
 191 as the rewiring between nodes or the addition of new nodes  
 192 and links. In recent years there has been renewed interest in  
 193 the properties of evolving networks such as the *small-world*  
 194 *property*. In most networks the longest path between two  
 195 nodes increases proportional to the number of nodes in the  
 196 network. But at least since Milgram's Small World  
 197 experiment (Travers and Milgram 1969) scholars have  
 198 been aware of the fact that in some networks the distance  
 199 between any two nodes is surprisingly small. The small  
 200 world property has the advantage of very efficient trans-  
 201 mission of information on the network and high resilience  
 202 against errors of the network (although more vulnerable to  
 203 targeted attacks, see Albert et al. 2000). Although small  
 204 world networks are only a small set of possible networks,  
 205 the onset of Big Data has revealed that a surprisingly large  
 206 amount of actual networks exhibit this property, e.g. the  
 207 world wide web, neuronal networks, citation networks,  
 208 telephone call networks, food chains, electric power grids  
 209 and metabolite processing networks. By the late 1990's this  
 210 prompted physicists to start developing general theories  
 211 showing how possibly a network could have the small-  
 212 world property (Watts and Strogatz 1998; Barabasi and  
 213 Albert 1999). More generally, new and abundant network  
 214 data has led to a surge in the theory of evolving networks in  
 215 the last 15 years, drawing heavily on pre-existing tools  
 216 from condensed matter physics (Albert and Barabasi 2002).  
 217 It is increasingly clear that the topology and evolution  
 218 of networks is governed by robust organizing principles

(Newman 2011). They might provide exactly what Kuhn 219  
 was looking for by the end of his life: first principles that 220  
 are necessary characteristics of any developmental or 221  
 evolutionary process. 222

Insights from network theory can help articulating the 223  
 Kuhnian paradigm on all three levels. First, network theory 224  
 provides a formal framework originating from graph theory 225  
 in mathematics and condensed matter physics within which 226  
 notions such as "paradigm", "exemplar" and "incom- 227  
 mensurability" could acquire an interpretation of unprec- 228  
 edented detail. Secondly, it can suggest mechanisms for 229  
 how possibly a network of conceptual, theoretical, instru- 230  
 mental and methodological commitments could self-orga- 231  
 nize through local interaction rules into paradigms that 232  
 exhibit critical behavior. Thirdly it can help to operation- 233  
 alize Kuhnian phenomena, possibly leading to novel 234  
 empirical predictions and a clearer view on exactly what 235  
 data is relevant. For example the increasing ability to sta- 236  
 tistically identify phases and phase transitions on networks 237  
 might operationalize the notions of normal vs. revolution- 238  
 ary science and pre-paradigmatic vs. paradigmatic science. 239  
 The feasibility of this project is illustrated by the fact that 240  
 these methods have already successfully been employed by 241  
 Kiyono et al. (2006) to detect different phases in financial 242  
 data. Similar research on bibliometric data might reveal the 243  
 existence of normal and revolutionary phases in science. 244  
 This would constitute a significant philosophical result 245  
 achieved by empirical means. It also suggests a normative 246  
 agenda for a systemic philosophy of science aimed at 247  
 optimizing information flows *on* networks, increasing the 248  
 resilience *of* networks and optimizing the interconnectivity 249  
*between* networks. 250

## 251 2.2 Agent-Based Modeling

252 One reason why the study of complex adaptive systems is 252  
 relatively new is that the dynamics emerging from local- 253  
 ized interaction are often beyond reach of pure mathe- 254  
 matical methods (Bonabeau 2002). The alternative is to 255  
 explore possibility space by simulating the possible inter- 256  
 actions of the components or agents of a system in an 257  
*agent-based model*. An agent-based model is a computa- 258  
 tional model for simulating the interaction of autonomous 259  
 agents to observe the behavior of the aggregate system. 260  
 Simple agent-based models can produce surprisingly strong 261  
 results. For example Thomas Schelling (1978) demon- 262  
 strated with his exemplary checkerboard model that just a 263  
 small racial preference is already sufficient to produce 264  
 strictly segregated neighborhoods over time. Although Schelling 265  
 made his model using only paper and pencil, canvassing pos- 266  
 sibility space often requires the computation of a vast 267  
 number of possible scenarios and hence advanced agent- 268  
 based models require computational power of a size that 269



270 for long was not widely available. Moreover the develop-  
 271 ment of such models typically required substantial pro-  
 272 gramming skills. Only recently has low-barrier software  
 273 such as Netlogo enabled a broader use of these models,  
 274 accompanied by the emergence of methodological guide-  
 275 lines for their construction (Miller and Page 2007). Agent-  
 276 based modeling can be used to articulate the Kuhnian  
 277 paradigm by allowing to investigate how possibly local  
 278 interaction rules can produce Kuhnian aggregate patterns  
 279 such as “normal science”, “revolution”, “crisis”, “para-  
 280 digmatic” and “pre-paradigm” periods. If scientific activ-  
 281 ity behaves as a complex system, then science is a process.  
 282 Agent-based simulations are uniquely suited for investi-  
 283 gating not the outcomes but the process by which it was  
 284 reached.

### 285 2.3 Big Data

286 The study of complex systems is characterized by the heavy  
 287 use of statistics to study aggregate patterns emerging from  
 288 complex underlying interactions. The scarcity of large and  
 289 qualitative datasets has long been an impediment to its  
 290 expansion beyond physics. Complex systems typically  
 291 consist of a very large number of components, for example  
 292 economic agents in a market, each with their own interac-  
 293 tions through time. Only recently do we have the technical  
 294 means to acquire, store and process such information. For  
 295 example the famous small world result was obtained  
 296 counting the steps it took letters to reach a given destination,  
 297 and the final result was based on only 64 such letters. This  
 298 situation has changed dramatically with the onset of “Big  
 299 Data”, vast datasets generated as a result of the digitization  
 300 of our world. To give just one example, in 2008 Jure  
 301 Leskovec replicated Milgram’s result using the Microsoft  
 302 Messenger instant-messaging network containing 255 bil-  
 303 lion messages sent by 240 million people. Scientometric  
 304 data is part of this Big Data revolution, containing infor-  
 305 mation about for example co-authorship, keywords and  
 306 citations of scientific papers. It is a fresh and vast source of  
 307 empirical data about the dynamics of science through time.

308 Just as Milgram, Kuhn made claims about system  
 309 properties but had to content himself with anecdotal evi-  
 310 dence, in his case from historical case-studies. Neverthe-  
 311 less he initially had huge expectations about the role the  
 312 historiography of science could potentially play in his  
 313 project of an empirically better informed philosophy of  
 314 science. As noted above, this was a dead end. Kuhn seems  
 315 to have realized this fairly quickly. Already in the post-  
 316 script to SSR his hopes for identifying paradigms in  
 317 empirical data shift to statistical data about science: “for-  
 318 mal and informal communication networks including those  
 319 discovered in correspondence and in the linkages among  
 320 citations [...]. Typically it may yield communities of

perhaps one hundred members, occasionally significantly  
 fewer. Communities of this sort are the units that this book  
 has presented as the producers and validators of scientific  
 knowledge. Paradigms are something shared by the mem-  
 bers of such groups” (SSR, 178). Kuhn explicitly refer-  
 ences Eugene Garfield, the founder of the Web of Science.  
 Currently this is one of the largest scientometric databases  
 in the world but back then the whole project was still in its  
 infancy. Nevertheless Kuhn is convinced that this is the  
 way to go: “I take it that the job can and will be done”  
 (ibid.). Now more than 40 years after the postscript, this  
 data exists and is readily available for analysis.

Perhaps surprisingly, the development of bibliometric  
 databases has largely gone unnoticed for many philoso-  
 phers of science. Philosophy of science has a long history  
 of focusing on the products of science in relation to the  
 world. But citations do not have, nor are they intended to  
 have, any justificatory value; citing a paper does not mean  
 one agrees with it. Yet citations anchor a paper in a net-  
 work of similar papers. They are similar not in their  
 opinion but in the more abstract sense of sharing what the  
 question should be and what counts as a solution. Thus  
 citations are, and are intended to be, anchoring a paper in a  
 network of papers that address similar questions and  
 uphold similar standards. So while citation data is mean-  
 ingless from a justificatory point of view, for the Kuhnian  
 paradigm it captures an elementary relation: citations  
 indicate membership of the same paradigm, viz. sharing the  
 same conceptual, theoretical, instrumental and methodo-  
 logical standard. A highly cited paper indicates that many  
 other papers use it for anchoring. Such papers are exem-  
 plary. Citation networks are typically characterized by a  
 power law distribution of citations. As a consequence there  
 are “hubs” in the network, nodes with a disproportionately  
 large amount of links. This operationalizes the notion of an  
*exemplar*. Such an exemplar exemplifies the problems and  
 standards for the papers to which they are connected. This  
 cluster in turn operationalizes the notion of a *paradigm* as a  
 network of scientific practices connected by conceptual,  
 theoretical, instrumental and methodological commitments  
 (SSR, 42). Although the nodes connected to the hub tend  
 not to be connected to other hubs, the hubs themselves are  
 (see Fig. 1). This could explain Kuhn’s claim that the most  
 exemplary scientists typically contribute to multiple para-  
 digms: “Usually individual scientists, particularly the ablest,  
 will belong to several such groups either simultaneously or  
 in succession” (SSR, 178). Shifts in growth rates of con-  
 tributions to different paradigms can be used as an empirical  
 proxy for a *revolution* described as “an increasing shift  
 in the distribution of professional allegiances” (SSR, 15)  
 This might lead to the production of conclusive statistical  
 evidence about the existence of Kuhnian scientific  
 revolutions.



**Fig. 1** An example of a network with 500 nodes with links distributed as a power law

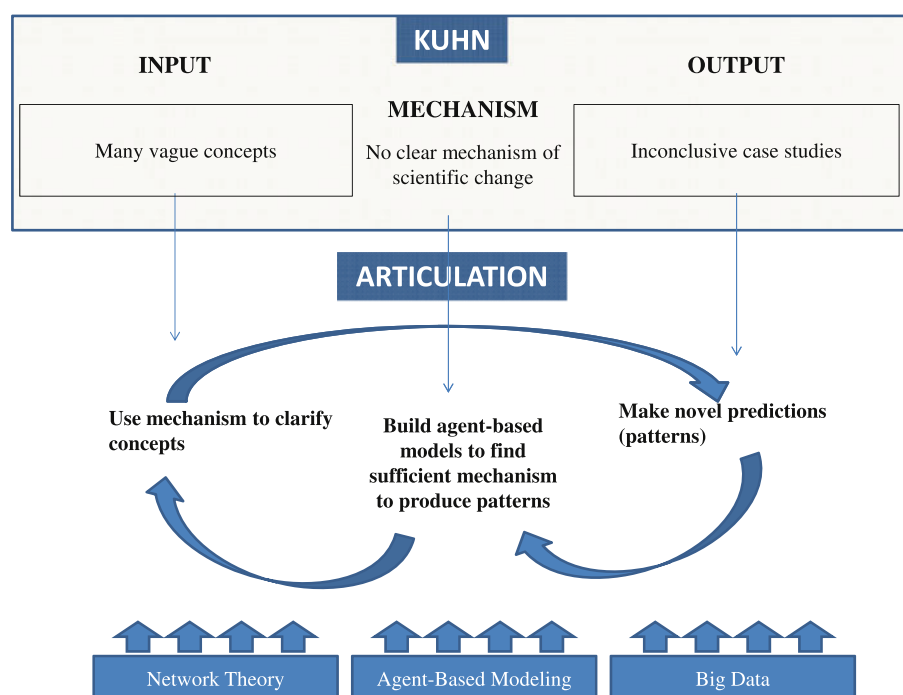
374 The complex, evolving networks revealed in bibliometric data (co-author networks, citation networks, key-  
 375 word networks,...) are the material reflection of science as a complex adaptive system. Because the Kuhnian paradigm  
 376 can give meaning to a citation, it is able to incorporate this fresh and vast source of empirical data. It is one step closer  
 377 to Kuhn's dream of a descriptively more adequate normative philosophy of science (Fig. 2).  
 378  
 379  
 380  
 381

### 382 3 Toward a Systemic Philosophy of Science

383 Kuhn is often credited with undermining the logical  
 384 empiricist research consensus in his time. But because of  
 385 its lack of articulation he failed to install a new one. In line  
 386 with the spirit of the "new image of science" that Kuhn put  
 387 forward, philosophers of science after SSR started to take  
 388 more serious actual scientific practice. But the lack of  
 389 generality of discipline-specific case-studies has led to a  
 390 fragmentation of the discipline into philosophies of indi-  
 391 vidual sciences. Kuhn himself later came to regret  
 392 this emphasis on case-studies, calling them "misleading"  
 393 because specific cases do not provide a basis for the  
 394 extrapolation of general normative guidance, rather they  
 395 illustrate their sheer variation. This variation has played a  
 396 central role in philosophy of science from its inception,  
 397 when Reichenbach and Carnap were struggling with the

398 construction of alternative geometries. Too much variation  
 399 threatens philosophy of science's ability to make general  
 400 claims about science. And as for making normative rec-  
 401 ommendations for specific situations, dependent on thor-  
 402 ough knowledge of an often partly tacit context, there is no  
 403 reason to assume that philosophers have privileged access  
 404 to this context over and above the scientists concerned.  
 405 Reichenbach proposed that philosophers should refrain  
 406 from making definite proposals but rather construct a map  
 407 of possibilities that can be used by scientists who can,  
 408 *given* their particular context, use it to find out what their  
 409 commitments are. "It is a kind of logical signpost which  
 410 we erect; for each path we give its direction together with  
 411 all connected directions and leave the decision as to  
 412 his route to the wanderer in the forest of knowledge"  
 413 (Reichenbach 1938, 14). Different frameworks serve different  
 414 purposes, and given a particular scientist's purpose this  
 415 map will help them find the optimal framework. Just as  
 416 biologists confronted with the diversity of species had done  
 417 before Darwin: they explained away variation using a fixed  
 418 set of purposes. But where do these purposes come from?  
 419 Philosophers' self-imposed restriction to the context of  
 420 justification effectively put this question outside of phi-  
 421 losophy of science. As far as philosophers were concerned,  
 422 the purposes were *given*, just as biologists before Darwin  
 423 had assumed as given the purposes species serve. But like  
 424 Darwin, Kuhn had realized from his empirical observations  
 425 the large amount of variation in these purposes. So instead  
 426 of using the purposes as an explanation for the variety of  
 427 frameworks, he made that variety of purposes itself an  
 428 explanandum. Although there is no general story to be told  
 429 about particular purposes, he realized that there might still  
 430 be a general pattern to be found in the process of their  
 431 change: "many of the most central conclusions we drew  
 432 from the historical record can be derived instead from first  
 433 principles" (Kuhn 2000, 112), and these principles "are  
 434 necessary characteristics of any developmental or evolu-  
 435 tionary process" (ibid. 119). Kuhn's intervention can hence  
 436 be understood as a Darwinian one: from different frame-  
 437 works for different purposes to a realization that variation  
 438 is the norm and general results can be found only in their  
 439 patterns of change. Hence focus shifts from the products of  
 440 science to its process. The components of a complex sys-  
 441 tem cannot be understood exhaustively without taking into  
 442 account their past and their relation to each other. A phi-  
 443 losophy seeing science as a complex system therefore  
 444 makes descriptive and normative claims about a new level  
 445 of analysis, the *systemic* level. Even though traditional  
 446 philosophy of science is increasingly fragmented because  
 447 of the lack of generality of discipline-specific case-studies,  
 448 a *systemic* philosophy of science might reveal that there is  
 449 indeed an across-the-board story to be told about which  
 450 philosophers of science can claim exclusive expertise.

**Fig. 2** A better articulation of the Kuhnian paradigm is a necessary condition for its operationalization



451 In the previous section I have sketched a number of  
 452 substantial developments in the tools that can be used to  
 453 articulate the Kuhnian paradigm. But the articulation of the  
 454 Kuhnian paradigm should not be an end in itself. Rather the  
 455 goal is to operationalize it: to make it constitutive of a  
 456 research agenda for the philosophical investigation of science  
 457 including both a descriptive and a normative component; it has to *describe* what science is, and *explain* why  
 458 it works.  
 459

### 460 3.1 Descriptive: Science as a Process

461 Describing science as a complex system means science is  
 462 essentially a process rather than just the sum of its parts.  
 463 Traditional philosophy of science takes the finished products  
 464 of science as a starting point and considers their change  
 465 only in response to external factors. A Kuhnian philosopher  
 466 of science takes a step back and wonders how possibly an  
 467 even remotely coherent and successful body of knowledge  
 468 can emerge from the localized interactions of individual  
 469 scientists through time. This allowed Kuhn to thematize  
 470 systemic, dynamic phenomena such as schools, paradigms  
 471 and revolutions, phenomena emerging from the interactions  
 472 of their components and not reducible to those components.  
 473 Understanding these phenomena normally not considered  
 474 part of the domain of philosophy of science requires  
 475 understanding science as a *process* rather than as the sum  
 476 of its products because they cease to exist when their  
 477 components stop interacting and their current structure  
 478 can only be understood as the result of the localized

479 interactions of its components in the face of chance events  
 480 through time. This is why Kuhn considers the historical,  
 481 the social and the contingent to be integral parts of the  
 482 domain of philosophy of science. Without them science  
 483 could not be seen as a process. The benefits of describing  
 484 science as a complex system are then twofold. First it  
 485 extends the domain to a range of novel phenomena at the  
 486 systemic level such as schools, paradigms and revolutions,  
 487 and second it can do so without putting excessive demands  
 488 on the individual agents in terms of the amount of information  
 489 they can process and the amount of oversight they have.  
 490

491 The main descriptive challenge of the Kuhnian paradigm  
 492 is then to describe these systemic features such as schools,  
 493 paradigms, revolutions, disciplines and the power-law  
 494 distribution of citations as emergent phenomena. If the domain  
 495 of traditional philosophy of science is the question how  
 496 science reacts to exogenous chance events such as new  
 497 observations resulting from technological advances, the domain  
 498 of Kuhnian philosophy of science is to investigate to what  
 499 extent these are reinforced by and simultaneously change  
 500 the self-organized structure of science. Why do some  
 501 discrepancies become anomalies and others don't? Why  
 502 do some solutions become exemplars and others not? Why  
 503 do disciplines form? Why does disciplinary diversity  
 504 persist? To what extent can these boundaries be explained  
 505 by external properties of the subject matter, and to what  
 506 extent are they self-organized? How can there be rational  
 507 disagreement within disciplines? How possibly could  
 508 something like a "paradigm" or a



509 “revolution” occur? Both the suitability and the timeliness  
510 of using the above tools for these purposes is evidenced by  
511 the fact that physicists themselves have started using agent-  
512 based models based on network theory to model the emer-  
513 gence and decline of Kuhnian paradigms (Bornholdt et al.  
514 2011).

515 Kuhn’s description was flawed by a lack of conceptual  
516 clarity, the lack of an explicit mechanism of change and the  
517 inconclusiveness of the adduced historical evidence. The  
518 new developments in the previous section now provide the  
519 means to overcome these descriptive flaws. Network theory  
520 offers an overview of a wide range of phenomena that can  
521 emerge on networks, along with explanations of how they  
522 can possibly emerge. Firmly rooted in mathematics and  
523 condensed-matter physics, it provides a precise and pow-  
524 erful conceptual framework for description. The resulting  
525 conceptual clarity is a necessary condition for program-  
526 ming agent-based models. An agent-based model is a  
527 computational model for simulating the interaction of  
528 autonomous agents to observe the behavior of the aggre-  
529 gate system. They can be used to provide how-possibly  
530 explanations for emergent structures. By canvassing pos-  
531 sibility space, testable predictions can be made. The  
532 resulting hypotheses can then be tested against the new  
533 and vast set of empirical data available in scientometric  
534 datasets.

### 535 3.2 Normative

536 The shift in perspective that is brought about by seeing  
537 science as a complex system is that traditional philosoph-  
538 ical problems are injected with the dimensions of interac-  
539 tion and time. A system with multiple agents allows them  
540 to be different, divide labor over them and let the sum of  
541 their labor be more than its parts. With the introduction of  
542 time comes a conflict between the past and the future: a  
543 choice can be designed to optimize on current knowledge,  
544 but can also be aimed at increasing the amount of knowl-  
545 edge in the future. This conflict between exploitation and  
546 exploration was thematized by Kuhn as the “Essential  
547 Tension” (1977). It is a direct consequence of a process-  
548 view on science and the key to understanding the philo-  
549 sophical implications of Kuhn’s perspective on science.  
550 Restricted to the finished products of science, the norma-  
551 tive evaluation of scientists’ choices can be restricted to a  
552 static evaluation of the decision given the available infor-  
553 mation. However, if science is an ongoing process, scien-  
554 tists might choose to compromise on exploiting *existing*  
555 knowledge in favor of finding more knowledge in the  
556 *future*.

557 Rationality: In a situation that requires both exploration  
558 and exploitation, individual scientists are effectively faced

559 with a so-called “multi-armed bandit” problem, after the  
560 model where a gambler enters a hall full of slot machines  
561 each with an unknown payoff matrix and has to trade off  
562 playing the same slot machine to get better information  
563 about its payoffs against exploring other slot machines that  
564 might have a more favorable payoff matrix. Theory *choice*  
565 becomes theory *search*. Scientists not only must choose the  
566 best current theory but also the one that is most likely to  
567 lead them to better theories in the future. For Kuhn, sci-  
568 entific rationality is not about adopting what is currently  
569 best, but about “the fittest way to practice future science”  
570 (SSR, 172). Scientific rationality goes from being back-  
571 ward-looking in the case of the evaluation of “finished”  
572 products, to being both backward-looking (exploitation)  
573 and forward-looking (exploration) in the case of practicing  
574 future science. Finding a dynamic balance between  
575 exploration and exploitation is a fundamental normative  
576 challenge for a systemic philosophy of science. Mayo-  
577 Wilson et al. (2012) use multi-armed bandit models to  
578 analyze theory choice.

579 Virtues: Because of the different nature of exploitation  
580 and exploration, it is natural to assume there are different  
581 (sets of) potentially conflicting virtues governing theory  
582 choice depending on whether one aims more at exploration  
583 or more at exploitation. For example the virtues of sim-  
584 plicity and generality are often contested, but then again  
585 they seem to play a crucial role in the exploration of new  
586 frameworks; see for example Einstein (1934) for a defence  
587 of the virtue of simplicity. This suggests they are explor-  
588 ative virtues, and their contestation can be explained by the  
589 fact that traditional philosophers only have room for  
590 exploitative virtues. The conflict between these virtues and  
591 how to deal with them (e.g. resulting in different strategies  
592 for model-building) is the subject of an old but still  
593 growing literature (Levins 1966; Orzack and Sober 1993;  
594 Matthewson and Michael 2009).

595 Division of Labor: If theory search involves both an  
596 exploitative and an explorative dimension and different  
597 (sets of) virtues are associated with it, it can be expected  
598 that some scientists are better at exploring and others at  
599 exploiting. Hence it is rational to divide labor over  
600 explorers and exploiters, or as is more common in the lit-  
601 erature on the division of labor in science, between mav-  
602 ericks and followers (e.g. Kitcher 1990). A normative  
603 challenge that is still largely open is what an optimal mix  
604 of both would be, and if there is even a general answer to  
605 this question. Weisberg and Muldoon (2009) provide an  
606 exemplary treatment of this challenge. Kuhn moreover  
607 described how the aggregation of these individual actions  
608 leads to a striking pattern of balancing exploitation and  
609 exploration at the aggregate level, whereby entire disci-  
610 plines can go into an exploitative mode during “normal



611 science” and explore new frameworks during “revolu-  
612 tionary science”. Although this is a self-organized pattern,  
613 a normative challenge is to find out whether this is the most  
614 rational way to collectively trade off exploration and  
615 exploitation in a scientific community. An additional  
616 question is how such self-organization compares to  
617 so-called “pre-paradigmatic” situations where self-organization  
618 does not succeed.

619 Independence: In a complex system, individual-level  
620 properties do not necessarily carry over to the systemic  
621 level. Hence collective rationality does not require per-  
622 fectly rational individuals, but individual rationality does  
623 not guarantee systemic rationality. This property is coined  
624 the “independence thesis” by Mayo-Wilson et al. (2011).  
625 The future-orientedness of theory choice makes perfect  
626 rationality in principle impossible because of the funda-  
627 mental uncertainty (not just risk) associated with estimat-  
628 ing the fruitfulness of a framework that still needs to be  
629 developed. This uncertainty is evidenced in the multi-  
630 armed bandit problem, where in most cases there is no  
631 unique strategy that can be proven to be optimal.  
632 Remarkably simple *rules of thumb* often outperform more  
633 complex strategies in these situations. It becomes then an  
634 important normative challenge for philosophers of science  
635 to determine what these rules of thumb should be. Kuhn  
636 himself admitted that he did not see how he could articulate  
637 how scientists following simple rules of thumb could  
638 together produce successful science.<sup>2</sup> The work of Giger-  
639 renzer et al. (2000) on simple heuristics that make us smart  
640 breaks new ground in this respect.

641 Progress: Complex systems need not have an endpoint  
642 and when they have it need not be optimal, making tradi-  
643 tional linear notions of scientific progress inapplicable.  
644 Moreover as a result of dynamic interaction rules scientific  
645 practice changes the very rules by which it operates,  
646 leaving no independent yardstick to measure progress with.  
647 This problem is common to all complex adaptive systems;  
648 as Kuhn notes in the final pages of SSR, also for biological  
649 evolution. De Langhe (forthcoming) represents the prob-  
650 lem of progress in a dynamic environment within the epi-  
651 stemic landscapes framework introduced by Weisberg and  
652 Muldoon (2009) as the discovery of a “dancing” land-  
653 scape, drawing on existing formalism from models of  
654 dancing fitness landscapes in biology such as Bak and  
655 Sneppen (1993).

2FL01 <sup>2</sup> “Even those who have followed me this far will want to know how  
2FL02 a value-based enterprise of the sort I have described can develop as a  
2FL03 science does, repeatedly producing powerful new techniques for  
2FL04 prediction and control. To that question, unfortunately, I have no  
2FL05 answer at all [...] The lacuna is one I feel acutely” (Kuhn 1977,  
2FL06 332–333).

## 4 Conclusion

In recent decades there have been a number of major  
developments which should have a serious impact on our  
philosophical understanding of science. Vast new empiri-  
cal, computational, and theoretical resources have become  
readily available. In this paper I have argued that Thomas  
Kuhn can offer a framework within which the philosophy  
of science can exploit these new resources. The examples  
that were given in the previous section illustrate that the  
exploitation of these new resources has already started, but  
in a fragmented way. The Kuhnian paradigm unifies these  
novel contributions under the heading of a systemic phi-  
losophy of science, of which they are exemplars. I strongly  
believe they are pointing in the direction of the fittest way  
to practice future philosophy of science.

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