

An agent-based model of interdisciplinary interactions in science

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2 ABSTRACT

3 An increased interdisciplinarity in science projects has been highlighted as crucial to tackle
4 complex real-world challenges, but also as beneficial for the development of disciplines
5 themselves. This paper introduces a parcimonious agent-based model of interdisciplinary
6 relationships in collective enterprises of knowledge discovery, to investigate the impact of scientist-
7 level decisions and preferences on global interdisciplinarity patterns. Under the assumption
8 of simple rules for individual researcher project management, such as trade-offs between
9 invested time overhead and knowledge benefit, model simulations show that individual choices
10 influence the distribution of compromise points between emergent level of disciplinary depth and
11 interdisciplinarity in a non-linear way. Different structures for collaboration networks also yield
12 various outcomes in terms of global interdisciplinarity. We conclude that independently of the
13 research field, the organization of research, and more particularly the local balancing between
14 vertical and horizontal research, already influences the final positioning of research results and
15 the extent of the knowledge front. This suggests direct applications to research policies with a
16 bottom-up leverage on the interactions between disciplines.

17 **Keywords:**

1 INTRODUCTION

18 [1] diversity/innovation

19 [2]: applied persp?

20 [3]

21 [4] mainsream / interdisc papers

22 [5] abm structure knowledge coevol

23 [6] assymetry of interdisc between social and hard sciences in the credit given to other disciplines

24 [7] competition for funding is inefficient

25 [8] importance of teams vs solo authors

- 26 [9] individual interdisciplinarity: in time, more scientists switch between topics
27 [10] : model for oa and research quality
28 [11] : empirical evidence of an optimal intermediate level of interdisciplinarity for impact
29 [12] paradigms persist in “low-power” sciences

30 1.1 Agent-based models of scientific knowledge production

- 31 [13] use an agent-based model to simulate the impact of a workflow to process data under different
32 collaboration scenarios.
33 [14] simulate citation dynamics, and more particularly the consequence of introducing a performance
34 index on citation patterns.
35 [15] surveys 46 simulation studies of peer review with numerous applications [16]

36 1.2 Perspectivism and Model Coupling

37 Beyond the simplifying opposition between fully constructivist and realistic approaches to science,
38 several alternatives have been developed, among which Perspectivism [17] is a way to tackle most of the
39 issues opposing these two by taking an agent-based approach to the production of scientific knowledge.
40 The principal feature of this point of view is to consider each scientific enterprise as a single perspective,
41 in which an agent aims to understand an aspect of the real world (the ontology) with the mean of a
42 medium, that is considered as a model. Constituted disciplines thus contains more or less compatible
43 perspectives. The explicitation of this approach has been done by [18] to embed it into knowledge domains,
44 as a generalization of knowledge domains introduced by [19].

45 We postulate that this approach to science may be a powerful tool to foster interdisciplinary collaborations,
46 if used in a reflexive way in the construction of projects. More precisely, an “Applied Perspectivism” would
47 imply a high-level of reflexivity for each agent implied, a mapping of the different layers of the enterprise
48 and the positioning of each agent regarding the domains of knowledge. This way, in the particular case of
49 model coupling, the explicitation of positioning and of the structure of each knowledge implied should
50 ease interactions. As Banos points out [20], transversal work must alternate with deeper investigations
51 in each discipline, in a kind of “virtuous circle”. [21] synergy more than interdisciplinarity in itself [22]
52 This raises the issue of, before individual researcher particularities, how a given collective structure of
53 scientific knowledge production should balance between these two. It is clear that this question is deeply
54 endogenous to each studied subject, and even each particular approach taken, but within the applied
55 knowledge framework described above, we have reasons to believe that certain structural properties may
56 be rather general. Indeed, each discipline is expected to bring components for each knowledge domain, and
57 the co-evolving perspective is built on their interrelations. We propose to investigate here basic aspects of
58 this issue, with means of agent-based modeling.

59 We aim at providing quantitative evidence of the feasibility of the epistemological point of view described
60 above and inform potential implementation for some of its processes, more precisely how can certain level
61 of coupling of perspectives (or overlap of ontologies) may be achieved given specializations of scientists
62 and a given dynamic of interaction.

1.3 Contribution

Various works have dealt with microscopic modeling of knowledge production, among which for example Chavalarias' Nobel game [23] which investigates the balance between falsification of previous theories and the elaboration of new theories. Giere also introduced an agent-based model of science in [24], consistently with the perspectivist approach described above. We develop here a simple agent-based model of scientific research focusing on the interplay between disciplinary and interdisciplinary research. The rationale relies on the basic assumption that scientists can choose when starting a new project between interdisciplinary collaboration and a work within their discipline. How can the choice patterns at the micro-level influence the overall interdisciplinarity level? The model is voluntary parcimonious to test if even many simplification some structural effects still hold.

[25] abm for group orga

[26] role of randomness in individual trajectories success

[27] model for research success

[28] research organization: public-private, small-large

2 AN AGENT-BASED MODEL OF INTERDISCIPLINARITY

Agents are N scientists A_i , characterized by a probability distribution $d(x)$ representing their disciplinary positioning in an abstract way: research is summarized by a one dimensional variable \mathbb{R} , and the disciplinary positioning on this axis is given by the distribution. The model is setup with normal distributions of width σ with an average distributed uniformly in $[0; 1]$. Scientists also have a time budget per day, that we will summarize as a future timetable $T(t_0) : t > t_0 \mapsto p(t) \in \mathcal{P}$ where \mathcal{P} is the space of scientific projects. The central feature of the model is the utility function $U(d_i, d_j)$ determining an abstract utility for scientist i to collaborate with j for a given project. It will be a function of the disciplinary overlap $o = \int_x d_i(x) \cdot d_j(x) dx$ and different assumptions on the form of this cost function can be tested. We take a linear cost in the overlap and a varying benefit, expressing the fact that researchers have different strategies regarding their interdisciplinary positioning. This way, we have $U(d_i, d_j) = o/i^\alpha - o$, assuming a fat-tail distribution of individual preferences for interdisciplinarity, given by a power law of parameter α . A discrete choice formulation gives the probabilities for a scientist i to choose among j collaborators by $p_j = \exp(\beta U(d_i, d_j)) / \sum_k \exp(\beta U(d_i, d_k))$. Given a social network of relations, that we take for now as a fixed scale-free social network, the temporal evolution of the model goes as follows: (i) one scientist with no current activity is picked up at random, and starts a project with one of its potential collaborators taken as its neighbors in the network that have free time, chosen with the probability p_j . The project has a random uniform duration and timetables are updated accordingly; (ii) current projects are updated and finished if necessary. The outcome of the model is measured by average depth across project, defined for one project as the overlapping areas between distribution, and average interdisciplinarity measured by total area covered.

3 RESULTS

The model is implemented in NetLogo and explored with OpenMole [29]. Source code and results are available on the open repository of the project¹. We run a basic grid exploration of the parameter space,

¹ on github at <https://github.com/JusteRaimbault/Perspectivism>

both with random and small-world social networks, for parameters α, β, σ with 50 repetitions of the model for each parameter points, corresponding to 158,400 model runs. Figure ?? shows indicators variation on a given subspace and the corresponding Pareto front between depth and interdisciplinarity. We show a second order influence of preference hierarchy α and non-linearity of model behavior as a function of all parameters. Convergence properties are reasonable with this number of repetitions. Large individual disciplinary width σ causes the choice parameter β to have no influence, whereas low values give an increasing interdisciplinarity and a decreasing depth as a function of β . Random behavior ($\beta = 0$) leads to a constant depth of projects. When examining the Pareto front between the two contrary objectives, the optimal points occur for intermediate β when σ is fixed, suggesting non-trivial behavioral optima at a fixed disciplinary configuration. These first exploration show the complex dynamics of interdisciplinarity even with simple interaction rules and network structure, and suggests further applications such as the exploration of policies by changing network structure or studying in a more refined way the influence of α .

4 DISCUSSION

4.1 Possible extensions

Possible refinements of the model, towards a less stylized and more behavioral and micro-based model, could for example include the introduction of time budgets, simultaneous projects and dynamical time investment for scientists. The assumption of two-person projects is also strongly constraining, and relaxing it would require the extension of depth and interdisciplinarity measures that is not necessary straightforward. Furthermore, the absence of learning and of evolution of the social network when completing a project suggests a short time scale of application: further refinements should include dynamics of individual distributions and of individual relationships.

In conclusion, we show with this simple model that the individual choices produce an emerging structure of the research front, suggesting that applied perspectivism requires a careful tuning of research structure and researcher behaviors since Pareto-optimal configurations correspond to non-trivial parameter points. Future developments should include more realistic behavioral assumption, and a formalisation of the applied perspectivism approach to include it in the agent-based model.

CONFLICT OF INTEREST STATEMENT

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

AUTHOR CONTRIBUTIONS

JR designed the study, coded the model, conducted the experiments, and wrote the paper.

FUNDING

ACKNOWLEDGMENTS

DATA AVAILABILITY STATEMENT

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FIGURE CAPTIONS