

Knowledge in socio-economic systems

Insights from collaboration, mobility, and data-driven models

Dr. Giacomo Vaccario

Knowledge in our world

- Knowledge assets reflect the market value of many leading firms
- The creation of new knowledge makes the way to scientific progress
- Knowledge is increasingly an outcome of collaborative efforts:
 - 1 rise of R&D alliances
 - 2 dominance of teams in scientific production
 - 3 Big Science (LHC, LIGO, EHT)

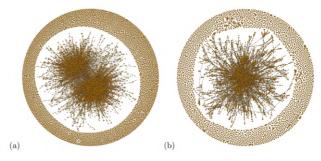
Key drivers

- \blacktriangleright social and technical capitals \rightarrow R&D alliances among firms
- \blacktriangleright scientists' expertise and "their (social) centrality" \rightarrow scientists' productivity and impact

OECD, (2013); Kuhn (1962); Wuchty et al., (2007); Hagedoorn, J. (2002); Walsh (1935); Mincer (1958); Schultz (1989); Stewart and Ruckdeschel (1998); Walker et al., (1997), Gulati and Gargiulo (1999), Stuard and Podolny (1996)

Unified vision: Collaboration networks in Industry and Science

- (a) R&D alliances and
 (b) co-authorship activities are different processes ...
- but they share similarities:
 - emergent network structures
 - ▶ knowledge is encoded: firms → patents scientists → publications
 - Key ingredients
 - Agents form networks
 - Co-produced knowledge artifacts

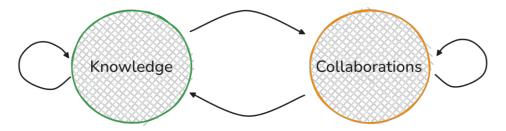






Tomasello, M. V., Vaccario, G., & Schweitzer, F. (2017). Data-driven modeling of collaboration networks: A cross-domain analysis. EPJ Data Science

Feedback loop between knowledge and collaboration

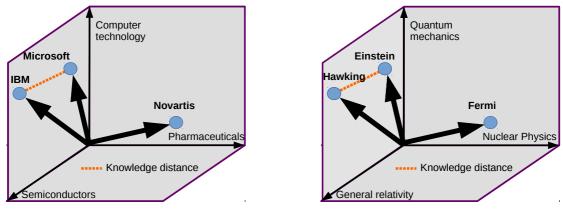


Questions

- 1 How does knowledge influence collaborations?
- 2 How do collaborations influence knowledge?
- 3 How do collaborations influence new collaborations?
- 4 How does existing knowledge influence new knowledge?

Vaccario, G. (2019). The structure, exchange, and transfer of knowledge in socio-technical systems. PhD Thesis

Embedding Firms and scientists in a knowledge space

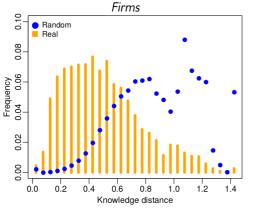


- ▶ 14K firms & 6 Million patents
- filed in 35 ISI-OST-INPI fields

- 100K scientists, 300K publications
- classified with 67 PACS codes

Vaccario, G., Tomasello, M. V., Tessone, C. J., & Schweitzer, F. (2018). Quantifying knowledge exchange in R&D networks. *Journal of Evolutionary Economics* Vaccario, G. (2019). The structure, exchange, and transfer of knowledge in socio-technical systems. *PhD Thesis*

Knowledge influences collaborations differently across domains

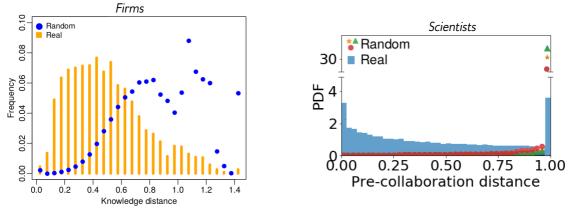


Firms prefer collaborators around an optimal knowledge distance

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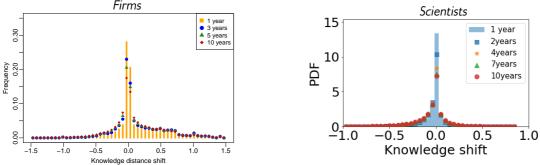


- Firms prefer collaborators around an optimal knowledge distance
- ▶ Scientists prefer collaborators with either very similar or very different knowledge

Vaccario, G., Tomasello, M. V., Tessone, C. J., & Schweitzer, F. (2018). Quantifying knowledge exchange in R&D networks. *Journal of Evolutionary Economics* Vaccario, G. (2019). The structure, exchange, and transfer of knowledge in socio-technical systems. *PhD Thesis*

Collaborations influence knowledge in unexpected ways

- Knowledge shifts = knowledge distance after collaboration knowledge distance before
 - $\blacktriangleright \ \ \textbf{Positive} \ \ values \rightarrow collaborators \ have \ more \ \ \textbf{different} \ knowledge$
 - ▶ Negative values → collaborators have more similar knowledge



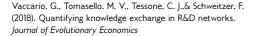
- $\blacktriangleright \quad \text{More negative values} \rightarrow \text{Majority of collaborators share more similar knowledge}$
- However, still many collaborations lead to more different knowledge

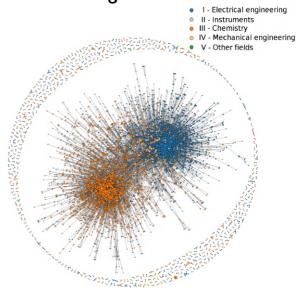
Tomasello, M. V., Tessone, C. J., & Schweitzer, F. (2015). The effect of R&D collaborations on firms' technological positions. *IFKAD*. Vaccario, G. (2019). The structure, exchange, and transfer of knowledge in socio-technical systems. *PhD Thesis*

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How to model the interplay between knowledge and collaborations

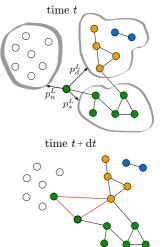
- Data-driven agent-based model to reproduce co-evolution
 - formation of the R&D network
 - knowledge exchange
- Network perspective
 - firms \rightarrow nodes
 - collaborations \rightarrow links





Old collaborations influence new ones: Label propagation model

- Existing collaborations define
 - ▶ firm's social capital
 - and influence new collaborations
- Firms $\rightarrow N$ agents
 - $\blacktriangleright \quad \text{Firm's social capital} \rightarrow \text{label}$
 - $\blacktriangleright \quad \text{new collaboration} \rightarrow \text{link formation}$
 - link formation probabilities depend on labels
 - ▶ with same label: *p*^{*L*}_{*s*}
 - with different label: p_d^L
 - with no label: p_n^L



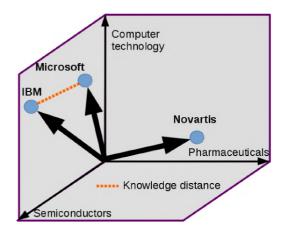
Tomasello, M. V., Perra, N., Tessone, C. J., Karsai, M., & Schweitzer, F. (2014). The Role of Endogenous and Exogenous Mechanisms in the Formation of R&D Networks. Scientific Reports

Tomasello, M. V., Burkholz, R., & Schweitzer, F. (2017). Modeling the formation of R&D alliances: An agent-based model with empirical validation. *Economics* Chair of Ecosystem Management | ecology.ethz.ch Dr. Giacomo Vaccario 31.10.2024 | 9 / 18

Knowledge distances decrease: Modelling knowledge exchange

- knowledge position
 - patent portfolio
- Collaborating firms exchange knowledge
- \Rightarrow approach in the knowledge space

$$\dot{\mathbf{x}}_i(t) = \mu \sum_{j \in \mathcal{N}_i(t)} \left[\mathbf{x}_j(t) - \mathbf{x}_i(t)
ight]$$

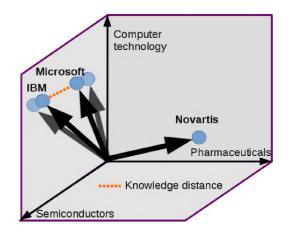


Tomasello, M. V., Tessone, C. J., & Schweitzer, F. (2016). A model of dynamic rewiring and knowledge exchange in R&D networks. Advances in Complex Systems Vaccario, G., Tomasello, M. V., Tessone, C. J., & Schweitzer, F. (2018). Quantifying knowledge exchange in R&D networks. Journal of Evolutionary Economics.

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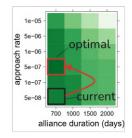
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Reproducing, learning, and proposing interventions



empirical

simulated



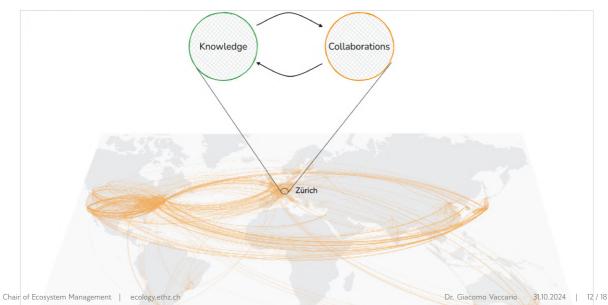
what-if analysis

Data-driven agent-based models

- Reproduces the empirical network and positive and negative knowledge shifts
- Learns about the current state using interpreting parameters values
- $\blacktriangleright \quad \text{What-if analysis} \rightarrow \textbf{Proposes} \text{ interventions to improve current state}$

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Collaborations and knowedge in the geographic space



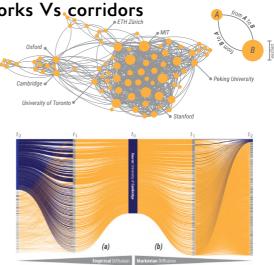
Collaborations and knowedge in the geographic space



Verginer, L., & Riccaboni, M. (2020). Cities and countries in the global scientist mobility network. *Applied Network Science* Verginer, L., & Riccaboni, M. (2021). Talent Goes to Global Cities: The World Network of Scientists' Mobility. *Research Policy* Chair of Ecosystem Management | ecology.ethz.ch Dr. Giacom

How tacit knowledge moves: Networks Vs corridors

- $\blacktriangleright \ \ Scientists \rightarrow Tacit \ knowledge$
- Networks
 - Freely move between locations
 - From one starting node career trajectories go to every other nodes
 - Only true at **city** level
- Corridors
 - Specific paths connect locations
 - From one starting node career trajectories go to a few other nodes
 - At institution and country level



Scholtes, I. (2017). When is a Network a Network? Multi-Order Graphical Model Selection in Pathways and Temporal Networks. *SIGKDD* Vaccario, G., Verginer, L., & Schweitzer, F. (2020). The mobility network of scientists: Analyzing temporal correlations in scientific careers. *Applied Network Science*

Reproducing mobility network: Data-driven agent-based model

Data

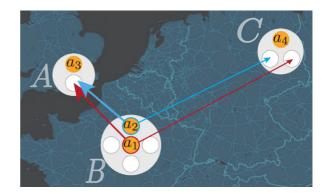
- MEDLINE: scientists' publications
- \Rightarrow Career trajectories

Model entities

- Scientists prefer closer location with higher fitness
- Locations prefer scientists with higher fitness

Constraints

- Limited space per location
- Scientists propose to move
- But locations decide to accept



Vaccario, G., Verginer, L., & Schweitzer, F. (2021). Reproducing scientists' mobility: a data-driven model. Scientific Reports

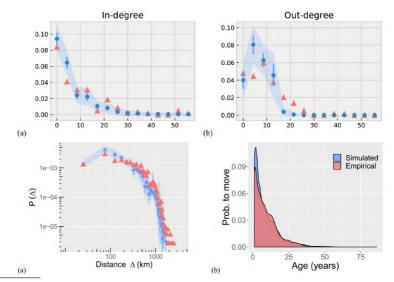
Reproducing mobility network: Data-driven agent-based model

Network level properties

- Degree distribution
- Clustering coefficient
- Path lengths
- ► ...

Scientist level properties

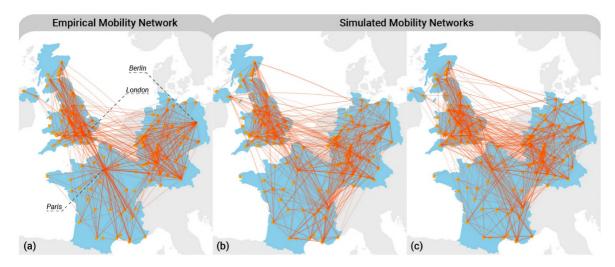
- Traveled distance
- Academic age when moving



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Reproducing mobility network: Data-driven agent-based model



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Conclusions

- Knowledge and collaborations are interdependent
- A unified vision of collaborations across industry and science is possible thanks to network models
- Knowledge is constrained in the geographic space
- Data-driven agent-based models reproduce the interdependence between knowledge and collaborations

