

Decoding the codebook of life

Eddie Lee, PhD

Data Science in Fundamental Physics

<https://eddielee.co>

June 6, 2024

Principles of Emergent Things (PoET)

What are “physics” theories for adaptive, collective, and intelligent things?

More about us at <https://eddielee.co>



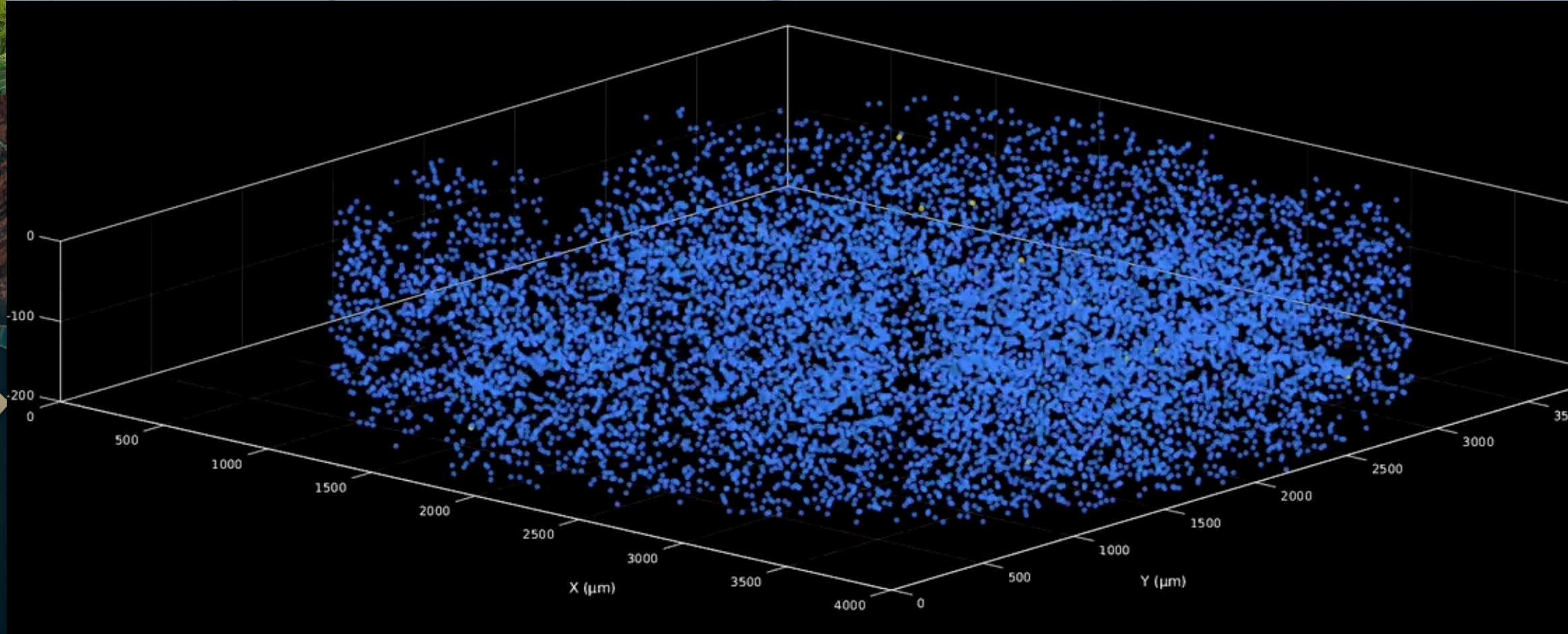
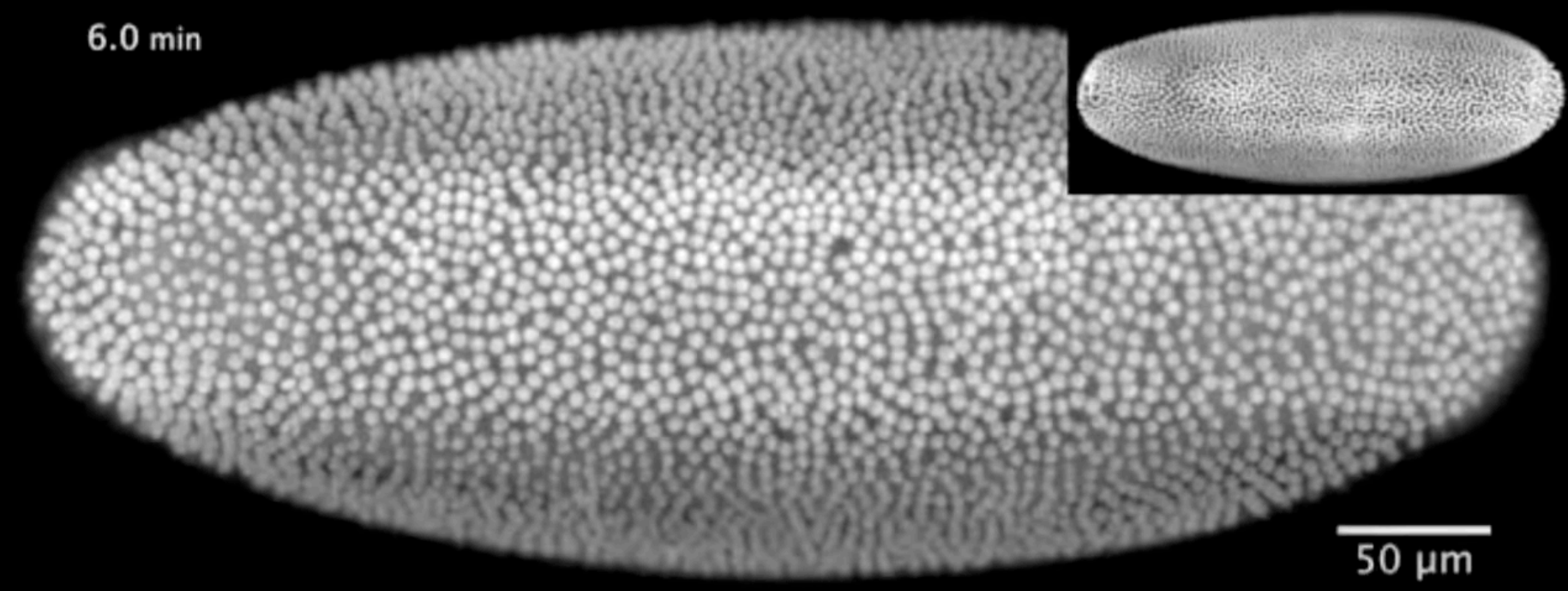
Gavin Rees
Visiting
researcher

Eddie Lee
Team leader

Ernesto Ortega
Postdoc
(2023-2024)

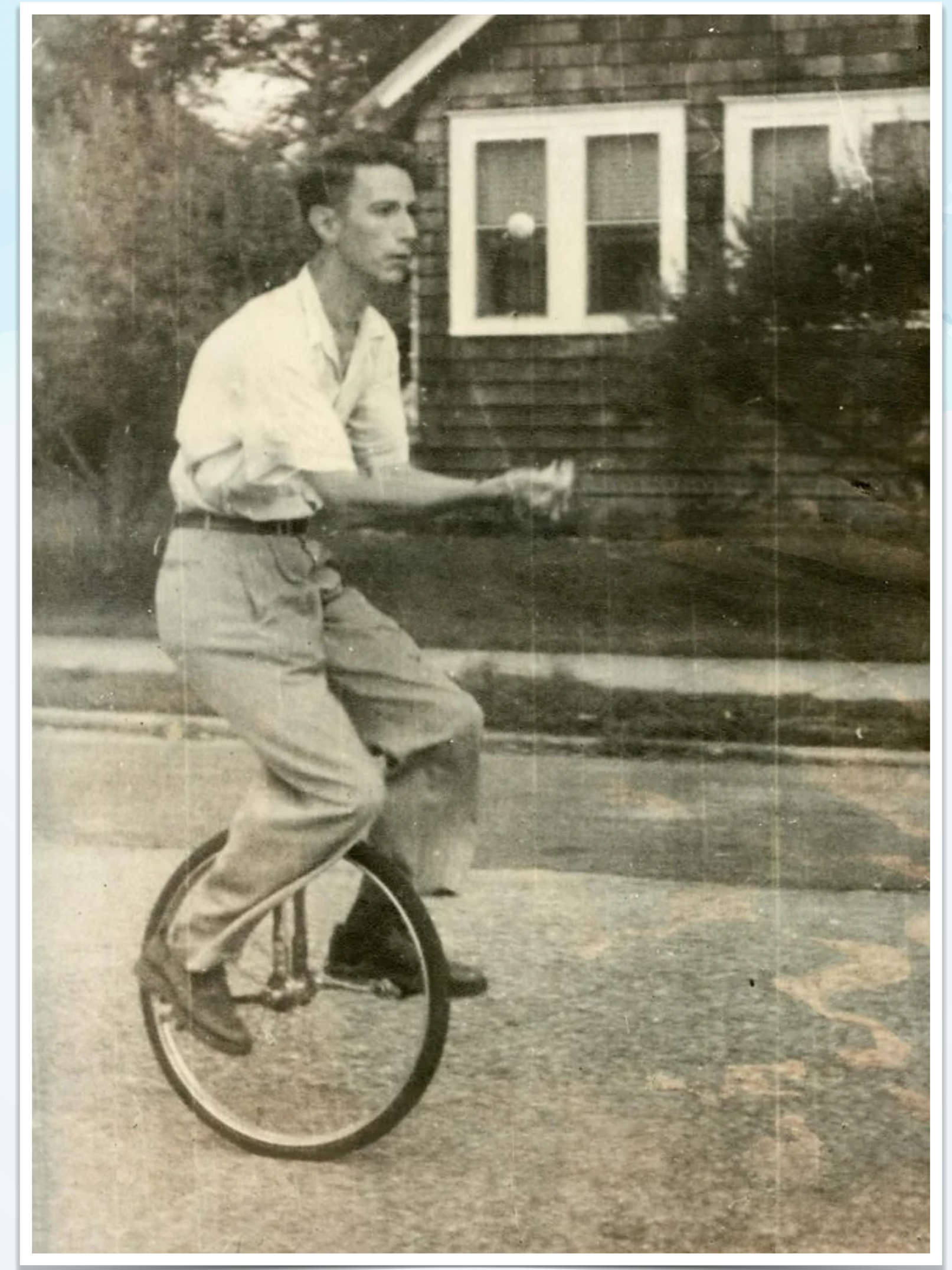
Niraj Kushwaha
PhD student

**Evelyn
McGonigle**
Research
intern
(2023)



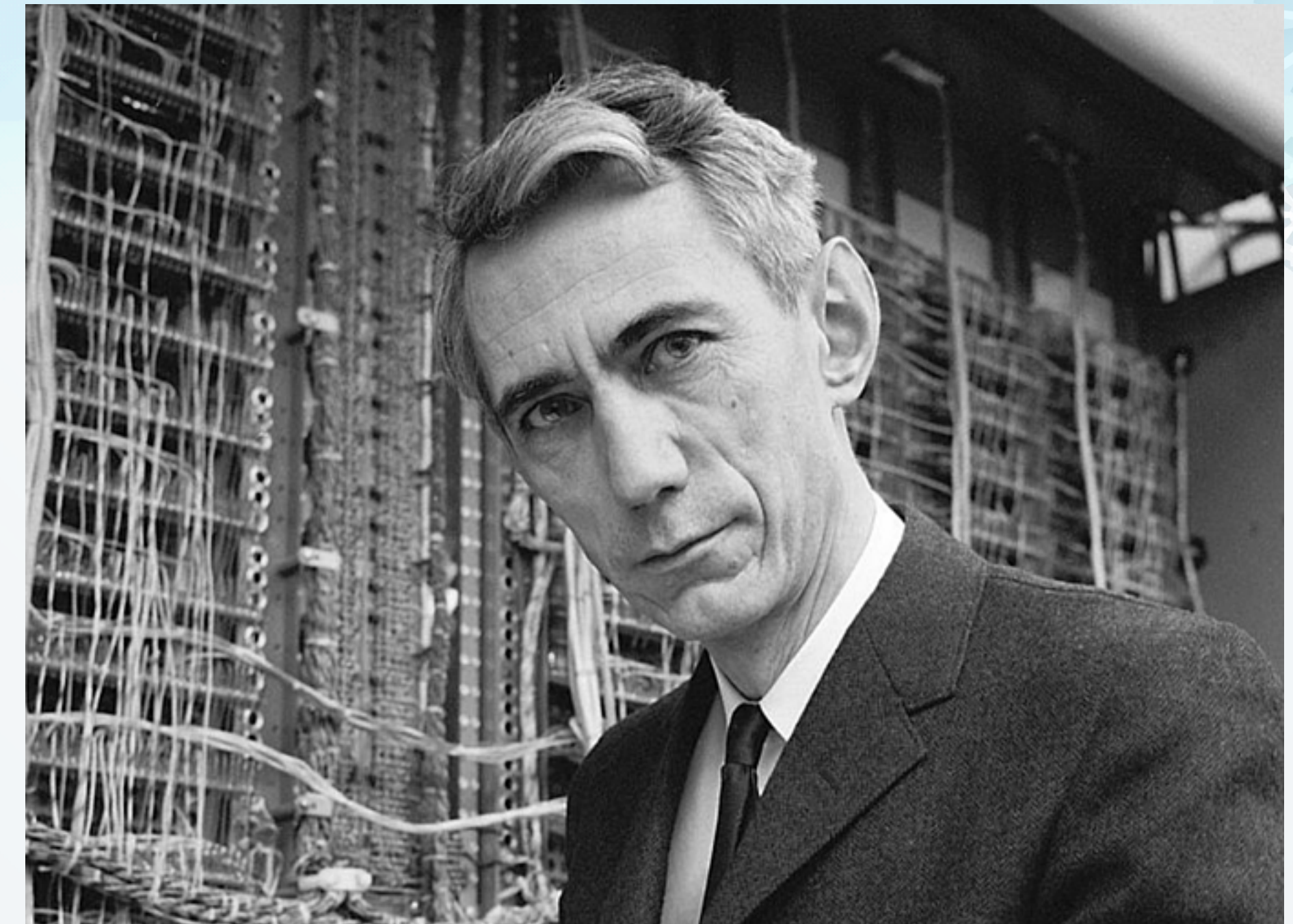
Claude Shannon & information theory

- Juggler and unicyclist
- Worked at Bell Labs
- Contributions to cryptography and artificial intelligence
- How do we characterize the information passing through a communication channel?
- A maximum (physical) rate at which information can be communicated through a channel
- Information entropy



Claude Shannon & information theory

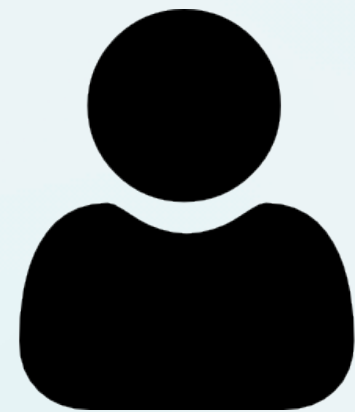
- Juggler and unicyclist
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1916-2001

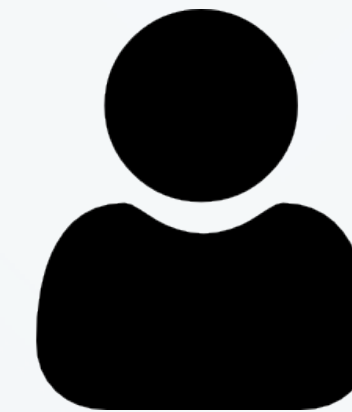
Information content of a message

Alice



message

Bob

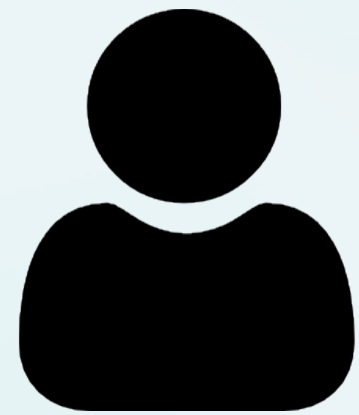


1 0 0 1 1 1 0 0 0 0 1 1 0 ...

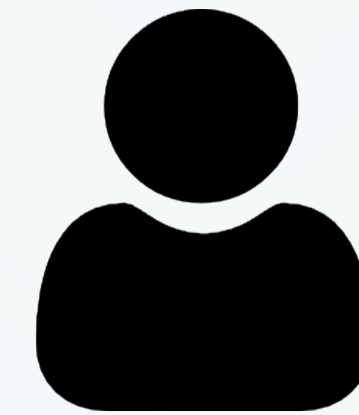


Information content of a message

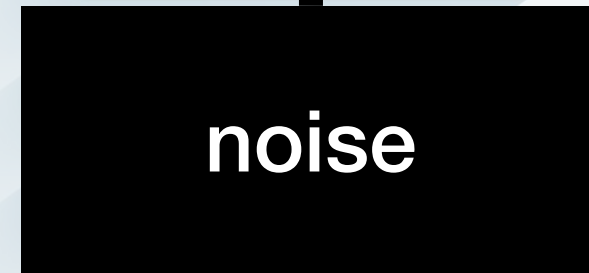
Alice

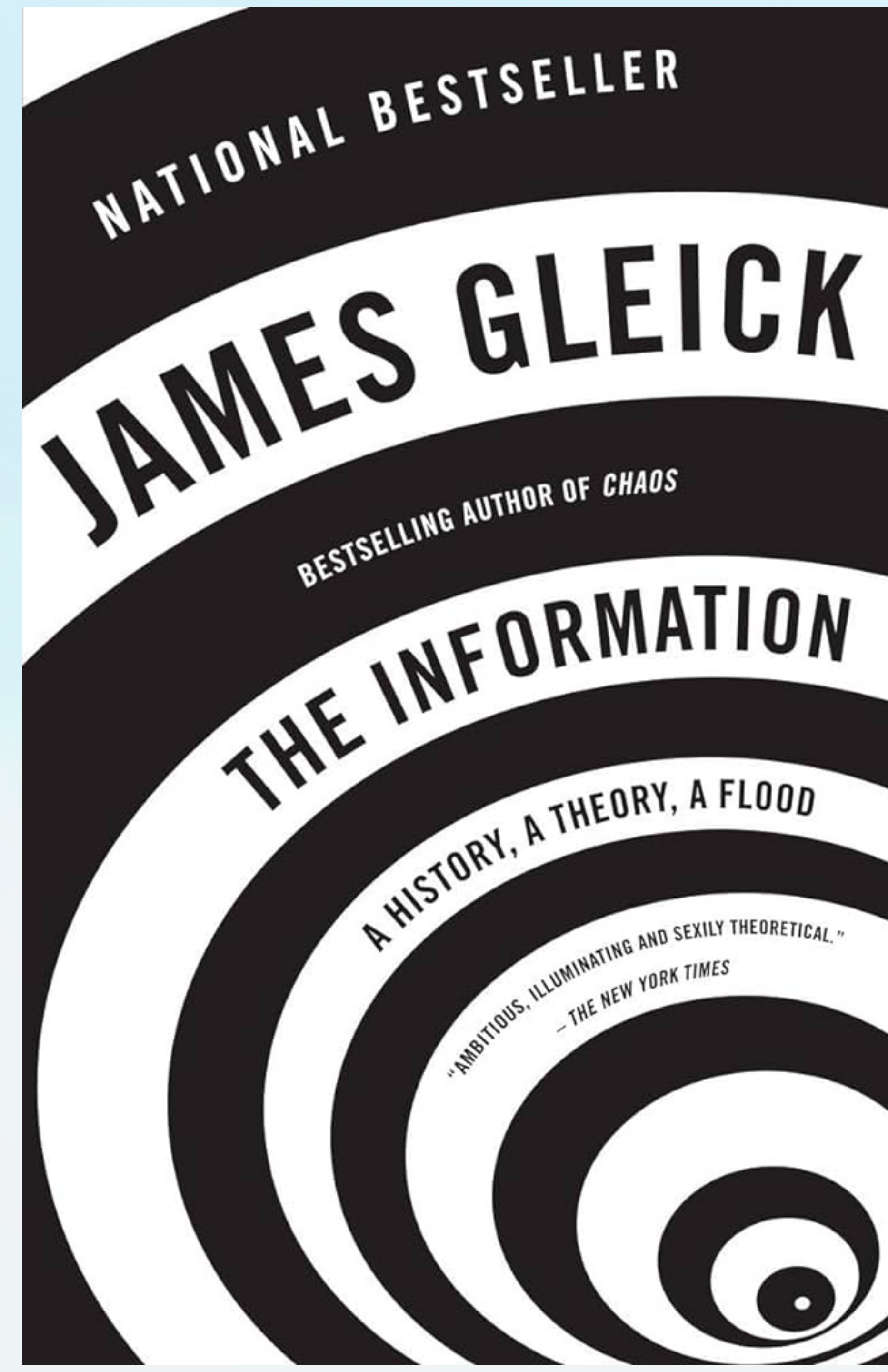


Bob



noise





NATIONAL BESTSELLER

JAMES GLEICK

BESTSELLING AUTHOR OF *CHAOS*

THE INFORMATION

A HISTORY, A THEORY, A FLOOD

"AMBITIOUS, ILLUMINATING AND SEXILY THEORETICAL."
- THE NEW YORK TIMES

4 projects at various stages of development

inference

multiscale data analysis

computation

innovation & obsolescence

Sociology is the “science of institutions, their genesis and their functioning.”

-Émile Durkheim



Lee, E. D., Broedersz, C. P., Bialek, W. "Statistical Mechanics of the US Supreme Court."

Journal of Statistical Physics, 160(2):275–301 (2015).

Lee, E. D. "Partisan Intuition Belies Strong, Institutional Consensus and Wide Zipf's Law for Voting Blocs in US Supreme Court." Journal of Statistical Physics 173(6):1722–1733 (2018).

Lee, E. D., Katz, D. M., Bommarito II, M. J., Ginsparg, P. H. Sensitivity of collective outcomes identifies pivotal components. Journal of The Royal Society Interface 17, (2020).

Lee, E. D. & Cantwell, G. T. Valence and interactions in judicial voting. Philosophical Transactions of the Royal Society A 20230140 (2024).

Rees, G. & Lee, E.D. Four dimensions of US legislative voting. In progress.

US Supreme Court (SCOTUS)

- Highest US court
- President nominates, Senate confirms
- 9 justices at most
- Selects own docket
- Decisions by majority vote
- “natural court” is a set of justices who sat on the bench together



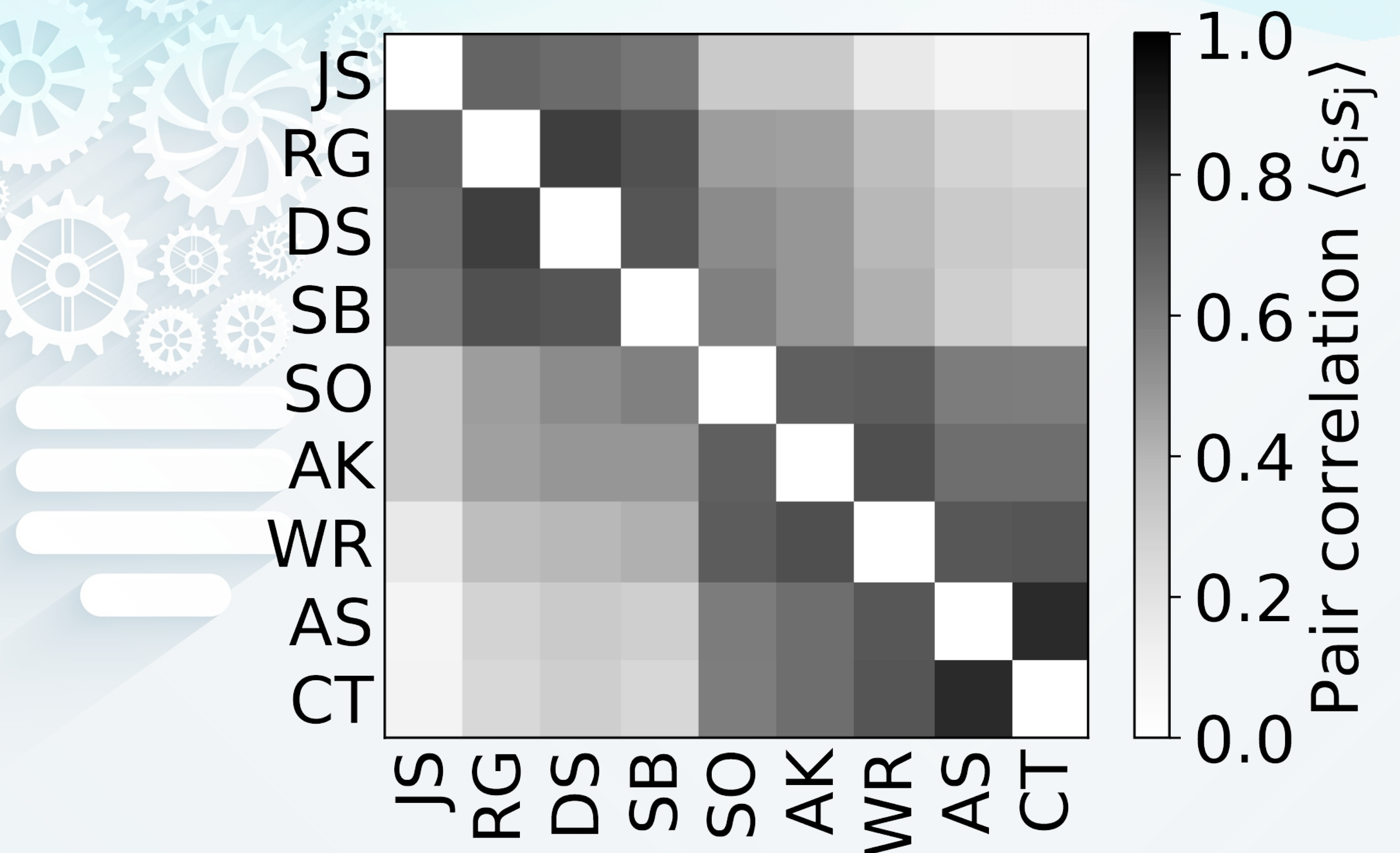
Maxent modeling with the US Supreme Court

$$\begin{aligned}\vec{s} &= \{ \text{yea}, \text{nay}, \text{nay}, \text{nay}, \text{nay}, \text{nay}, \text{yea}, \text{yea}, \text{yea} \} \\ &= \{ 1, -1, -1, -1, -1, -1, 1, 1, 1 \}\end{aligned}$$

Correlations imply interesting behavior

$$\begin{aligned}\vec{s} &= \{ \text{yea, nay, nay, nay, nay, nay, yea, yea, yea} \} \\ &= \{ 1, -1, -1, -1, -1, -1, 1, 1, 1 \}\end{aligned}$$

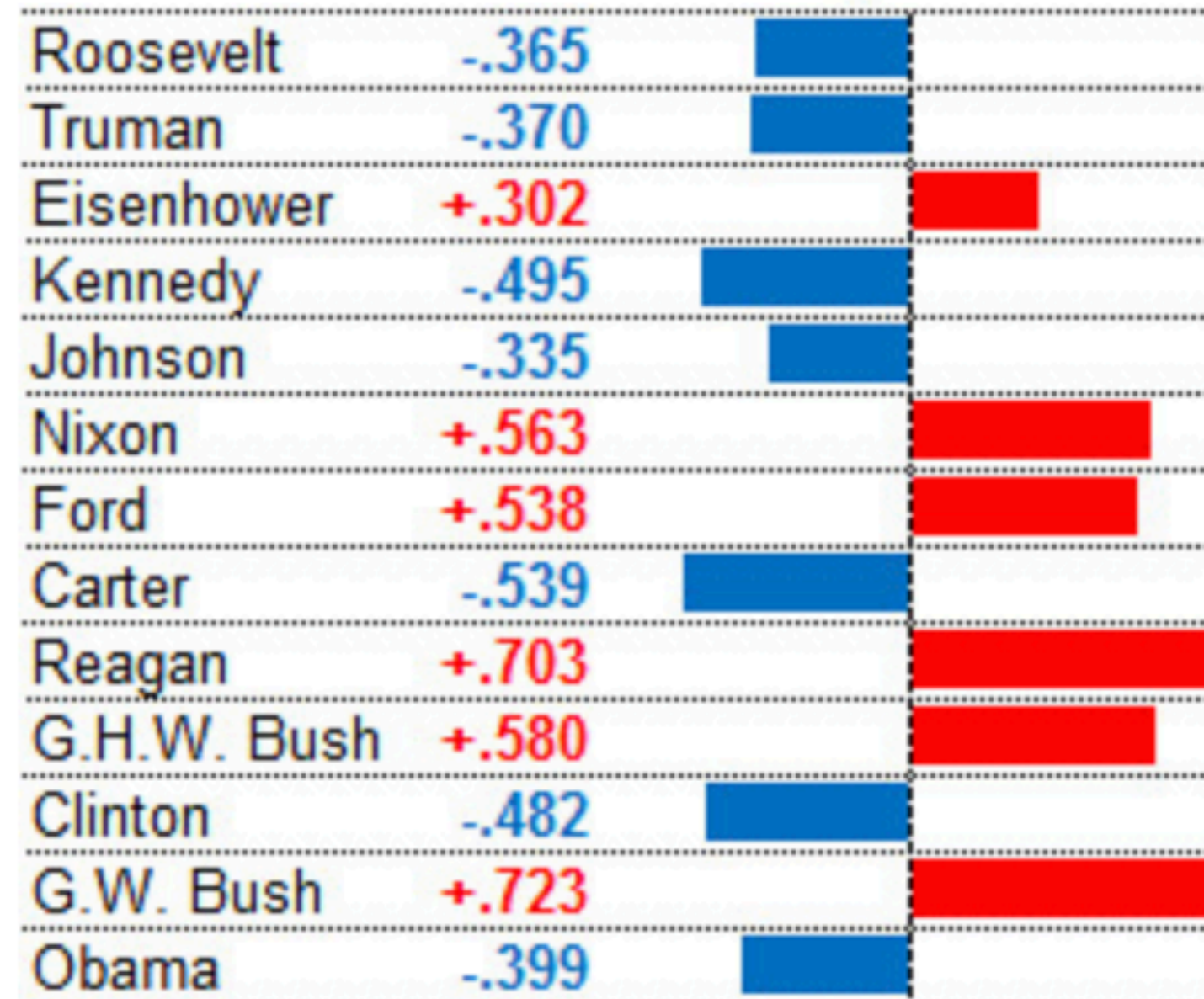
$$\langle s_i s_j \rangle = \sum_s p(\vec{s}) s_i s_j$$



(D)W-Nominate

- Spatial voting model originating in 1980s
- Assumption of independent voters maximizing utility along different political issues
- Equivalent to a kernel regression technique, Gaussian processes with radial basis function solved by maximizing the posterior
- Very parameter heavy, a problem with sparse voting data

DW-NOMINATE Common Space Scores for Recent Presidents



FiveThirtyEight
New York Times

Parameter counting with N=36 voters

Taking all justices from 1946-2016

K=8,737

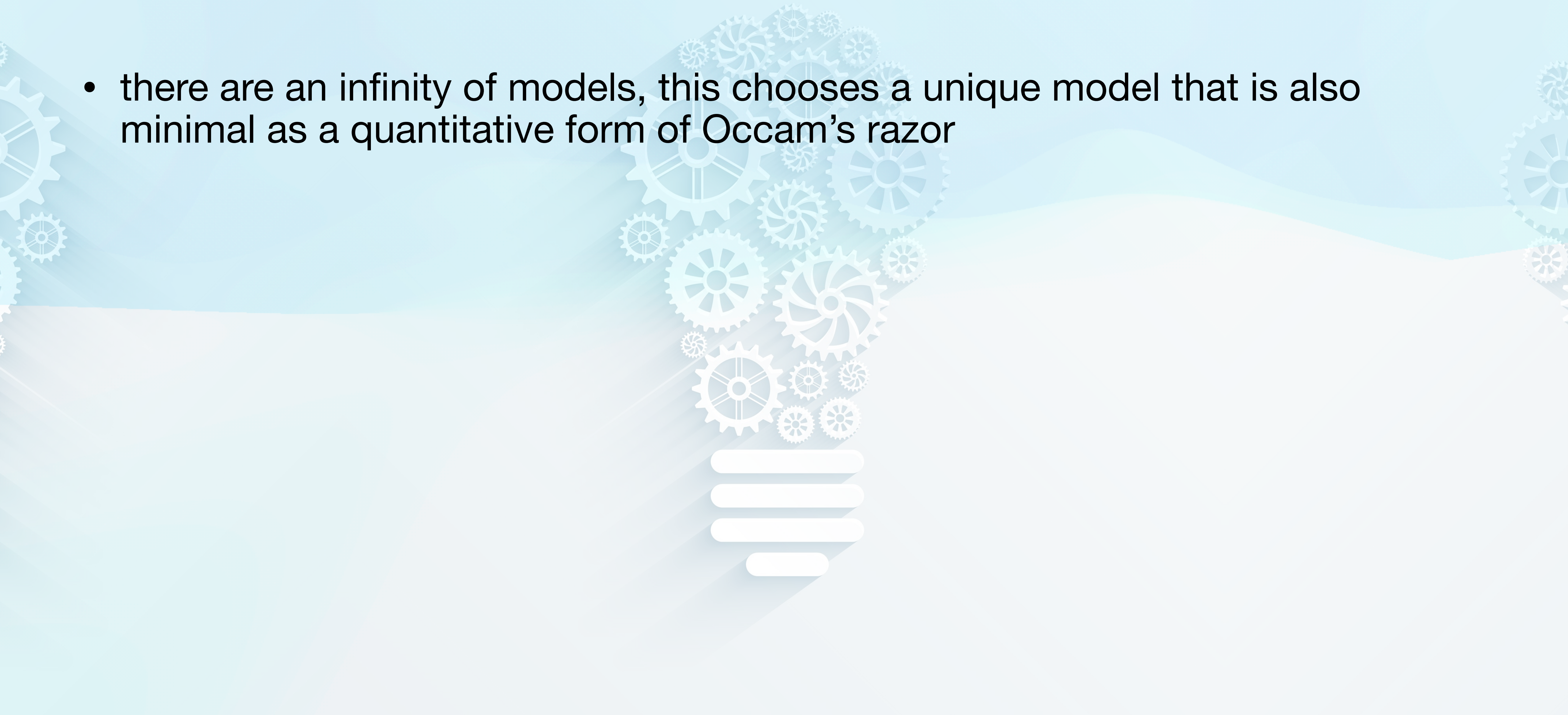
D=10

[number of voters] x [number of dimensions] + [number of votes] x [number of dimensions] x 2

~ 100,000 parameters

Maximum entropy principle

- there are an infinity of models, this chooses a unique model that is also minimal as a quantitative form of Occam's razor



Maximum entropy principle

- there are an infinity of models, this chooses a unique model that is also minimal as a quantitative form of Occam's razor
- maximize the entropy while fitting a limited set of important features of the system (Lagrangian multipliers from multivariable calculus)



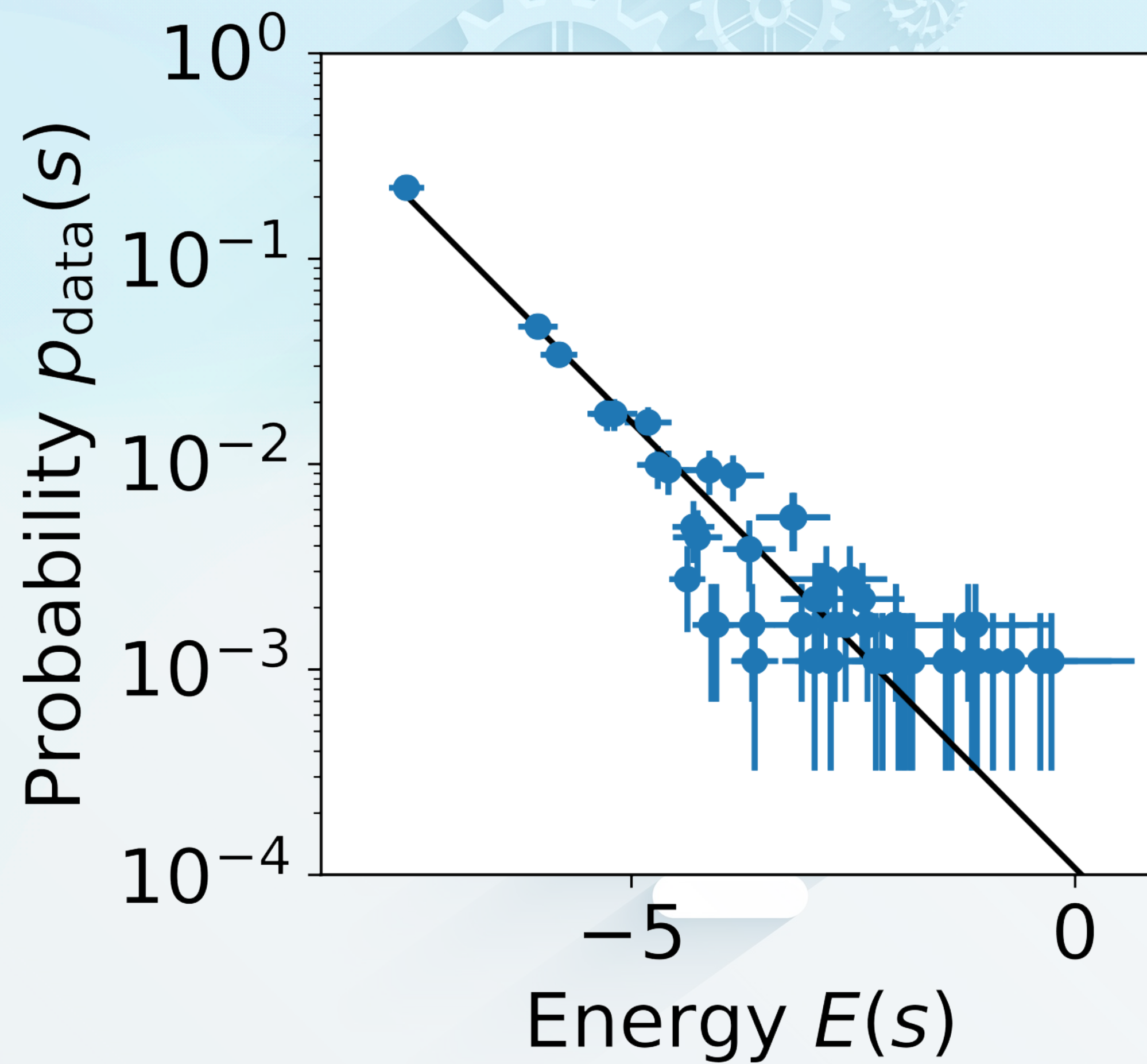
Maximum entropy principle

- there are an infinity of models, this chooses a unique model that is also minimal as a quantitative form of Occam's razor
- maximize the entropy while fitting a limited set of important features of the system (Lagrangian multipliers from multivariable calculus)
- leads to a "Boltzmann" probability distribution (a.k.a. exponential family, maximum entropy, restricted Boltzmann machines)

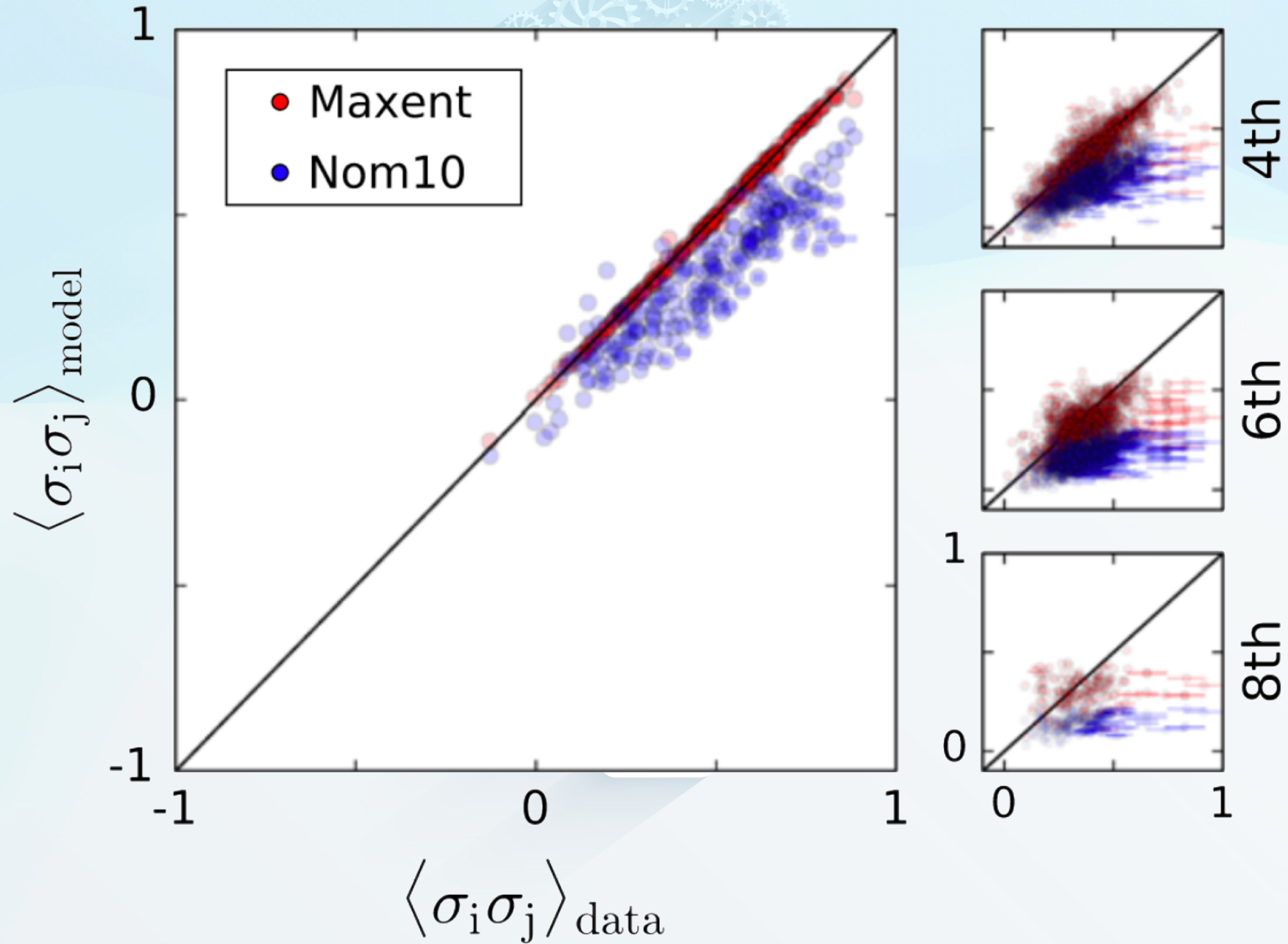
$$p(s) = \frac{e^{-E(s)}}{Z}$$

accounting for
interactions, 630
parameters

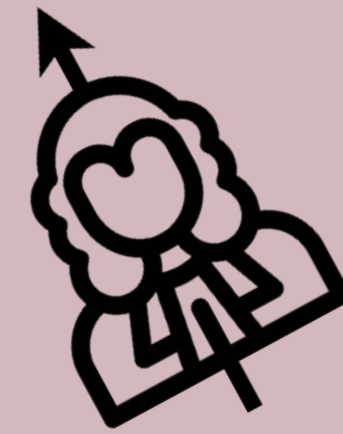
Model captures entire distribution



Maxent performs better than W-Nominate



**a powerful (elucidating and transparent)
way of building models with interactions**



Valence and interactions in judicial voting



Edward D. Lee, Complexity Science Hub
George Cantwell, Cambridge University
Philosophical Transactions Royal Society A



Lee, Broedersz, Bialek (2015)

Lee (2018)

Lee, Katz, Bommarito, Ginsparg (2020)

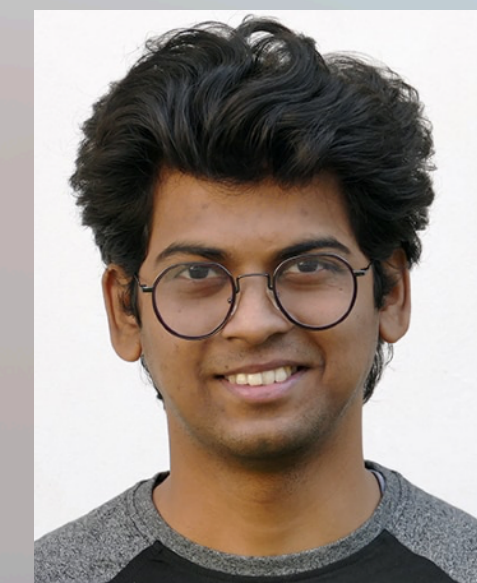
Testing the model on the Second Rehnquist Court

Model	Model evidence, $\log P(\mathcal{D})$
Bias, Eq. (7)	-3481.34
Interaction only, Eq. (9)	-3355.08
Combined, Eq. (10)	-3304.20

Next steps with analysis and experiments

- identify coalitions
- identify important players (“Partisan intuition” in 2018)
- identify sensitive points in the system that would tip the balance (“Sensitivity of collective outcomes identifies pivotal voters” in 2019; “Discovering sparse control strategies in neural activity” in 2022)
- go beyond one-dimensional ideology (Rees & Lee, in progress)

Cascades of conflict activity

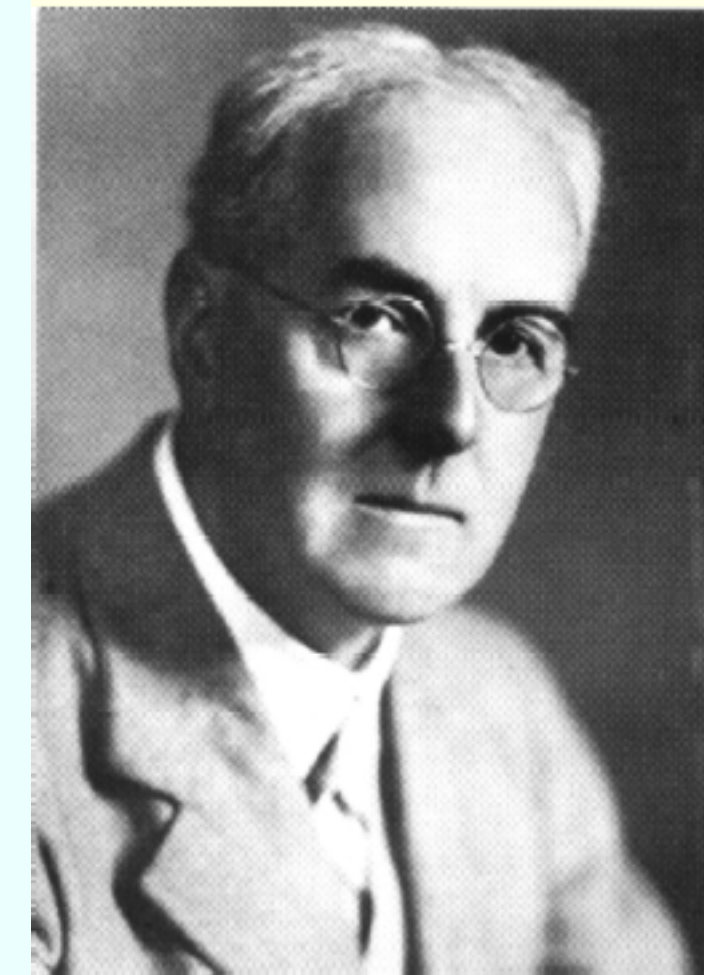


Niraj Kushwaha



Woi Oh Sok

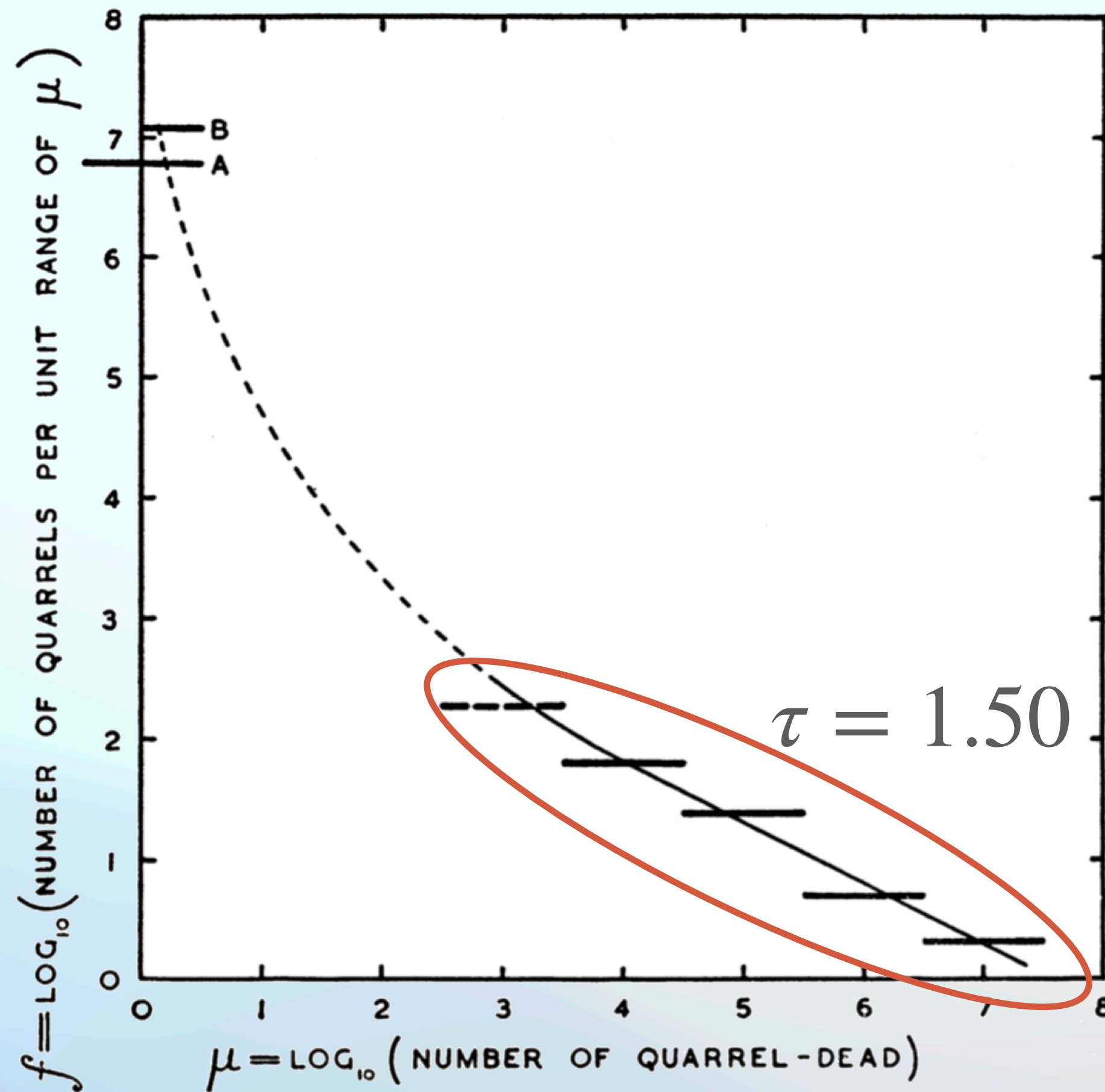
Richardson's law for fatalities



Lewis Fry Richardson
(1881-1953)

Interstate wars from 1820-1945

Normalized frequency $P(F)$



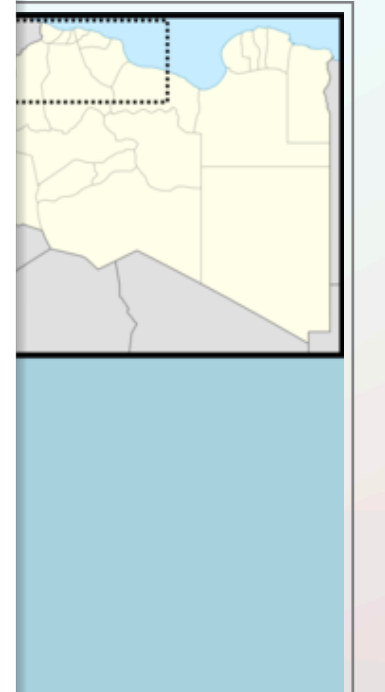
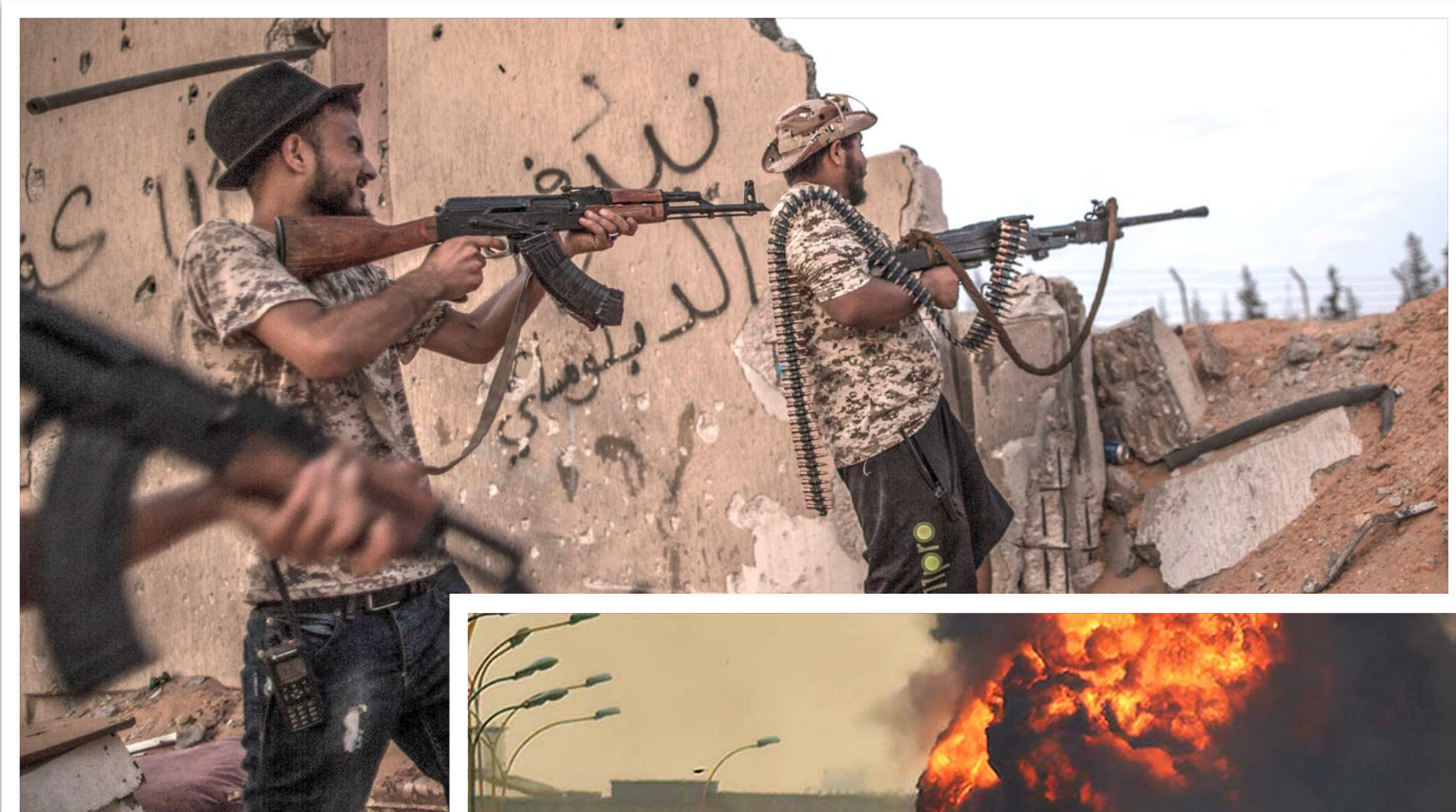
Number of fatalities F

Power law distribution

$$P(F) \sim F^{-\tau}$$

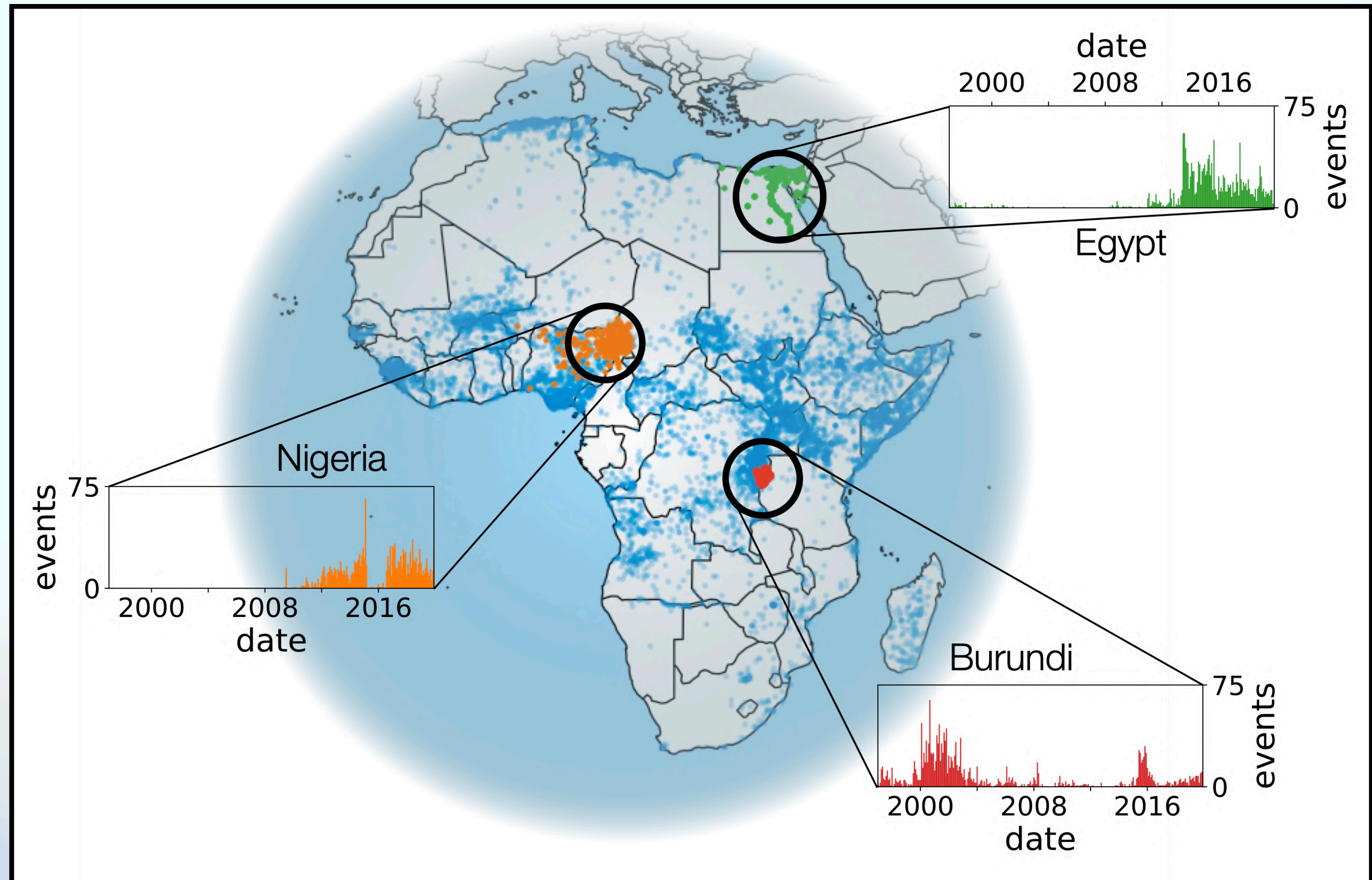
World War II

Libyan Civil War

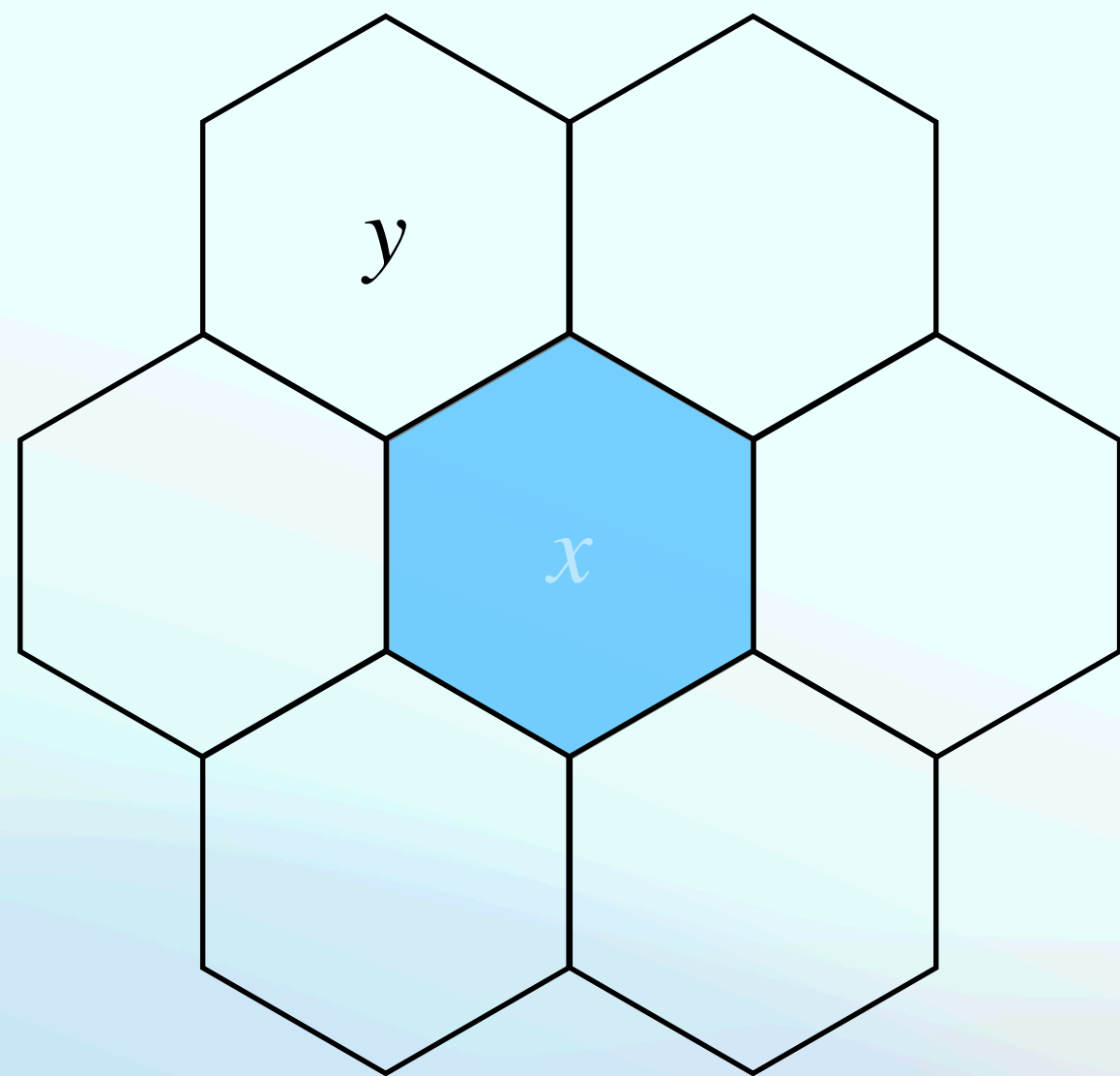


Armed Conflict & Location Event Data Project

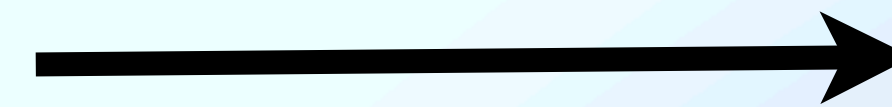
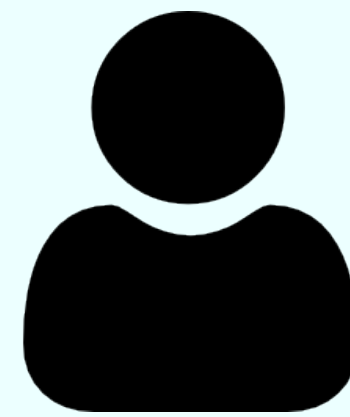
- 27 years (1997–today)
- Spans Africa (~8,000 km)
- 400,000 events (reports, fatalities, day, location)
- >700,000 direct fatalities



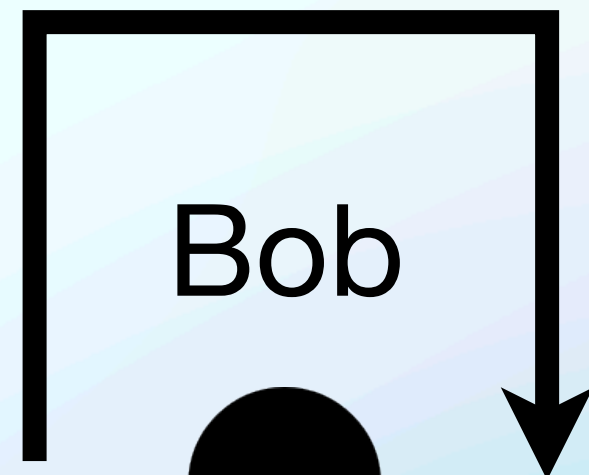
Inferring relationships between sites i and j

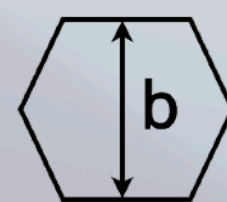
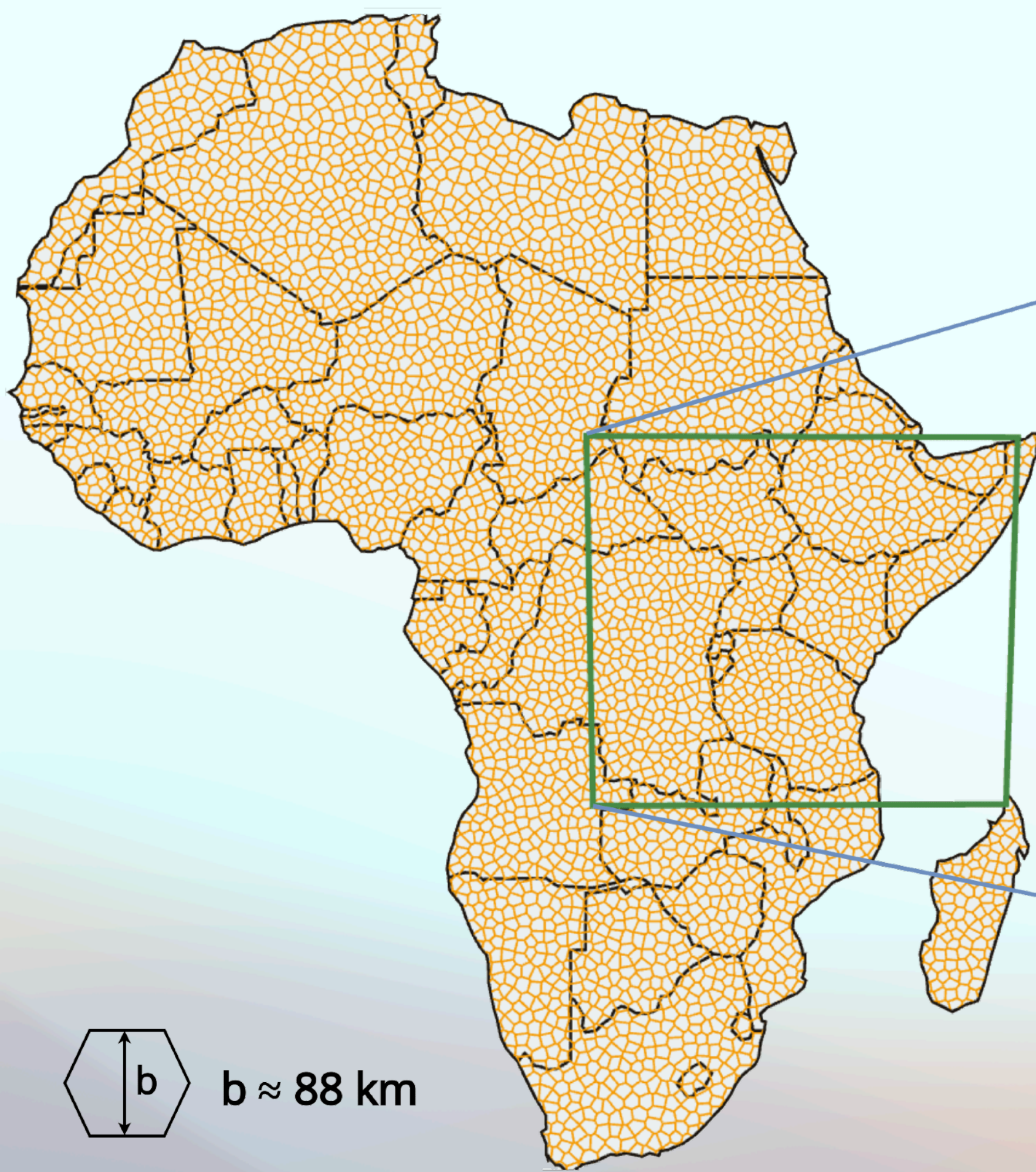


Alice

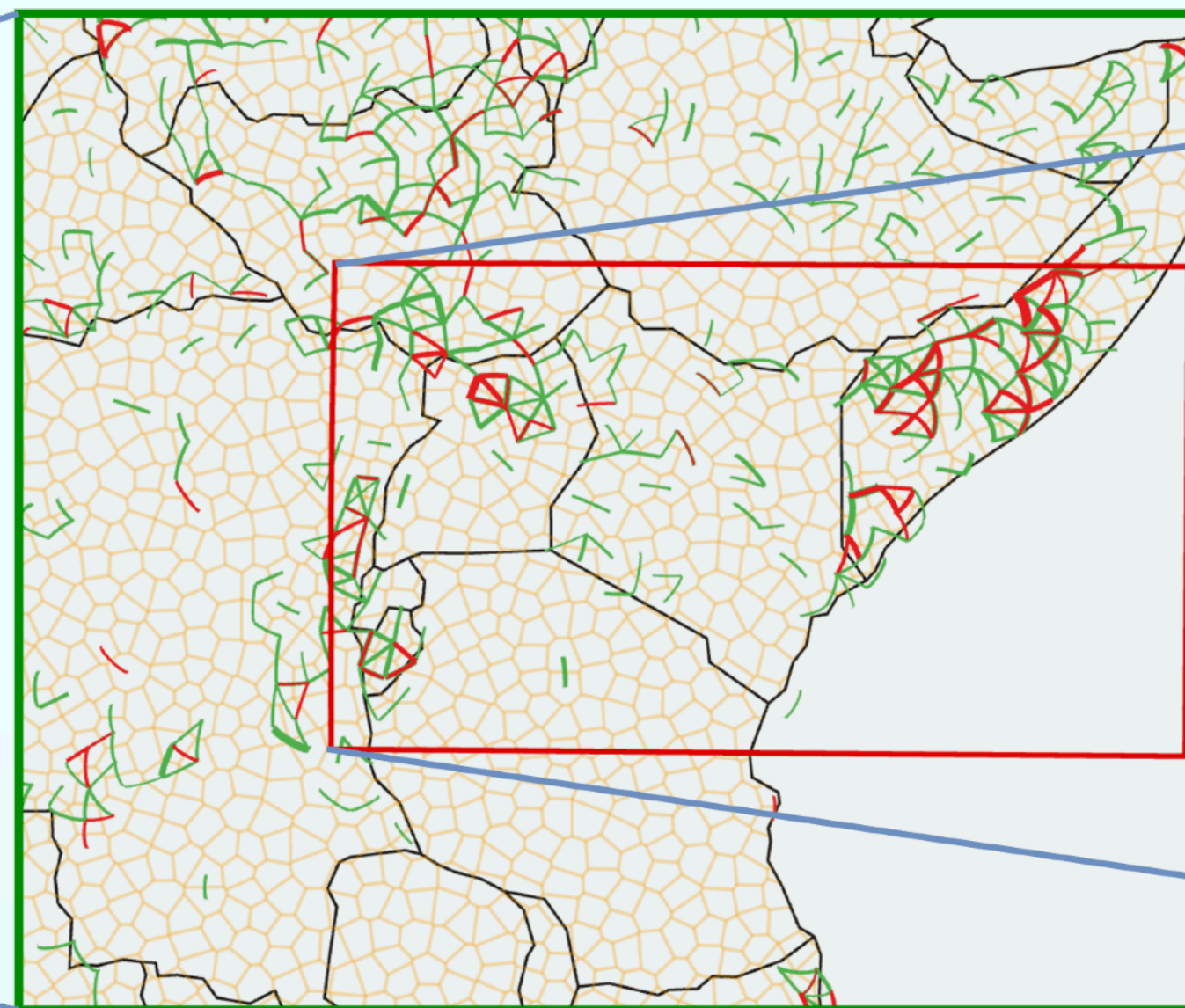


Bob

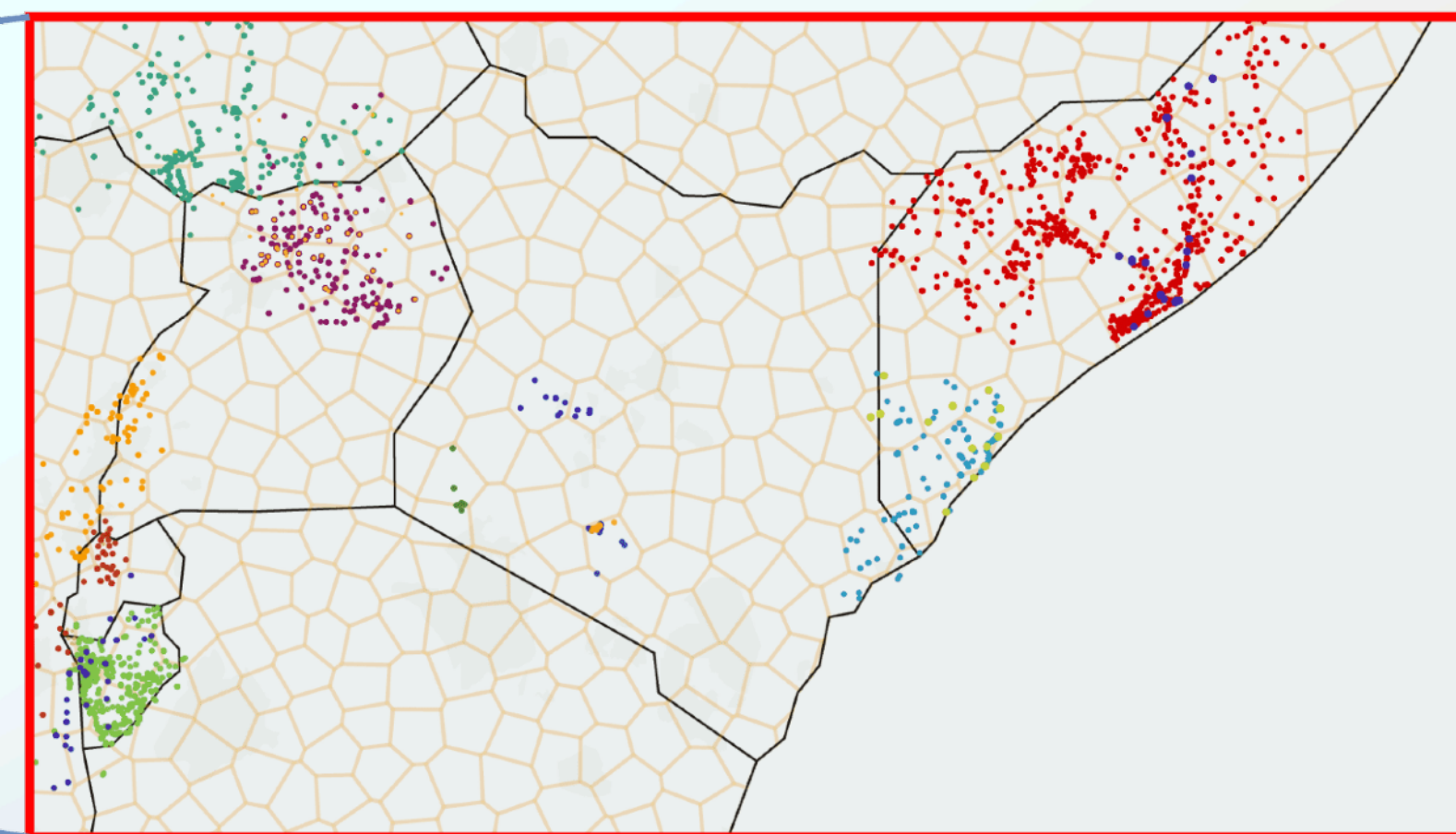




$b \approx 88 \text{ km}$



- Causal link (uni-directional)
- Causal link (bi-directional)

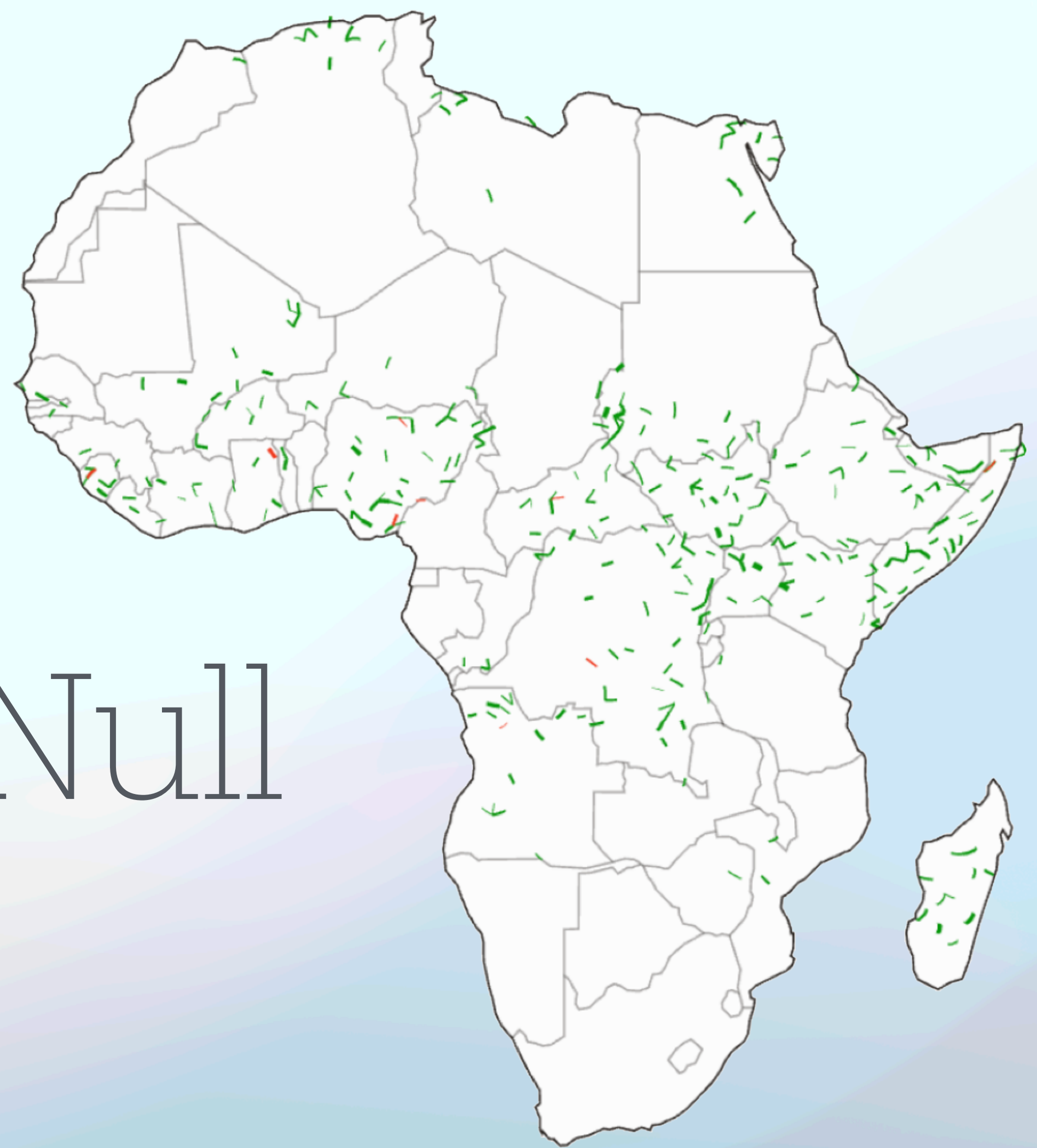


• Conflict event

Conflict network



Data



Null

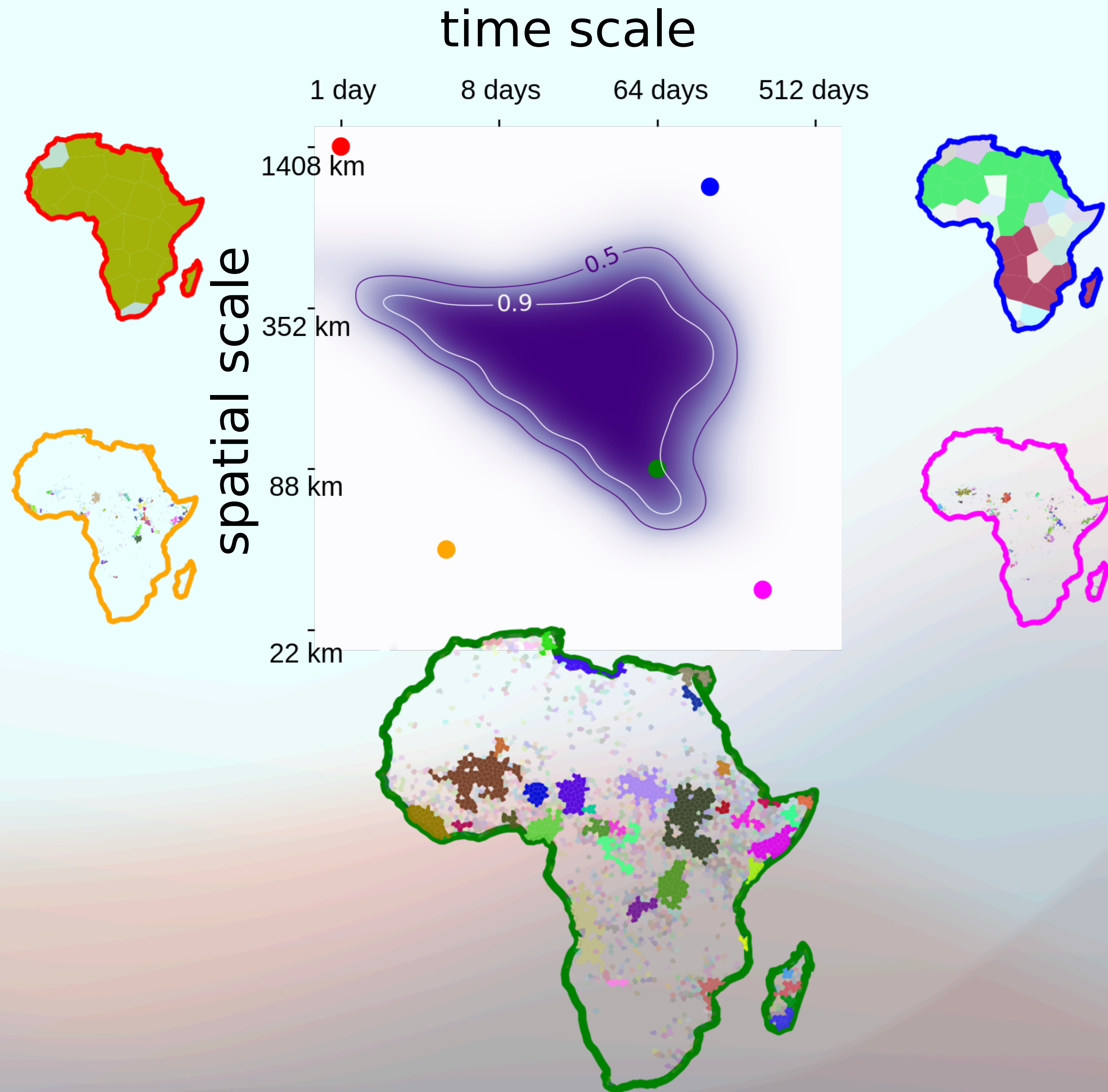
Conflict avalanches

Temporal bin=0

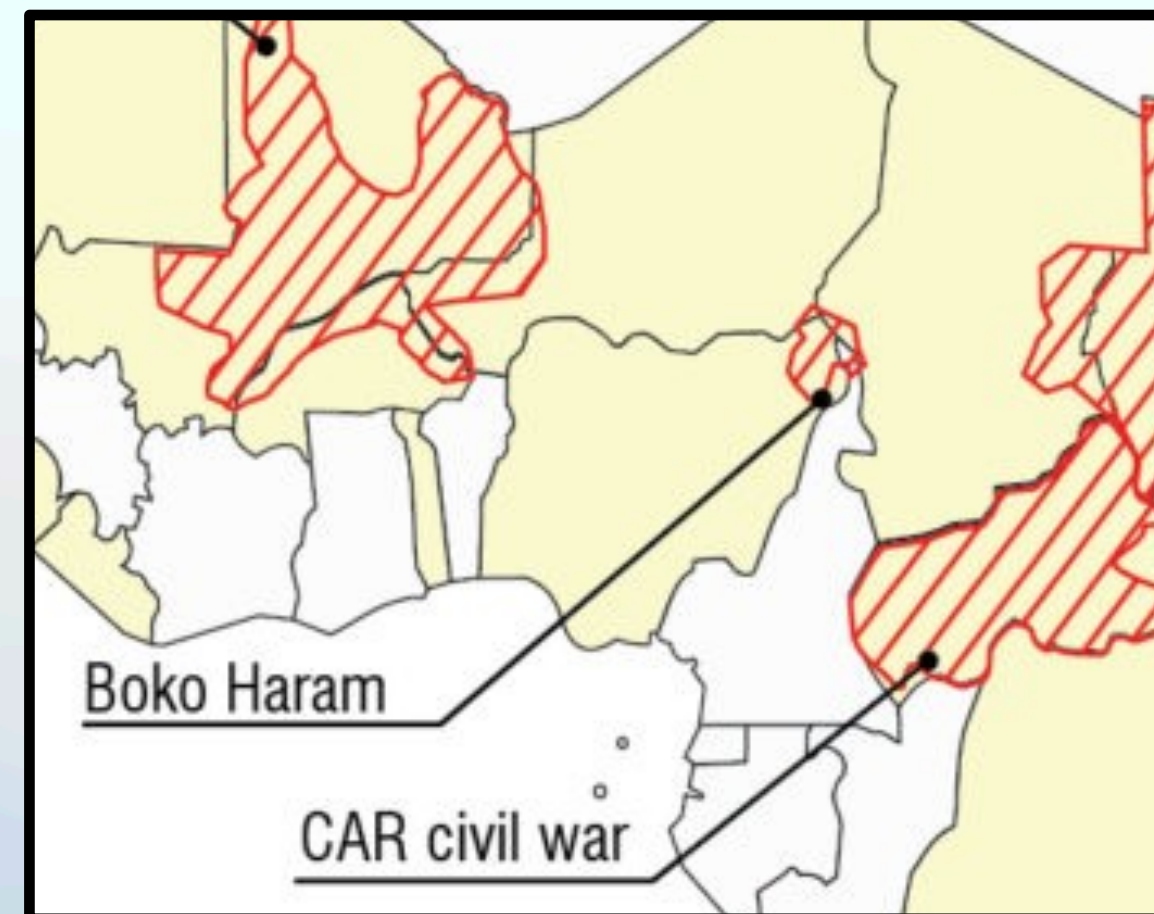
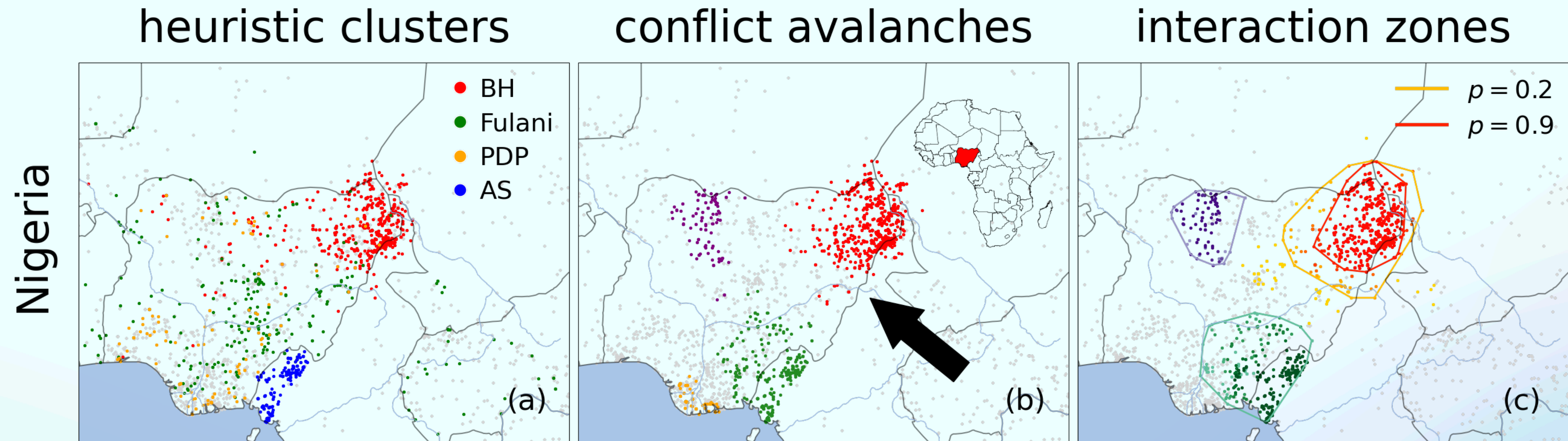


(01 Jan 1997 - 05 Mar 1997)

Discovering the mesoscale



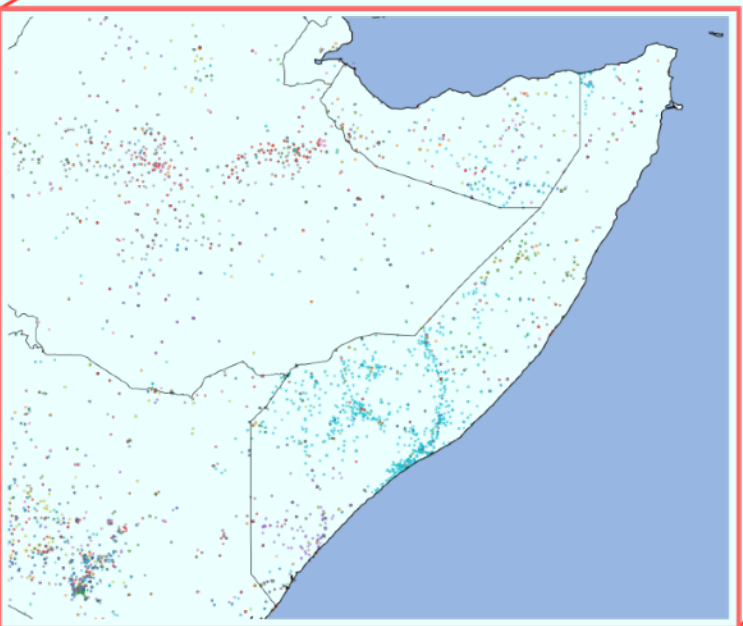
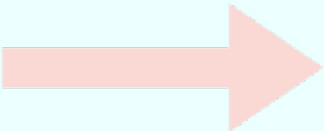
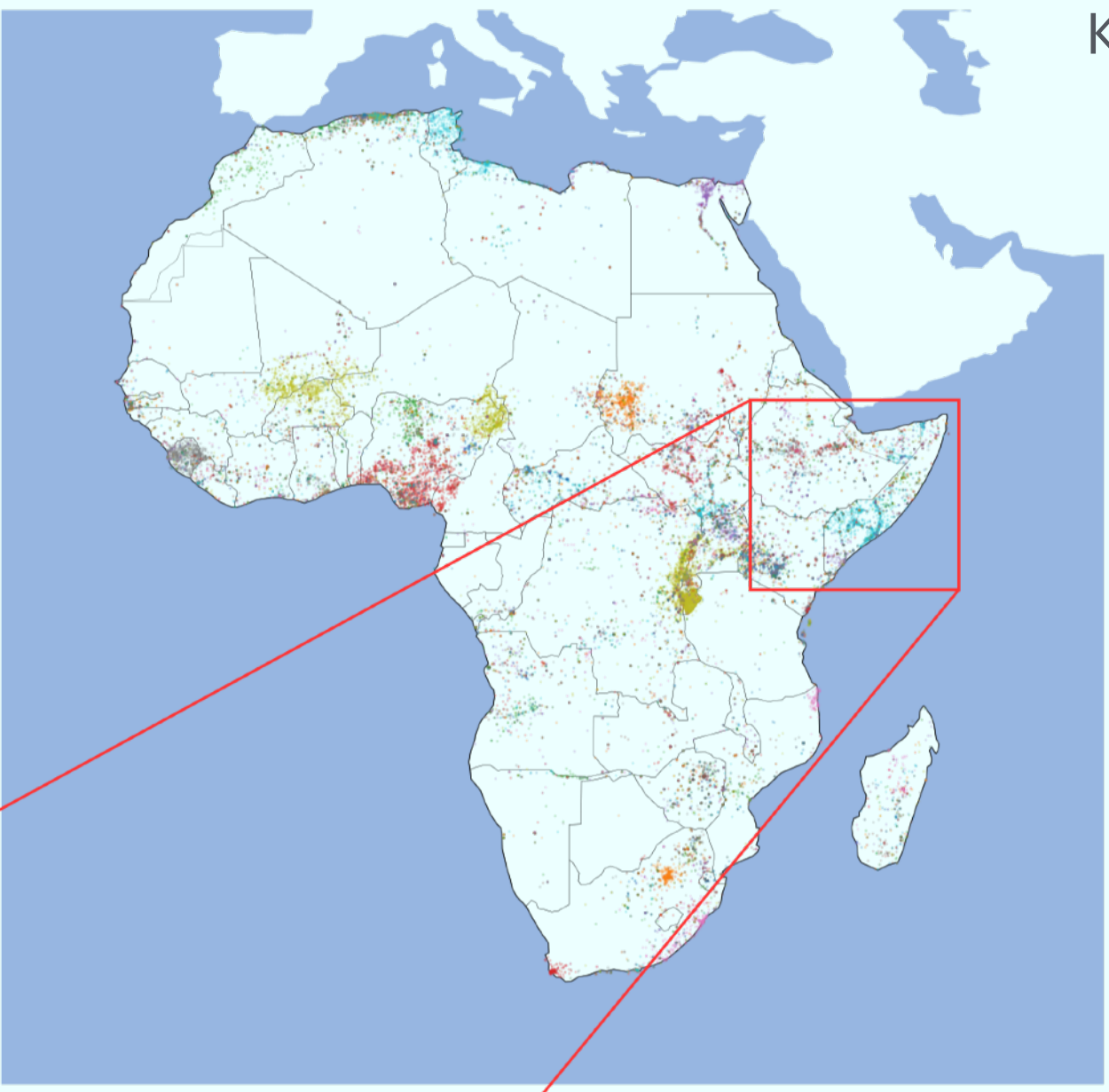
Systematic clusters discover mechanism



A

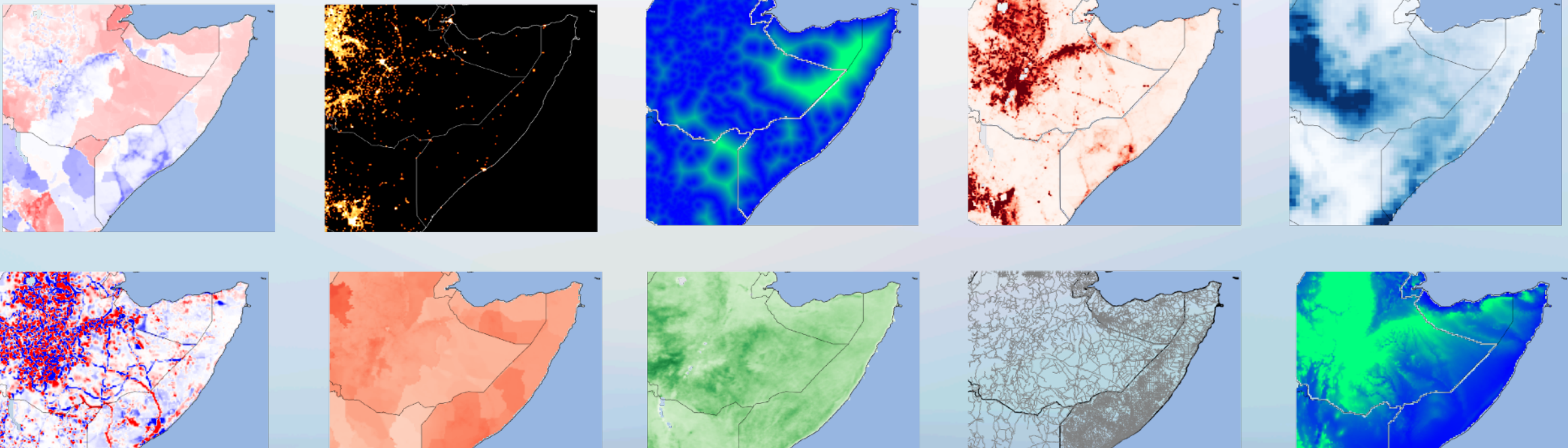


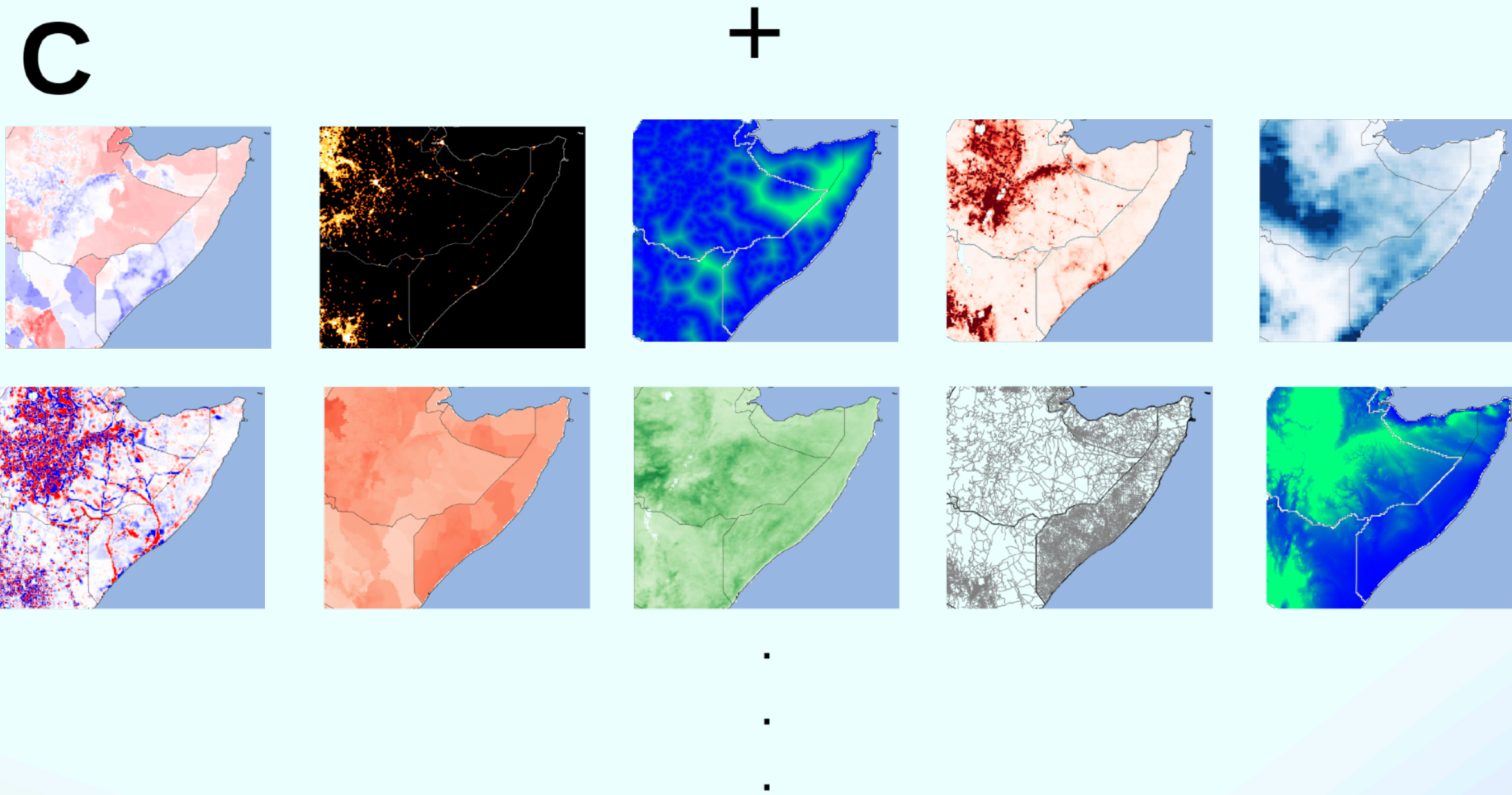
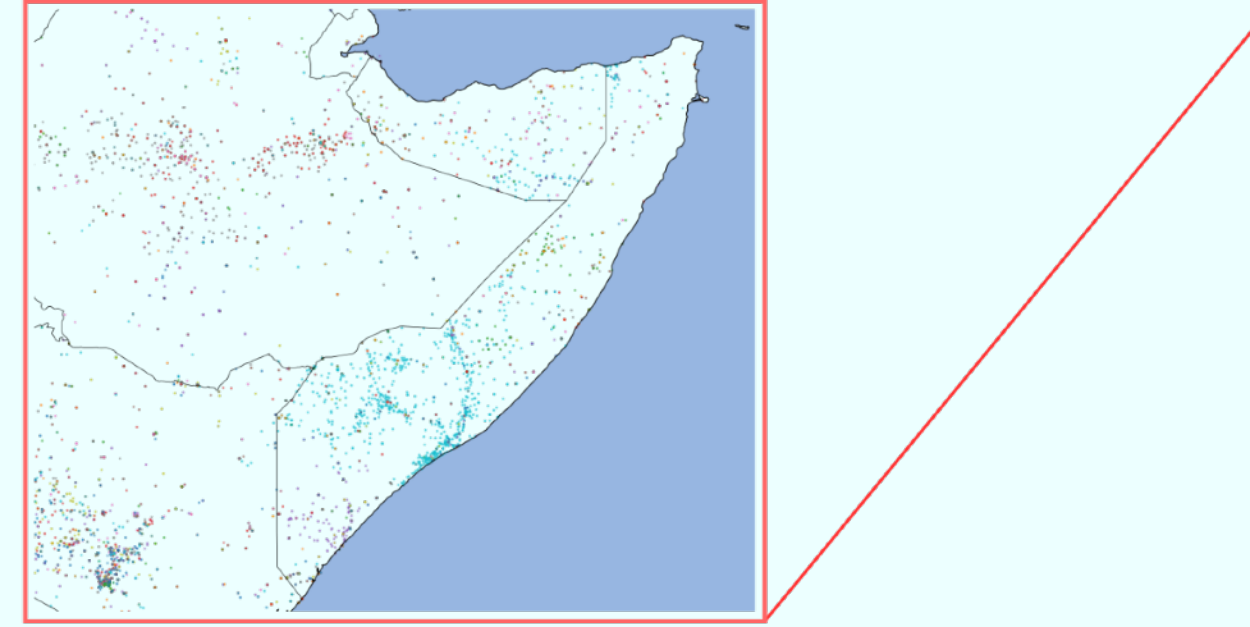
B



C

+





Climatic variables

Temperature
Precipitation
NDVI

Geographic variables

Distance from inland water bodies
Distance from coastline
Elevation

Population variables

Population count
Population density

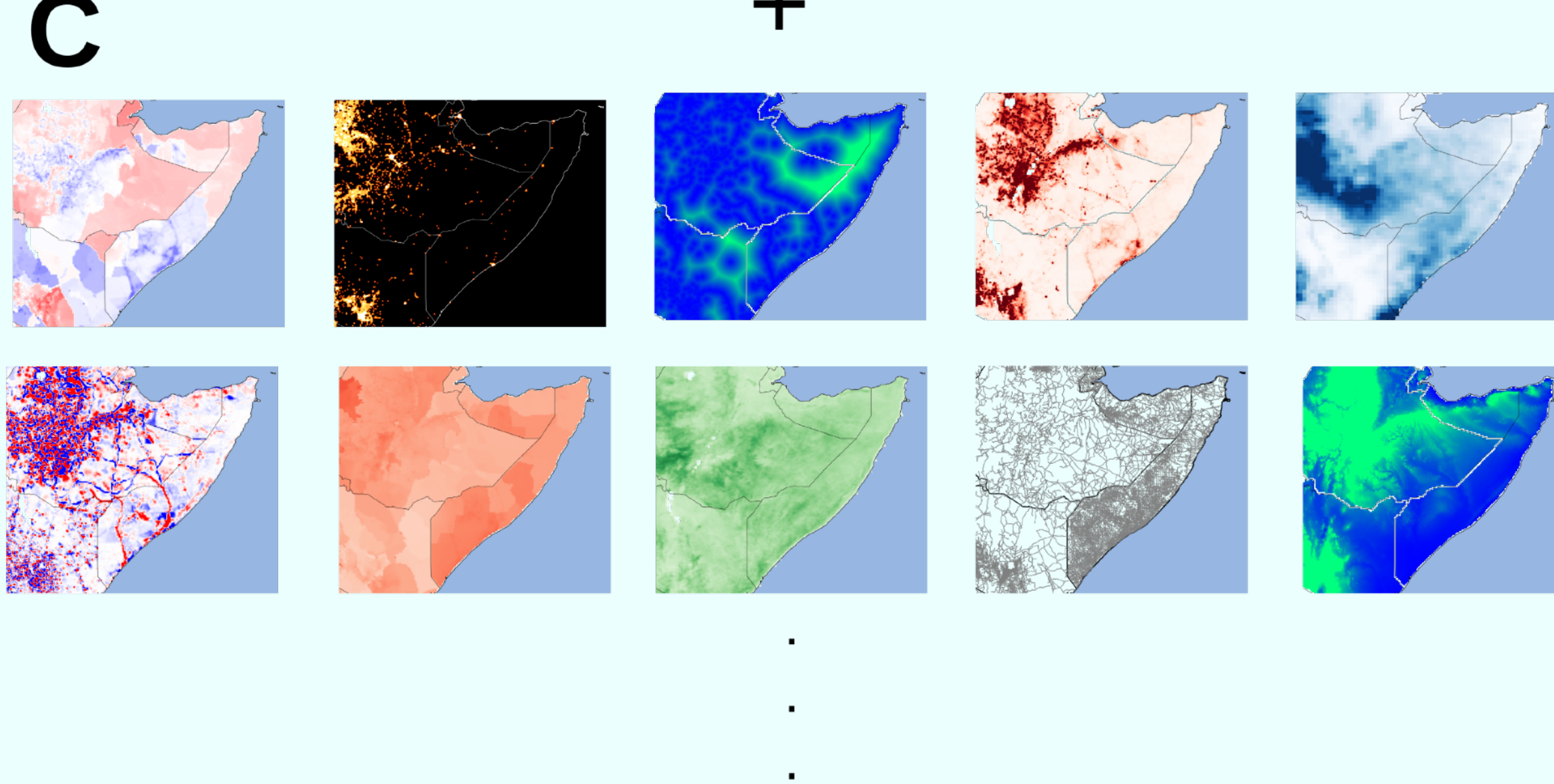
Avalanche variables

Total number of fatalities
Total number of news reports
Total duration

Economic variables

Demographic variables

Infrastructural variables



Climatic variables

Temperature
 Precipitation
 NDVI

Geographic variables

Distance from inland water bodies
 Distance from coastline
 Elevation

Population variables

Population count
 Population density

Avalanche variables

Total number of fatalities
 Total number of news reports
 Total duration

Economic variables

GDP
 GDP per capita
 HDI
 % increase in GDP
 % increase in GDP per capita
 % increase in HDI

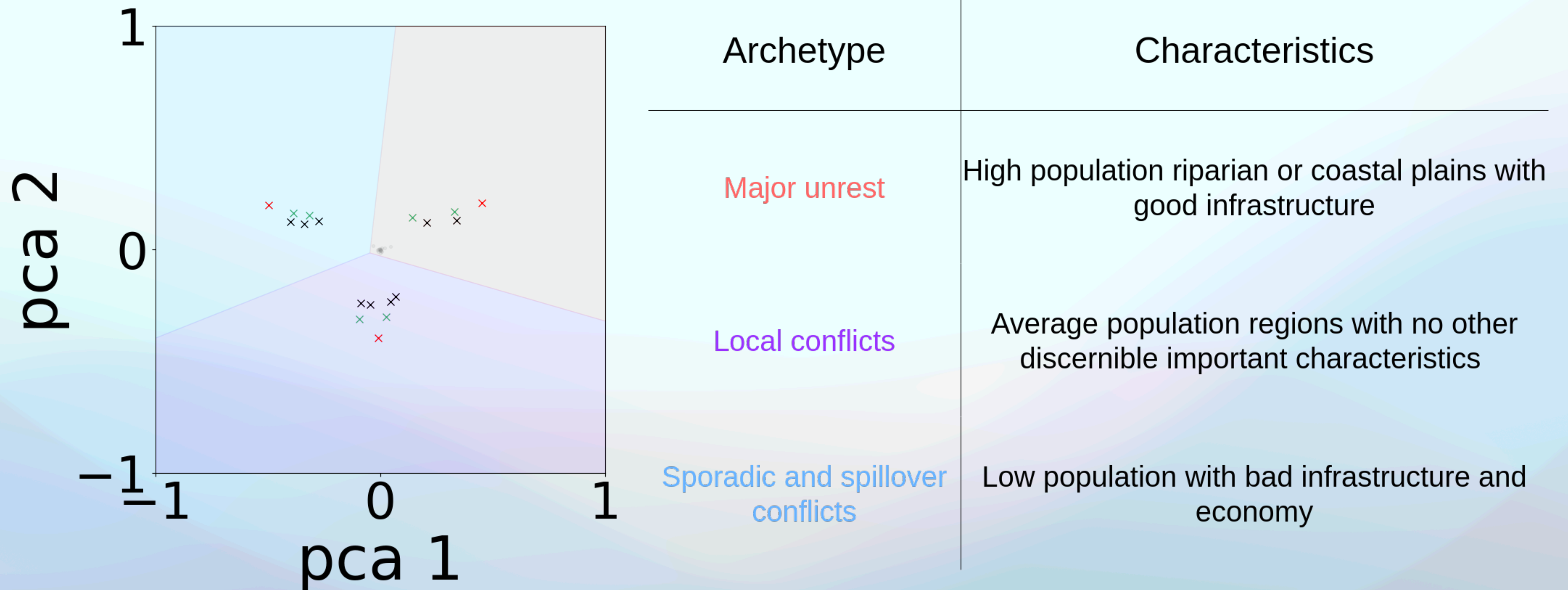
Demographic variables

Net migration
 Birth rate
 Death rate
 Interacting ethnic groups

Infrastructural variables

Cellular phone per 100 people
 Electric consumption
 Shortest distance to roads
 Night light

Triangle of madness



information consumption and firm size

Eddie Lee Complexity Science Hub

Alan Kwan Hong Kong University

Anjali Bhatt Harvard Business School

Rudi Hanel Complexity Science Hub

Frank Neffke Complexity Science Hub



Brenner, Bialek, & de Ruyter van Steveninck, *Neuron* (2000)

Endres & Wingreen, *PNAS* (2008)

Lee, Flack & Krakauer, preprint (2022)

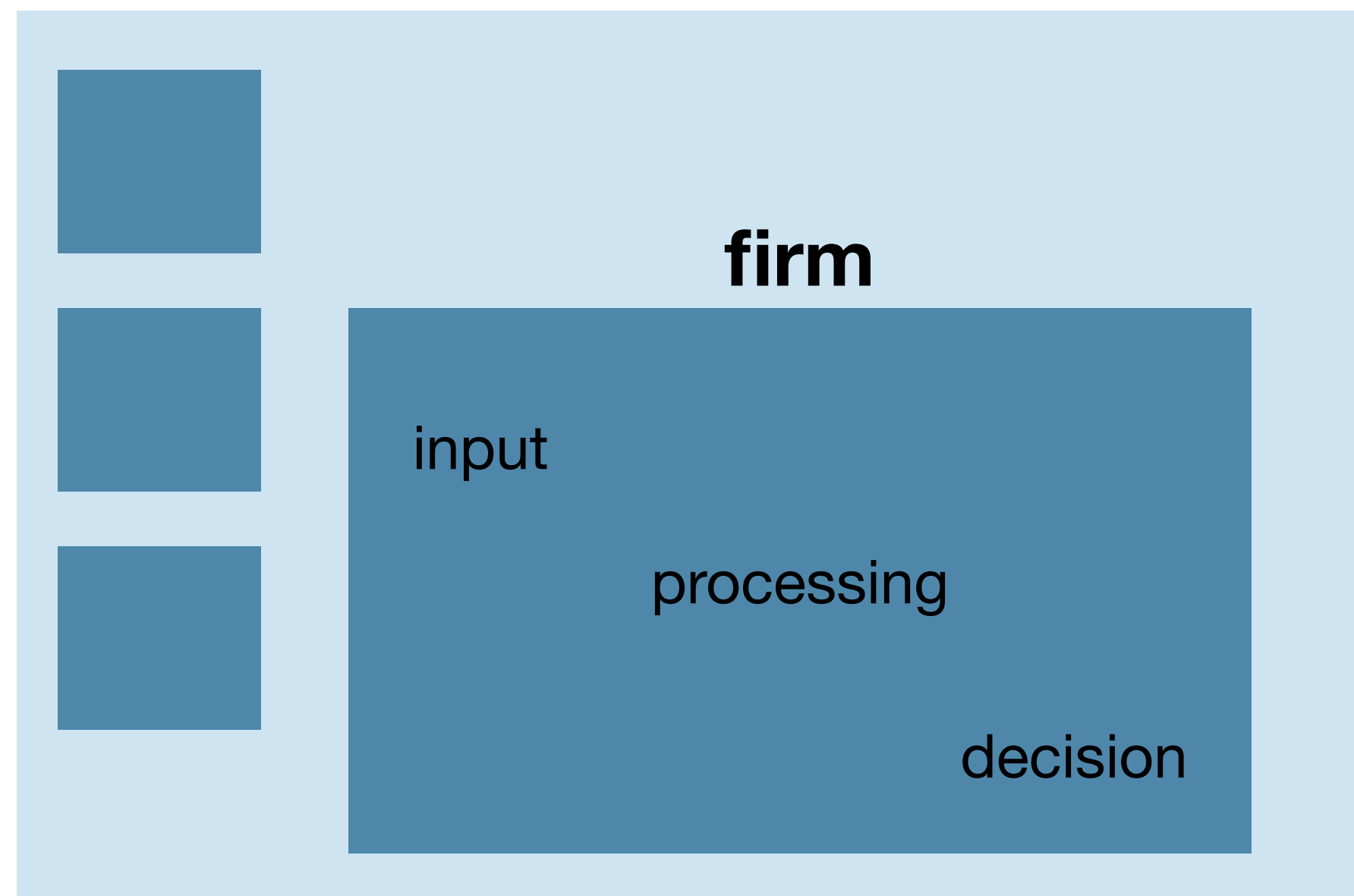
Fleig & Balasubramanian, preprint (2023)



Stentor coeruleus

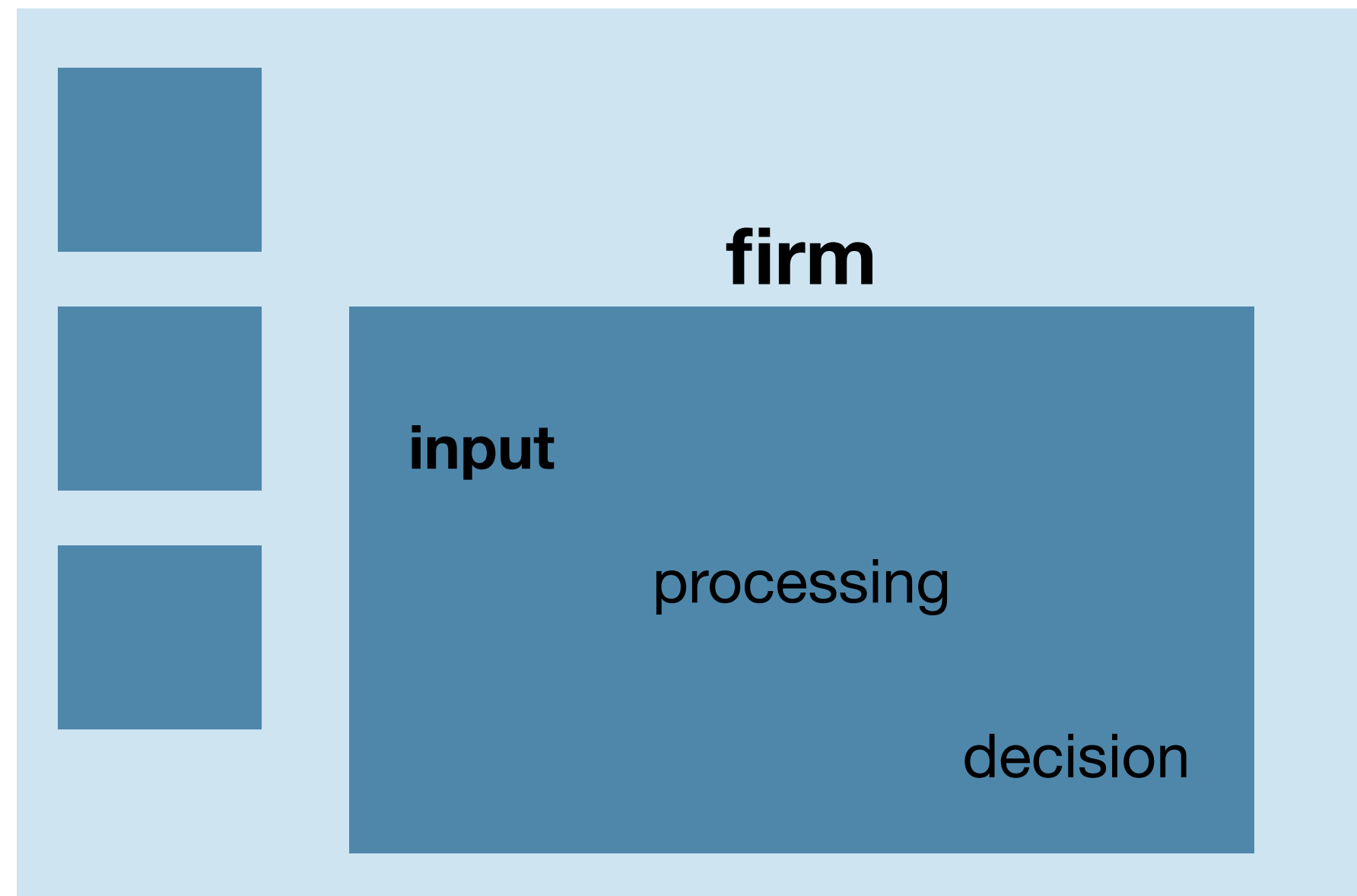
firms are information entities

environment



employees consume information

environment



data details

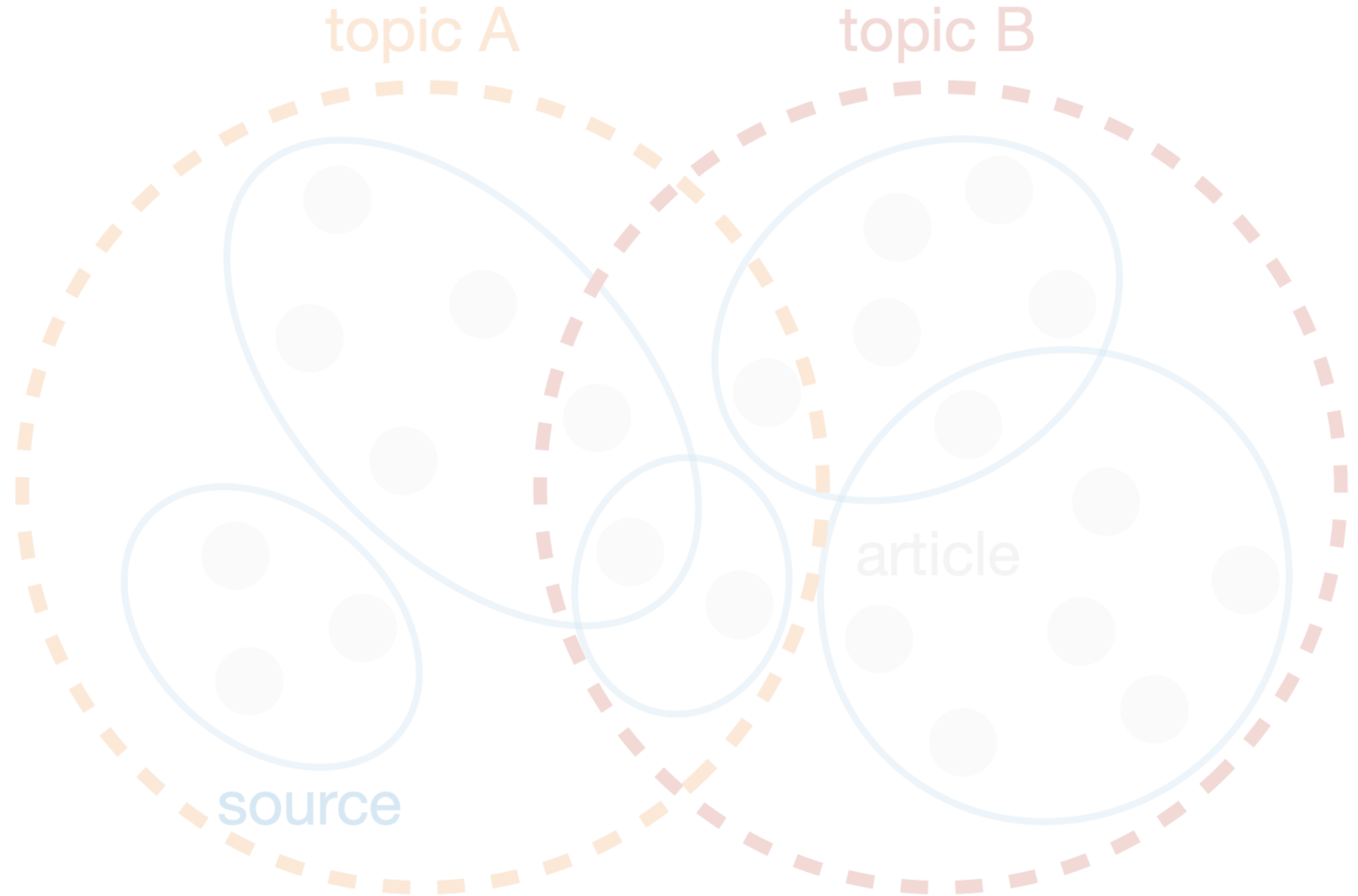
- publishers spanning technology, marketing, legal, biotech, manufacturing, and a wide range of business services
- 1 billion records —> 100 million records per day, filtering by to distinguish between sports, adult, and entertainment websites versus a set of articles from well-known business publishers
- two-week period between the dates of June 10 and June 23, 2018
- filters applied toward the content and visitors suggests that information consumption events observed in this study contain the subset of observations most likely to come from work-related visitors and work-related content

Sources like

Bloomberg News
Wall Street Journal
Forbes
Business Insider
CBSi
1105Media
ITCentral Station
Questex
etc.

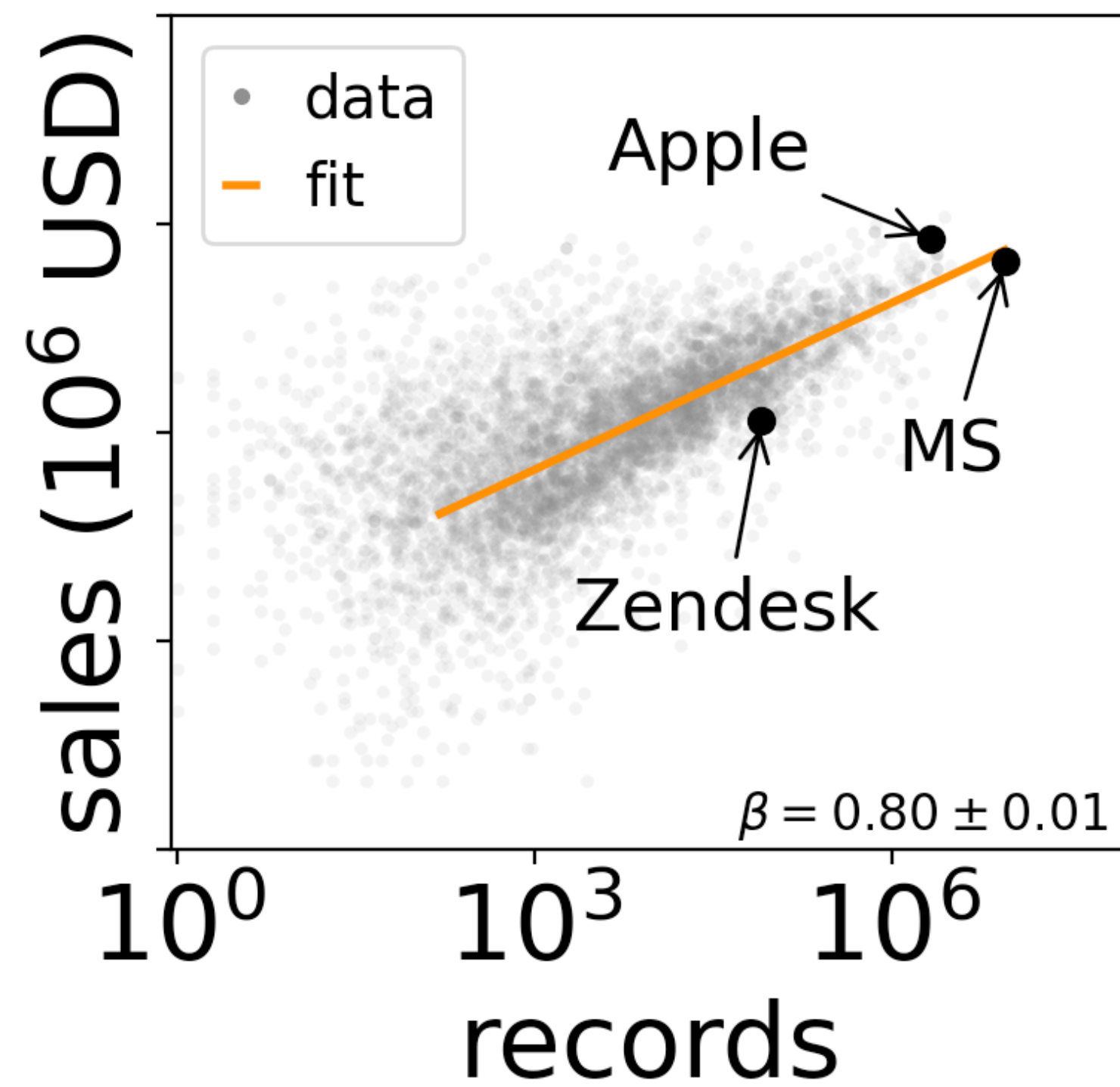
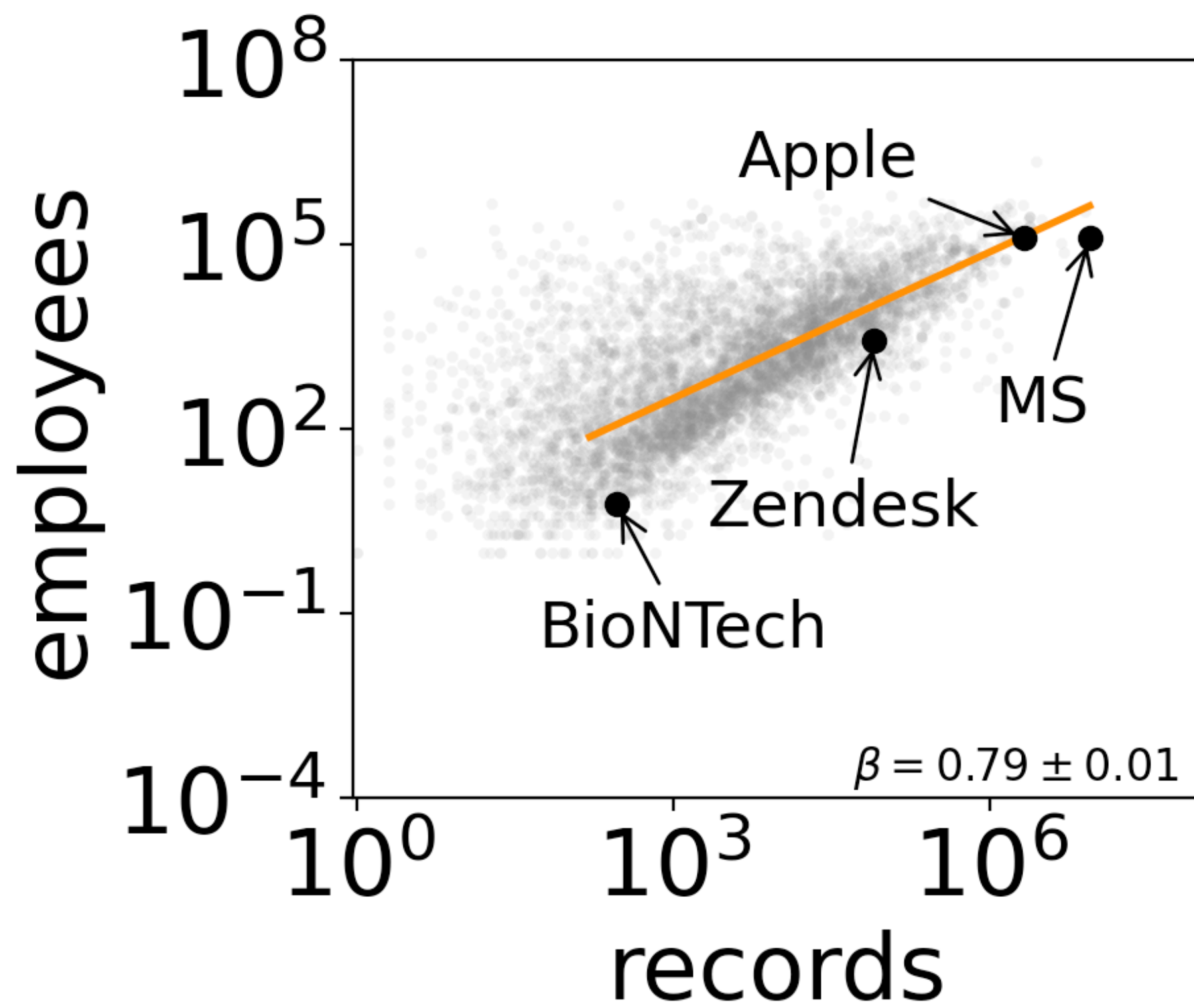
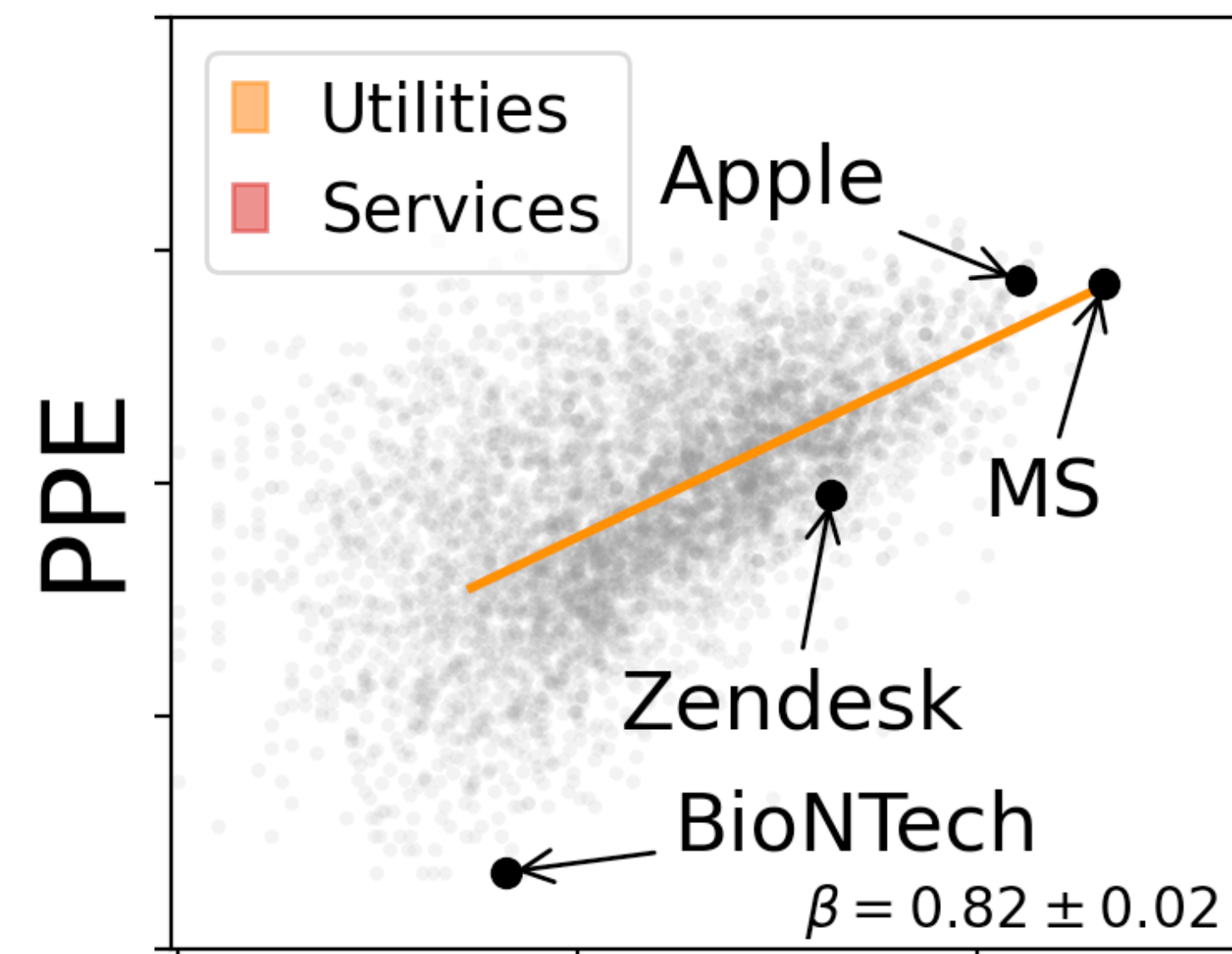
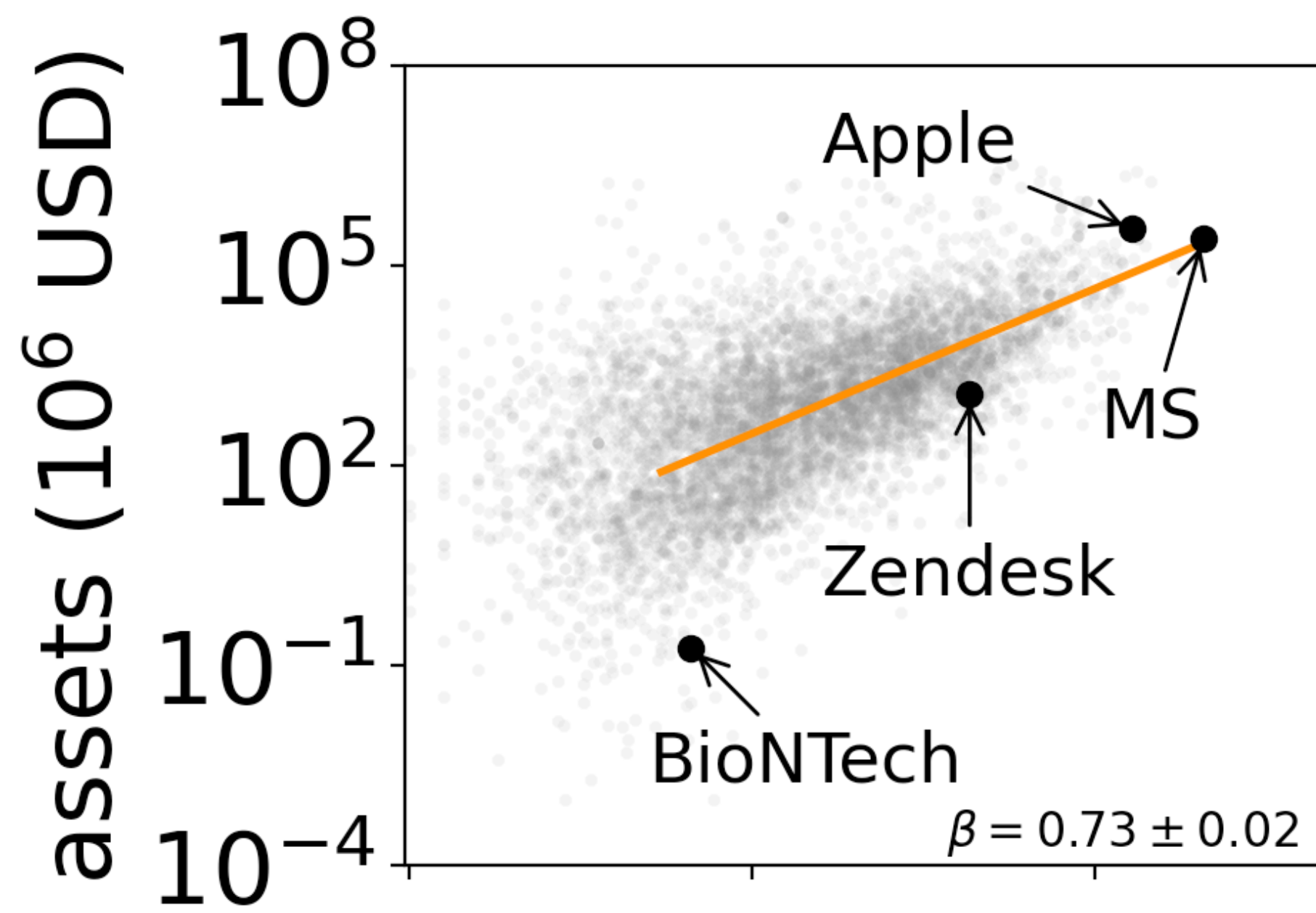
Business-relevant topics like

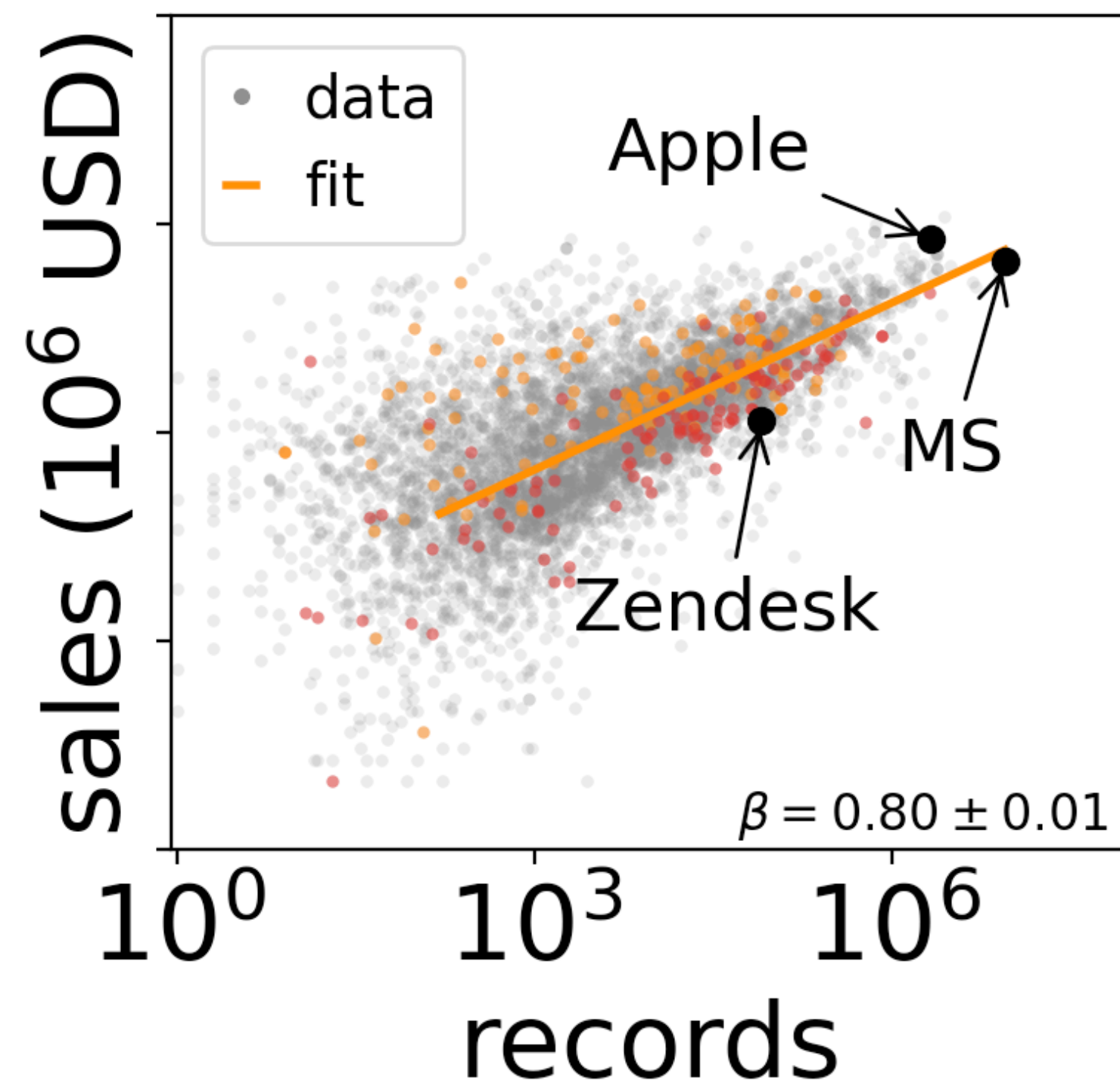
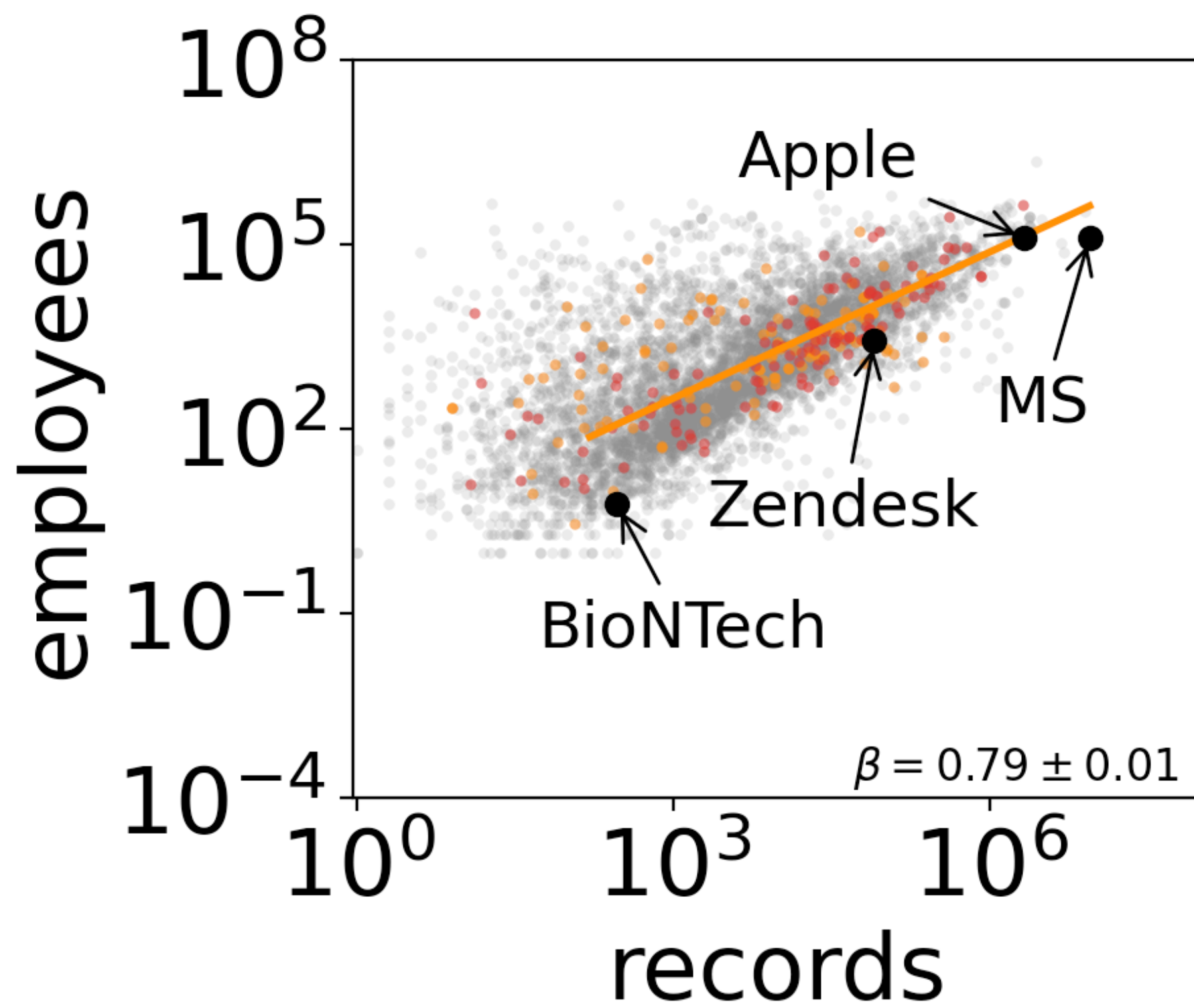
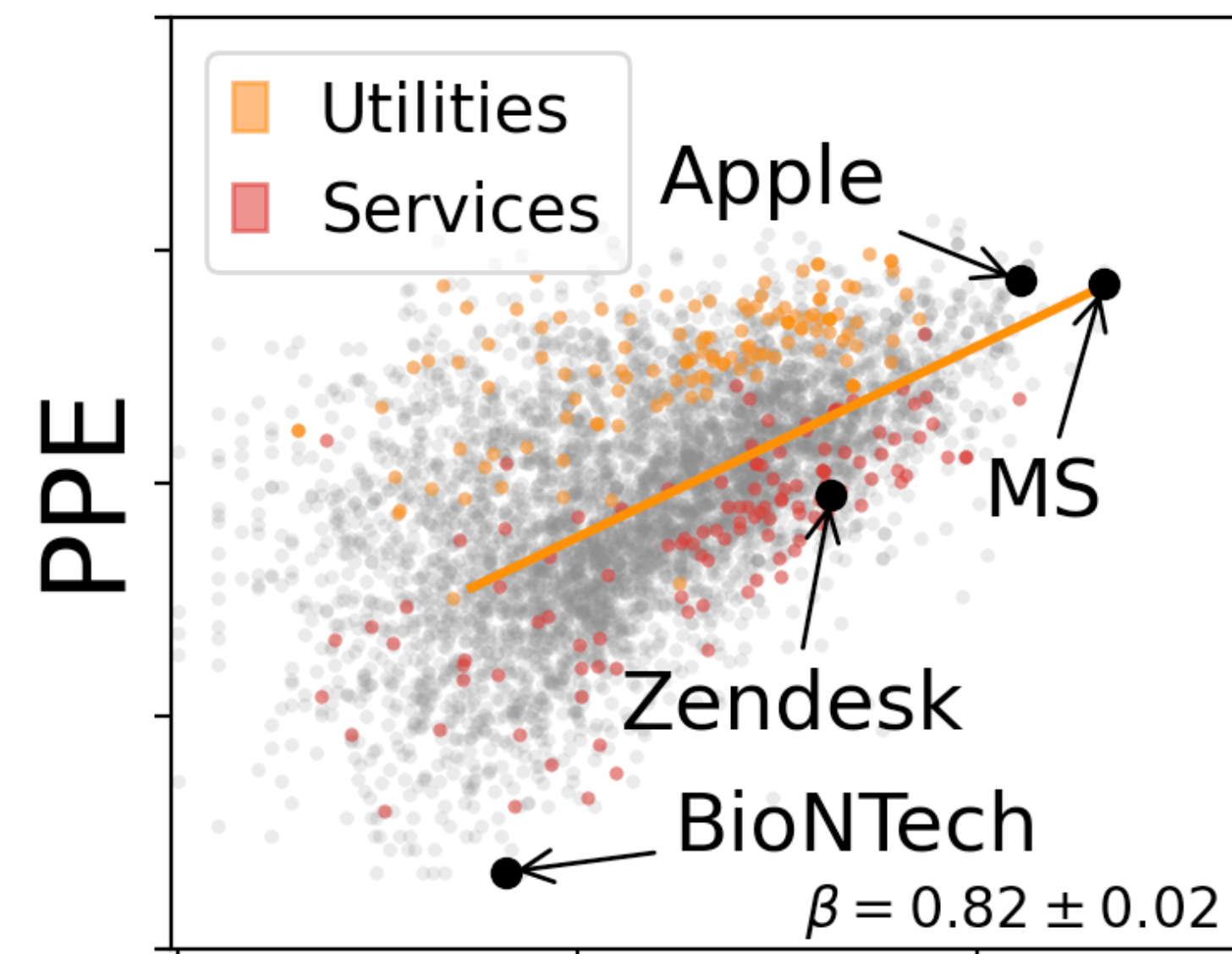
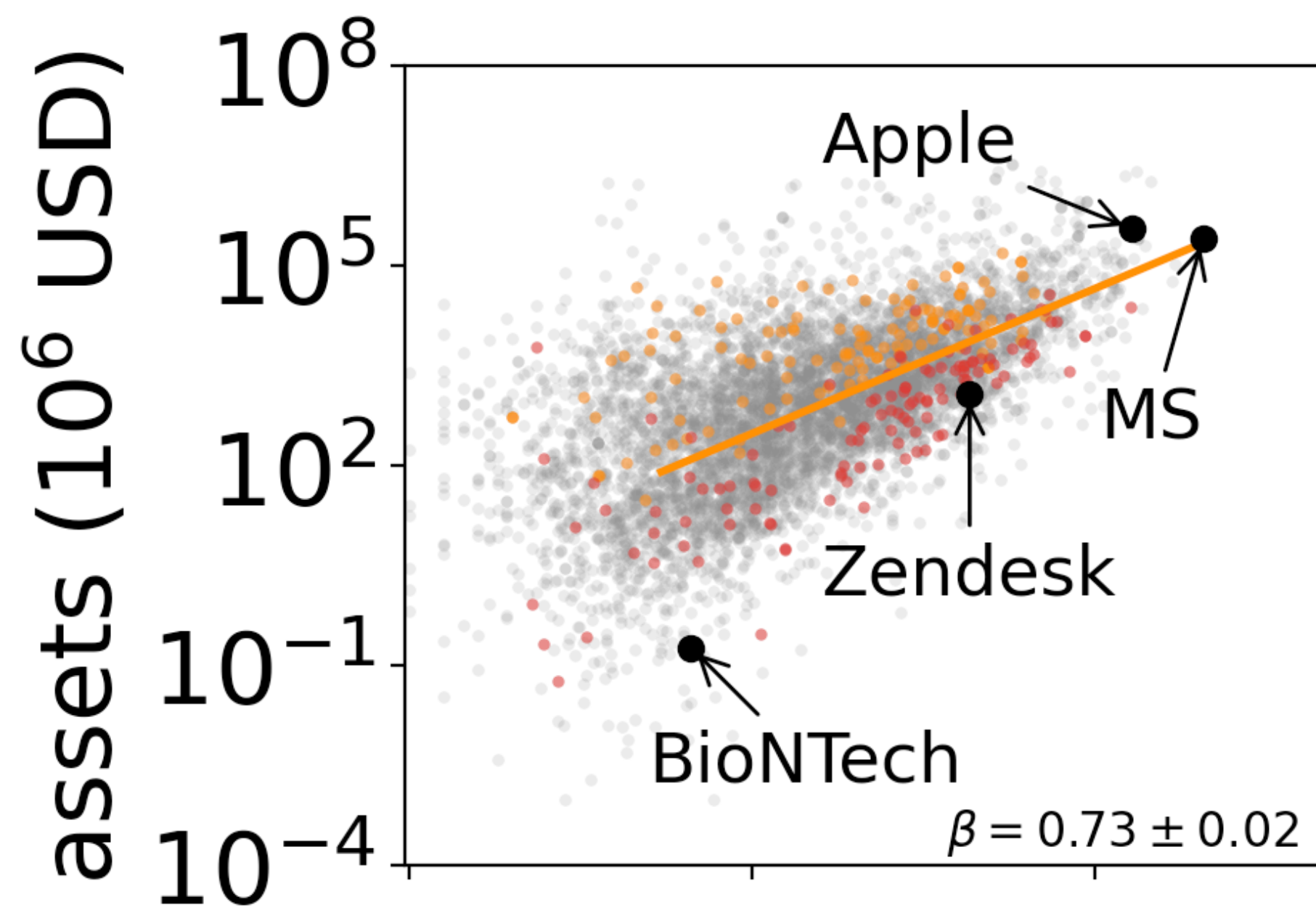
Social media
Discipline
Vacations
3D animation software
Trade notes
Blu-ray
Globalization
etc.



information economy of scale

Stanley et al. (1996), *Nature*
Axtell (2001), *Science*
Bettencourt et al. (2007), *PNAS*
Gabaix (2009), *Annual Review of Economics*
Zhang, Kempes, & West (2022), preprint
etc.

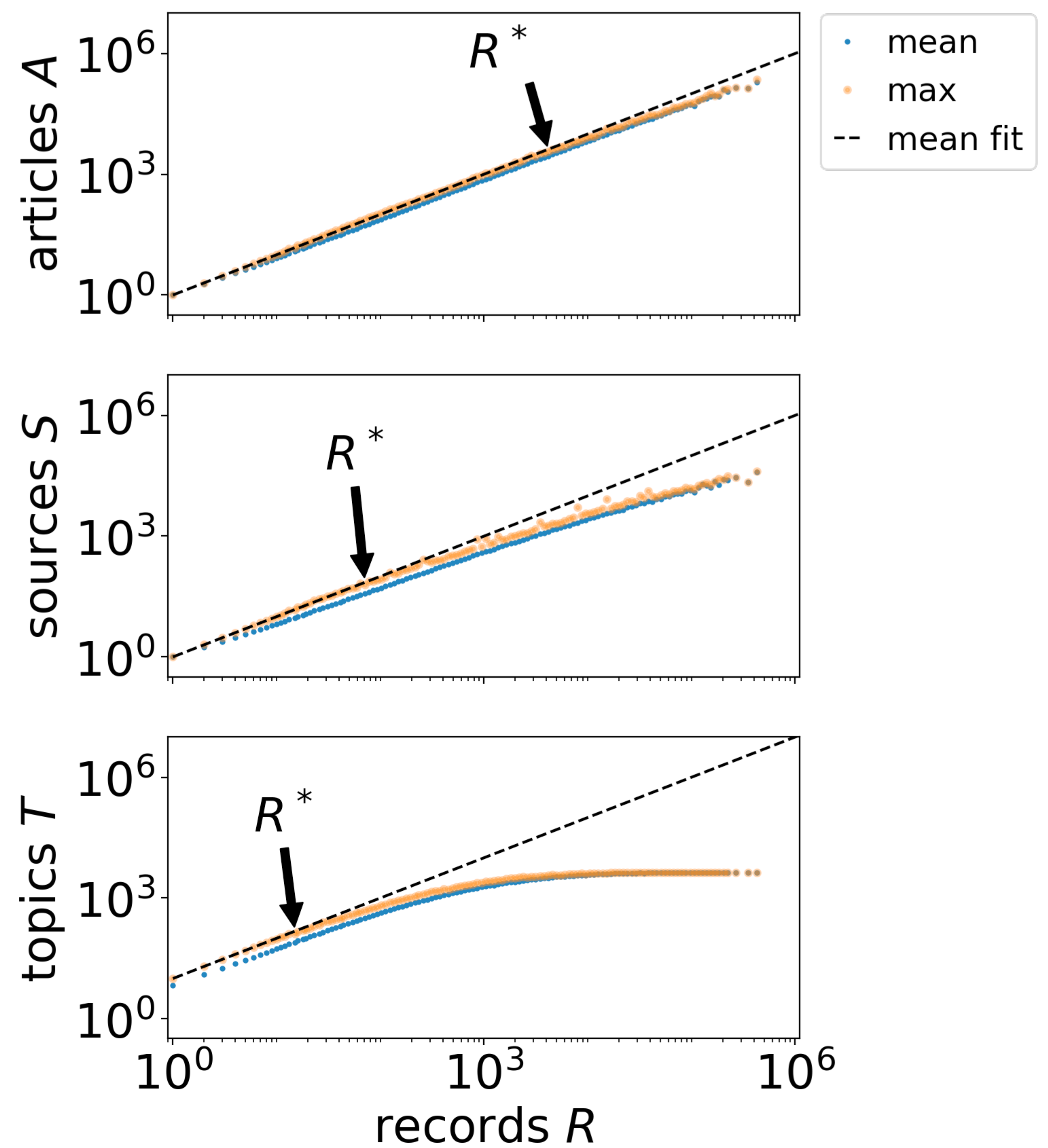




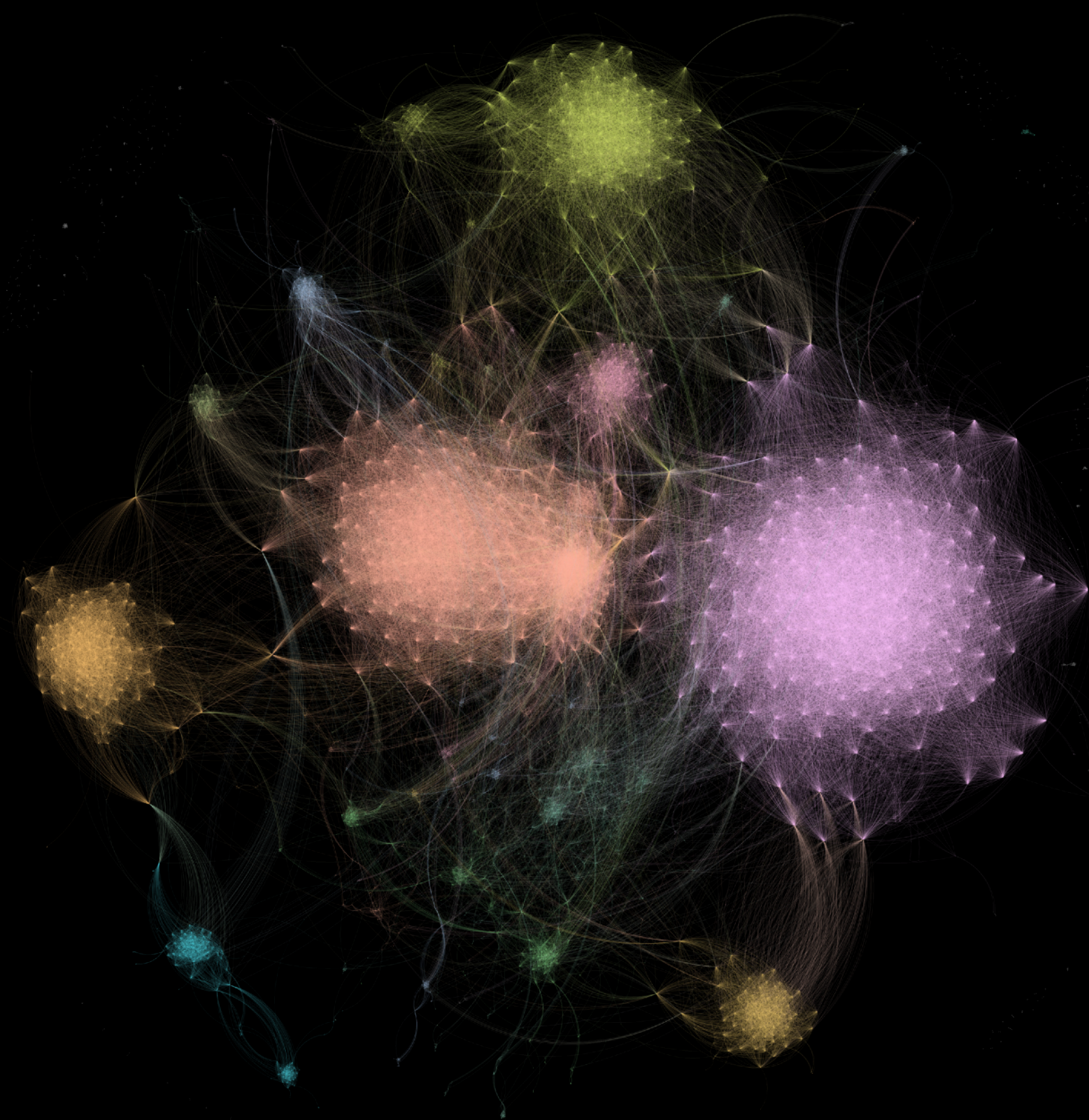
Deviations from baseline strongly correlated with future valuation and contemporaneous returns

	Tobin Q		Return	
	(5)	(6)	(7)	(8)
Constant	0.0005 (0.0837)		-0.4395*** (0.1196)	
Excess Reading (assets)	0.2926*** (0.0231)	0.2020*** (0.0263)	0.1970*** (0.0369)	0.1085** (0.0424)
Log assets	-0.0037 (0.0111)	-0.0017 (0.0111)	0.1491*** (0.0161)	0.1248*** (0.0167)
Observations	3,149	3,149	3,187	3,187
R ²	0.09974	0.15819	0.02669	0.06808
Within R ²		0.03870		0.02147
NAICS2 fixed effects		✓		✓

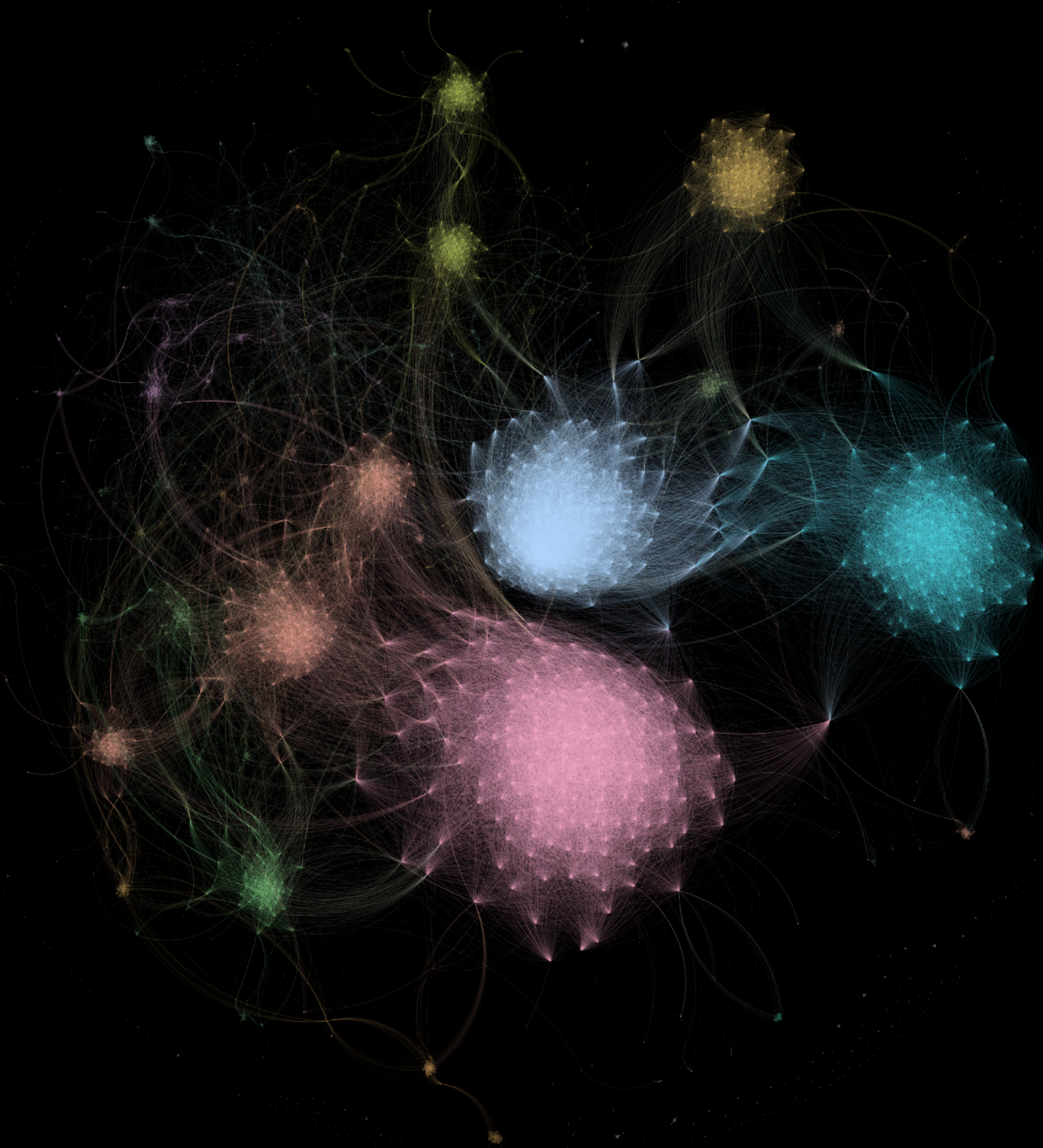
limited diversity of reading



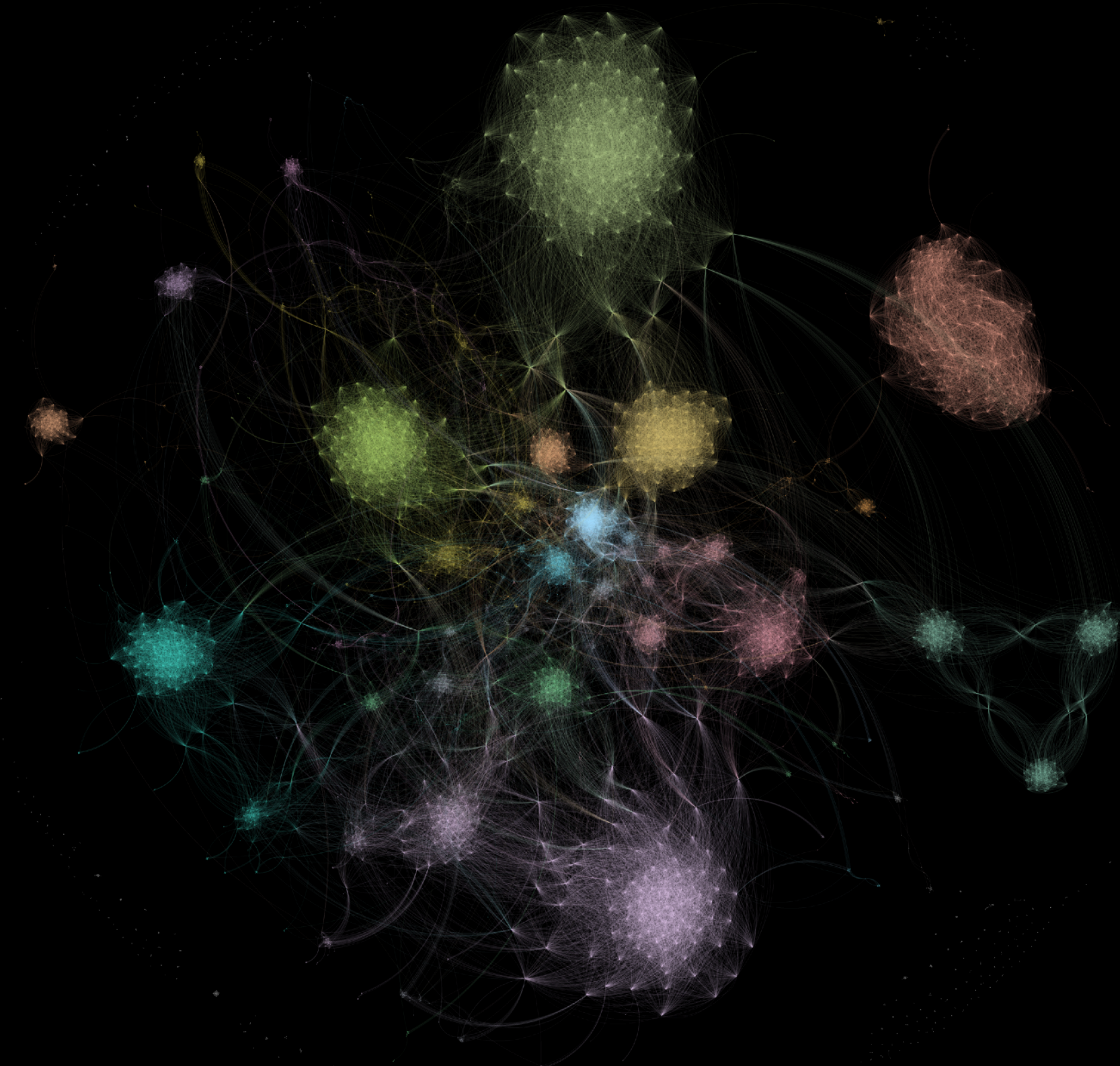
insurance

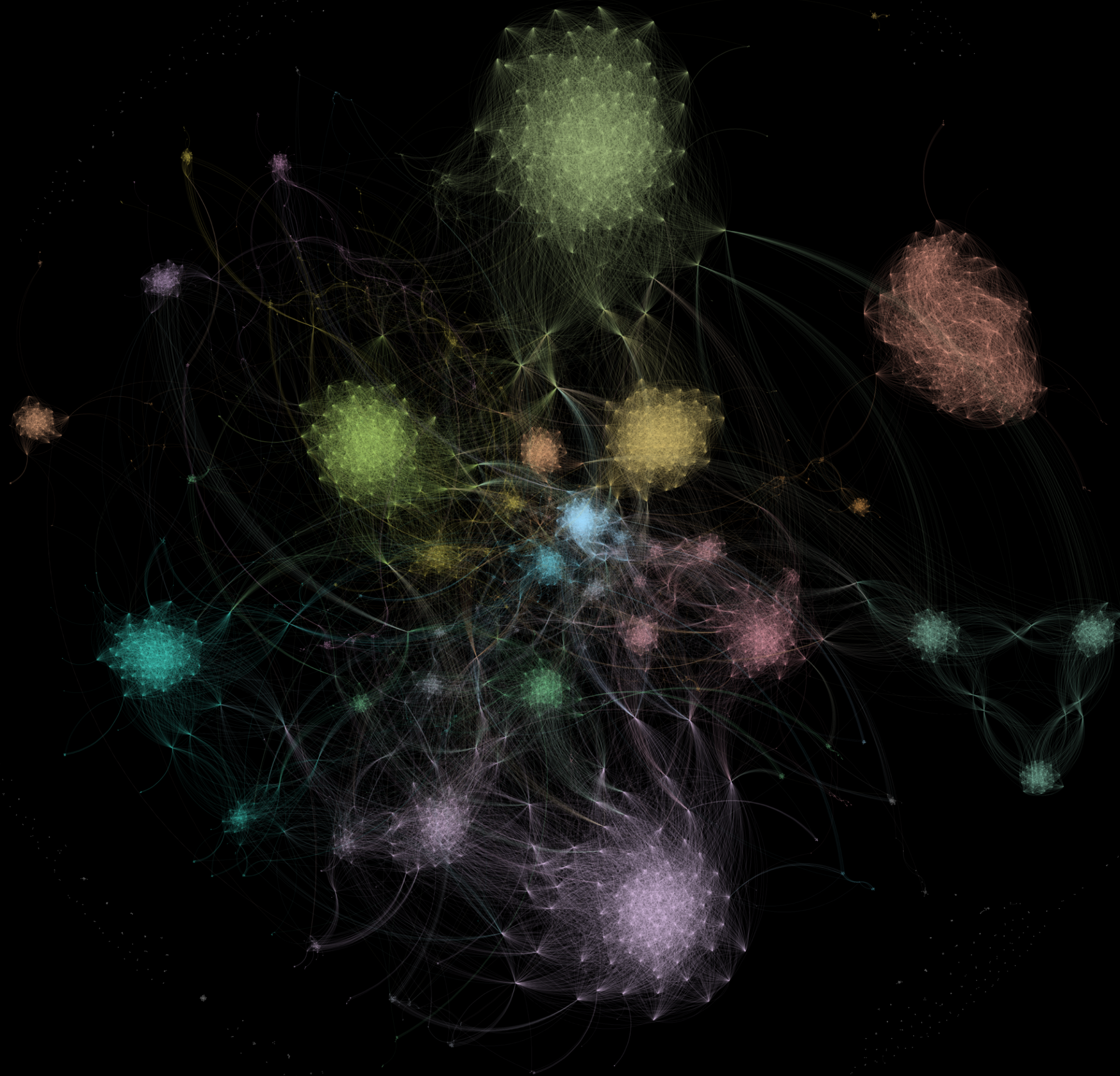


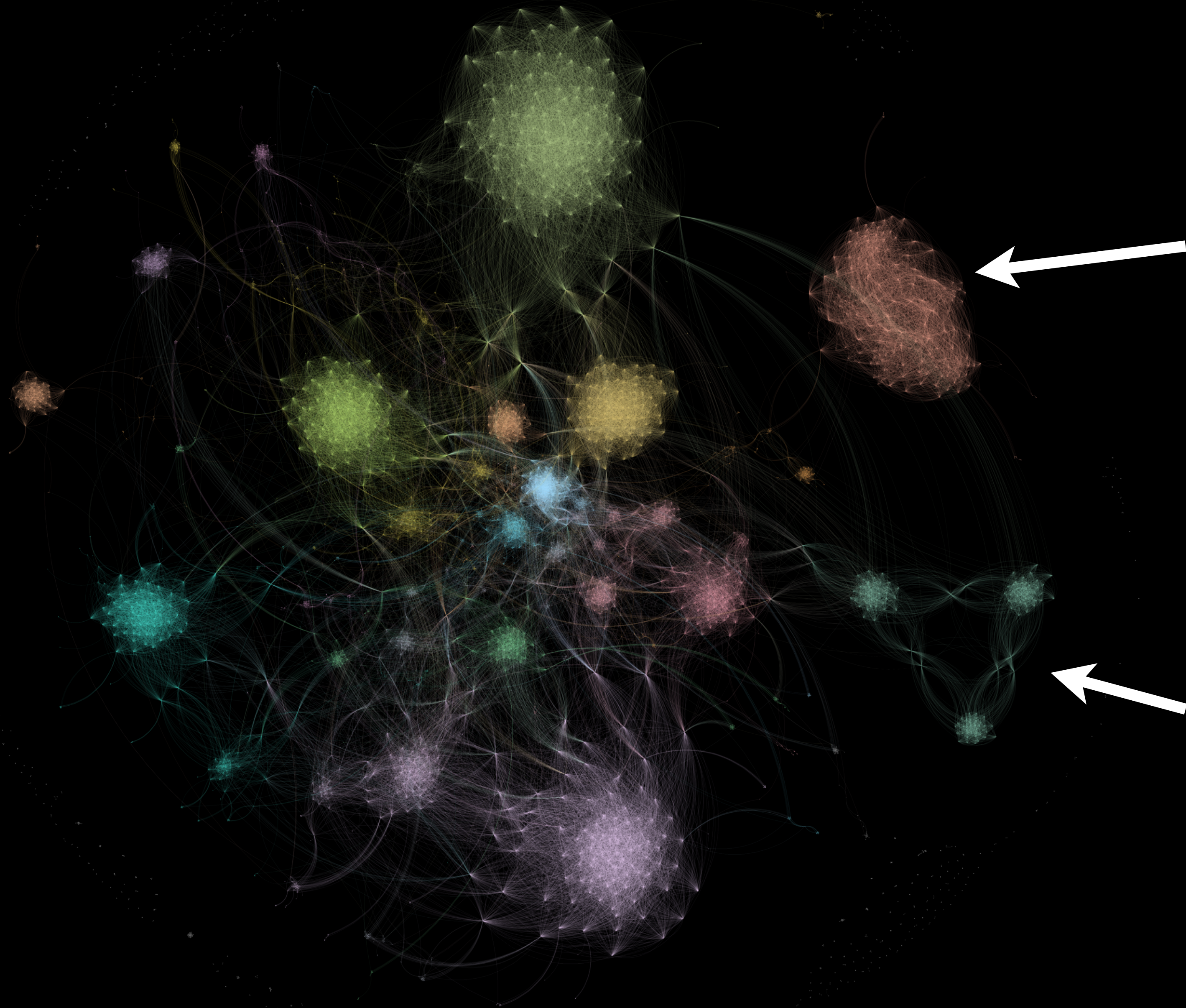
clothing



automobile



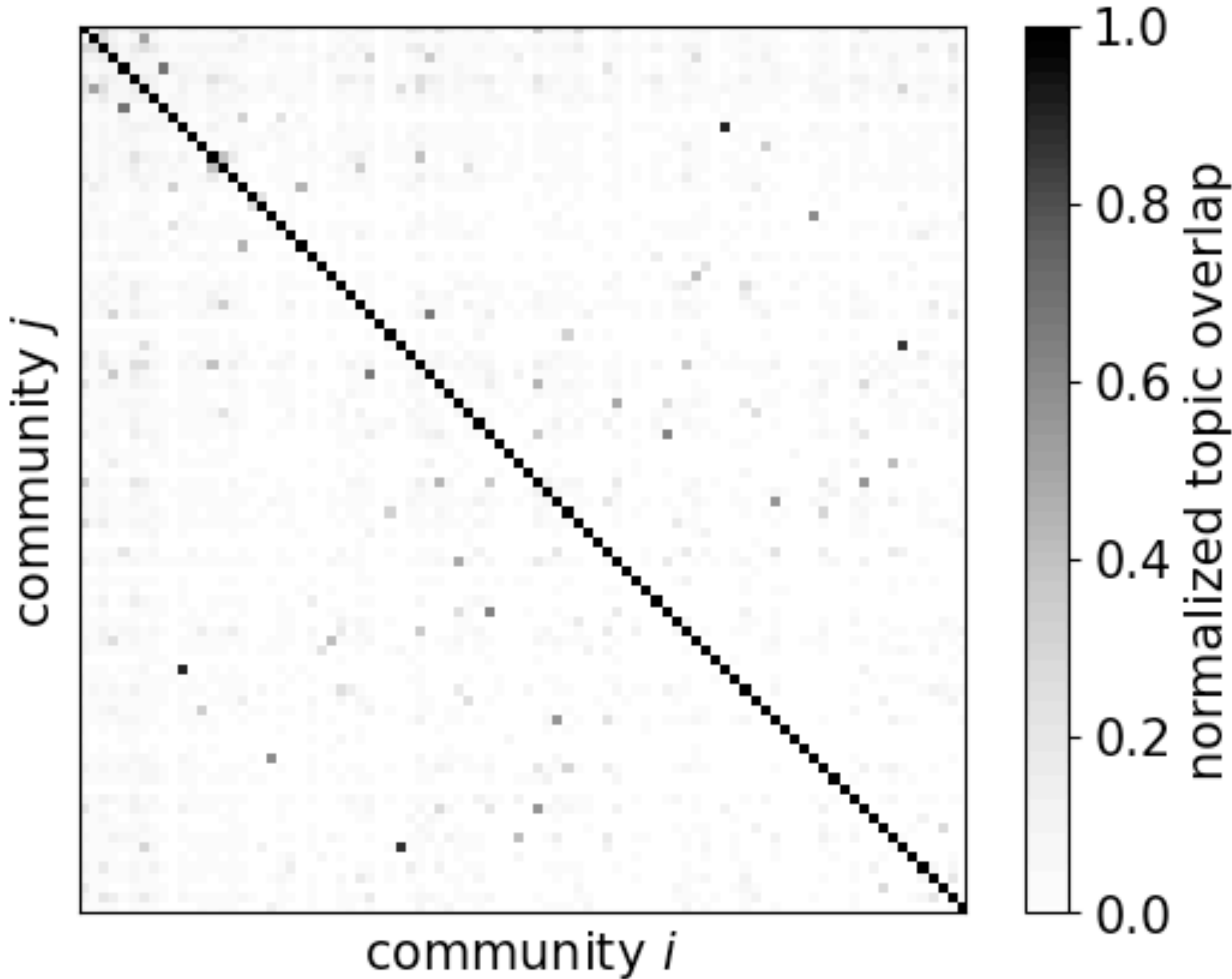




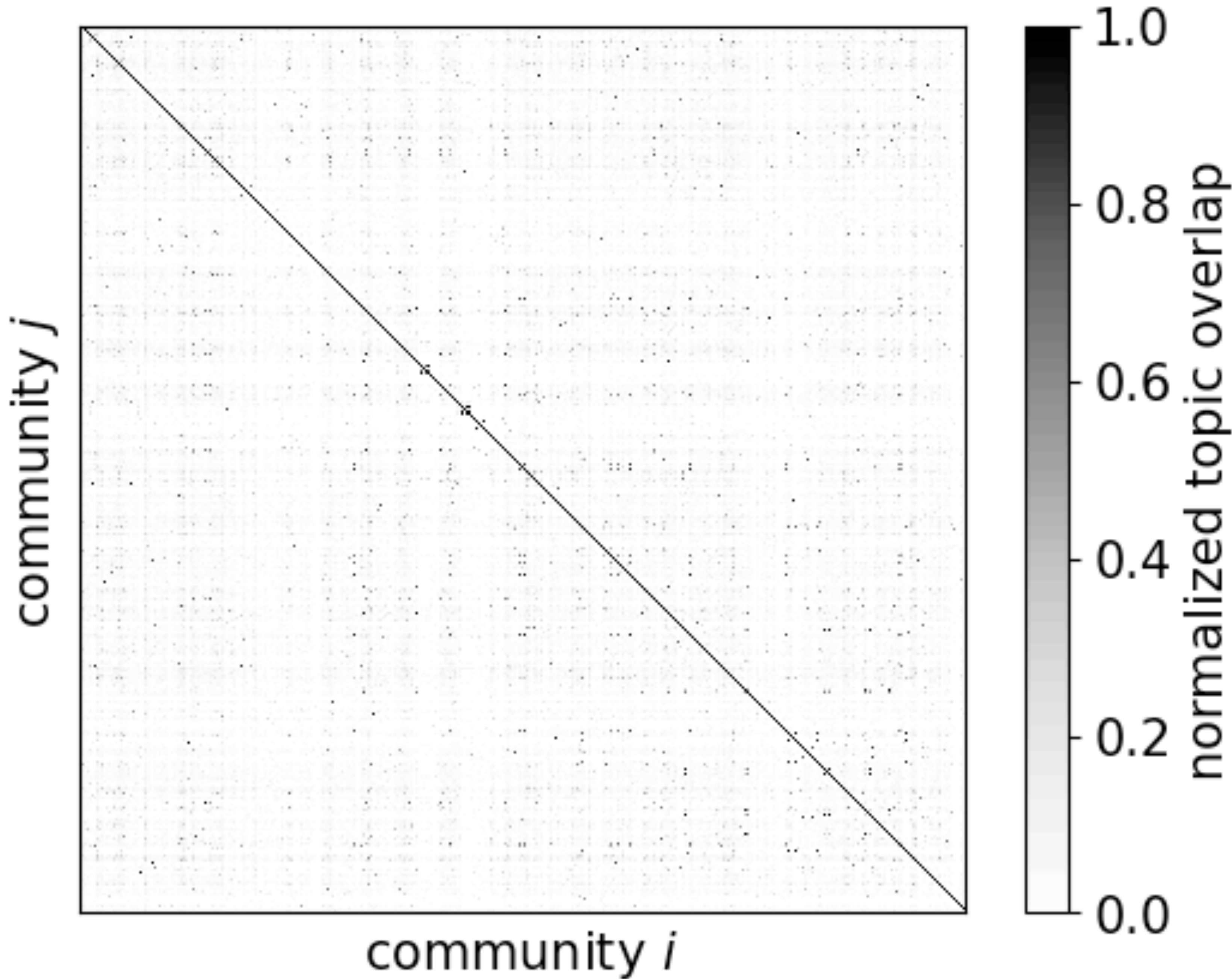
community A with
normalized topic vector
 \vec{t}_A

community B with
normalized topic vector
 \vec{t}_B

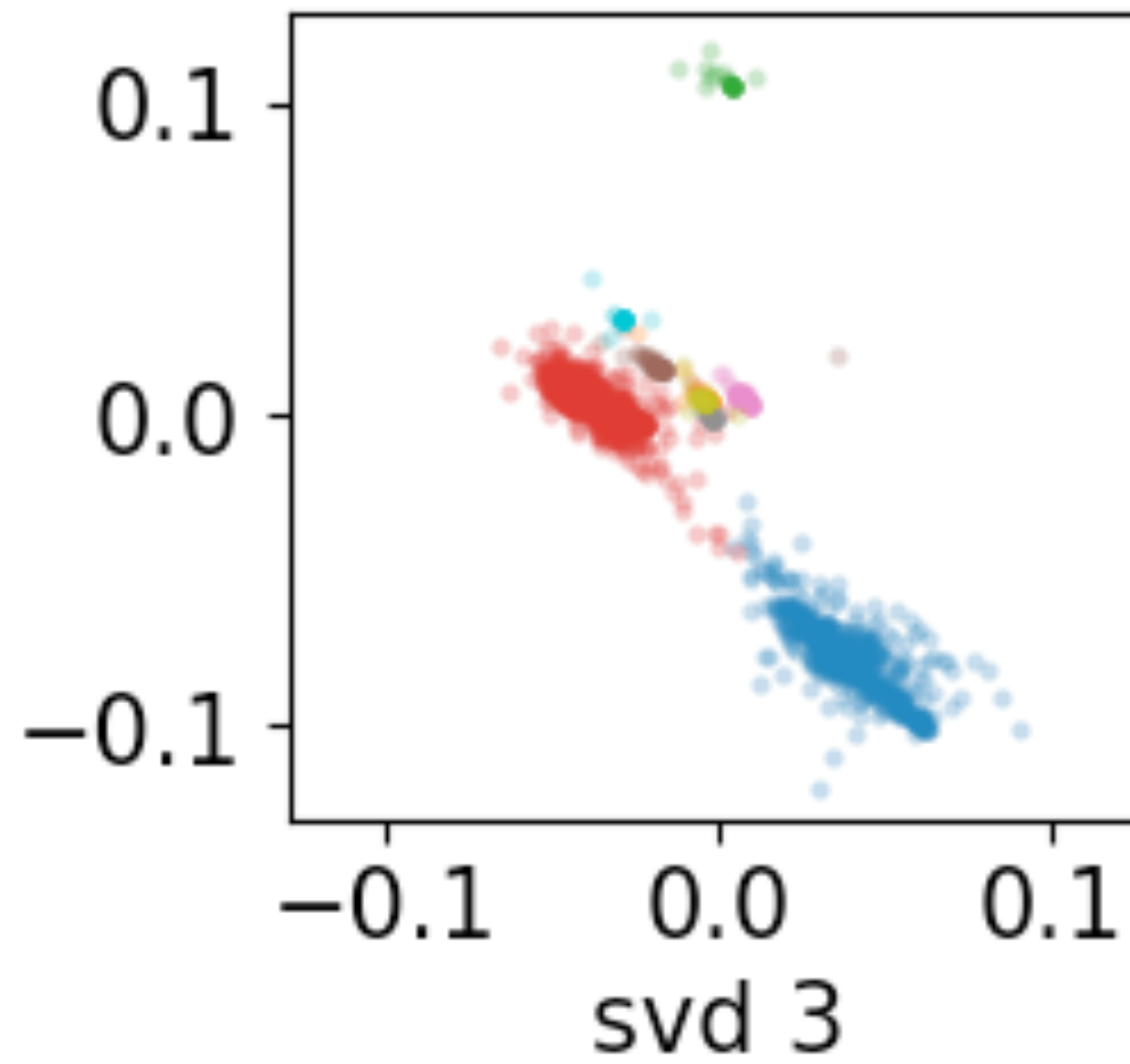
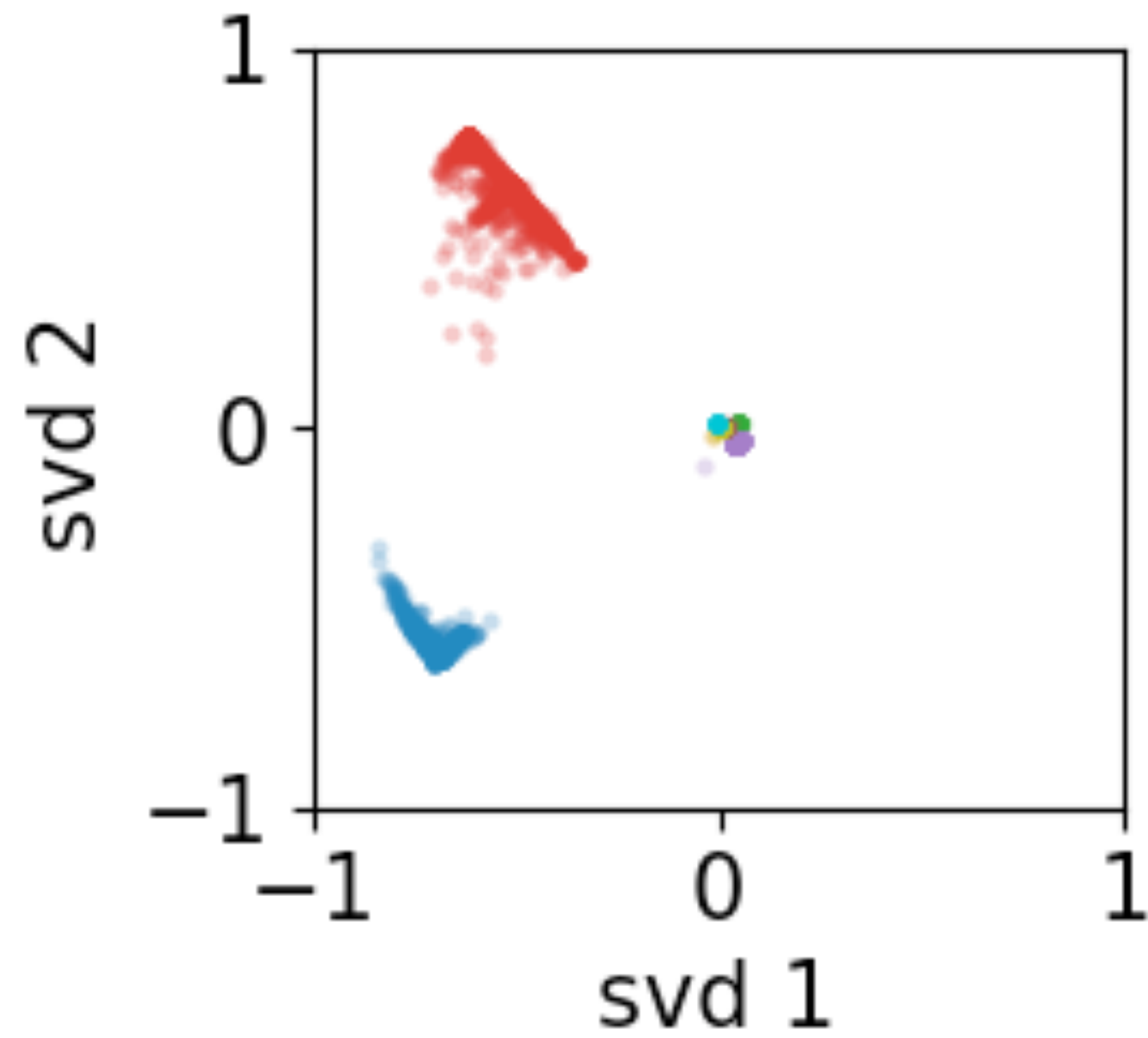
Similarity matrix of “departments” within an auto manufacturing firm



Similarity matrix of “departments” across many firms



Types of “departments”



Is science becoming less innovative?

American Economic Review 2020, 110(4): 1104–1144
<https://doi.org/10.1257/aer.20180338>

Are Ideas Getting Harder to Find?[†]

By NICHOLAS BLOOM, CHARLES I. JONES, JOHN VAN REENEN,
AND MICHAEL WEBB*

Long-run growth in many models is the product of two terms: the effective number of researchers and their research productivity. We present evidence from various fields showing that research effort is rising, while research productivity is declining sharply. The increasing number of researchers requires a growing number of ideas, but the increasing number of researchers also requires a growing number of computer chips, which increases the number of ideas required in the early stages of research. In the long run, we look we find that ideas, and the research effort required to find them, are getting harder to find. (JEL D2)

Monopolistic progress in large fields of science

James A. Evans^{b,c,d} 

^aNorthwestern University, Evanston, IL, 60208; ^bDepartment of Sociology, University of Chicago, Chicago, IL, 60637; ^cUniversity of Chicago, Chicago, IL, 60637; and ^dSanta Fe Institute, Santa Fe, NM, 87501

University of California, Berkeley, CA, and approved August 25, 2021 (received for review December 8, 2020)

number of papers published each year over time. Policy measures aim to increase research funding, and scientific progress is determined by the number of papers produced. The career trajectories of academic departments, institutions, and individuals are determined by how these increases in the numbers of

causing faster turnover of field paradigms, a deluge of new publications entrenches top-cited papers, precluding new work from rising into the most-cited, commonly known canon of the field.

These arguments, supported by our empirical analysis, suggest that the scientific enterprise's focus on quantity may obstruct fundamental progress.

This detrimental effect will intensify as the number of publications in each field continues to grow—unavoidable given the entrenched, interlocking nature of scientific publication quantity. Policy measures

Article

Papers and patents are becoming less disruptive over time


<https://doi.org/10.1038/s41586-022-05543-x>

Michael Park¹, Erin Leahey² & Russell J. Funk^{1✉}

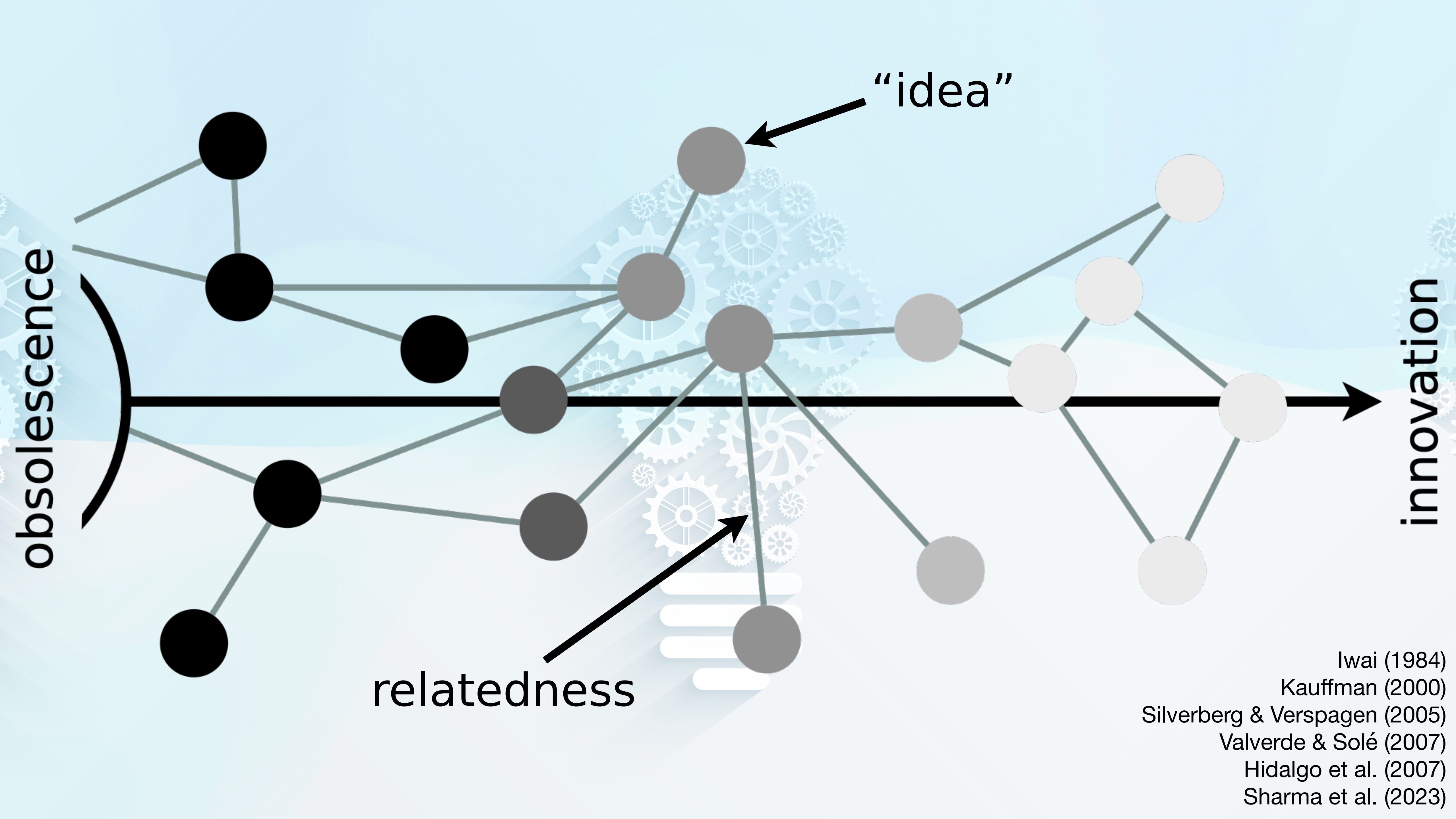
Received: 14 February 2022

Accepted: 8 November 2022

Published online: 4 January 2023

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Theories of scientific and technological change view discovery and invention as endogenous processes^{1,2}, wherein previous accumulated knowledge enables future progress by allowing researchers to, in Newton's words, 'stand on the shoulders of giants'^{3–7}. Recent decades have witnessed exponential growth in the volume of new scientific and technological knowledge, thereby creating conditions that should be



Iwai (1984)

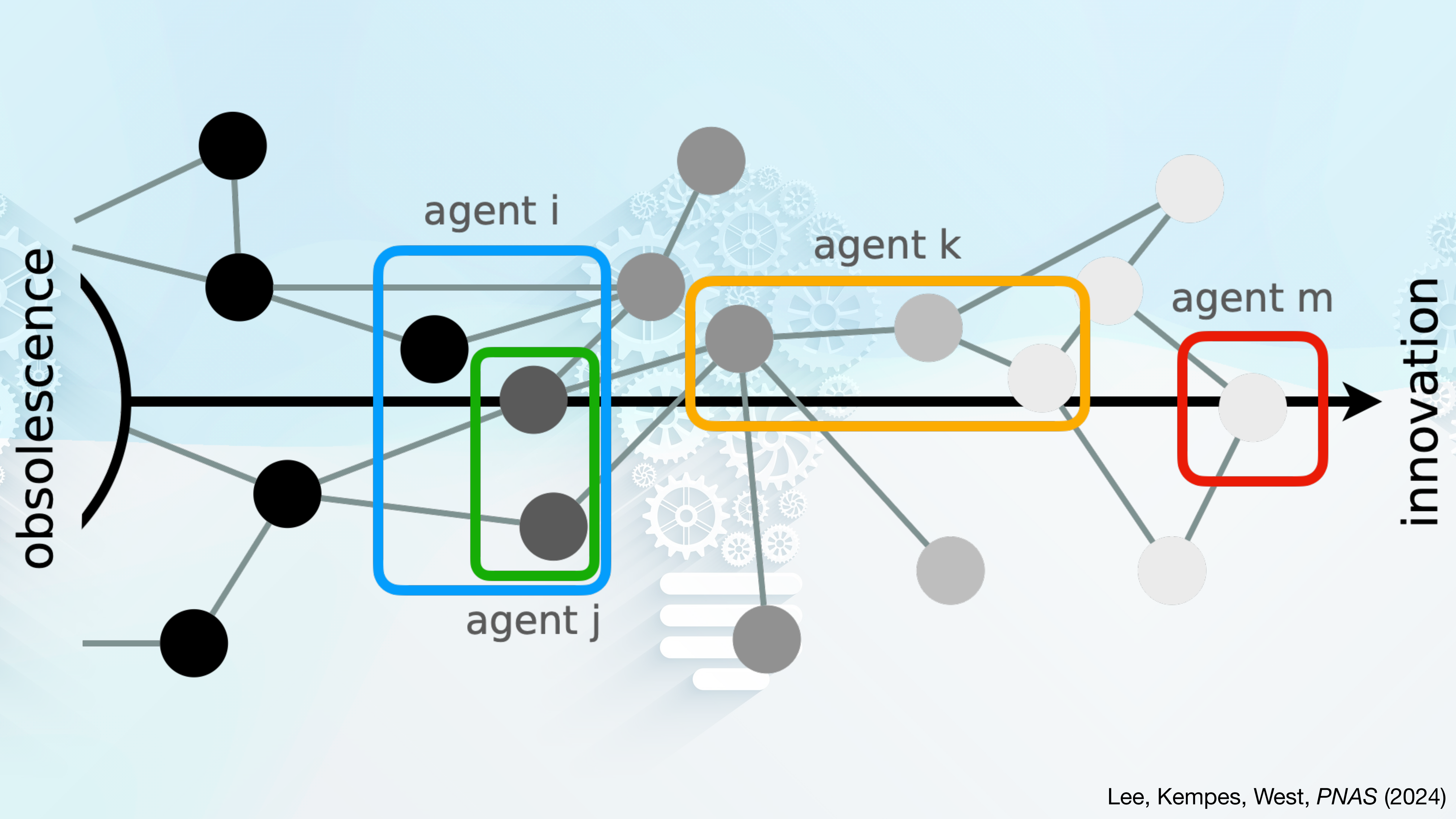
Kauffman (2000)

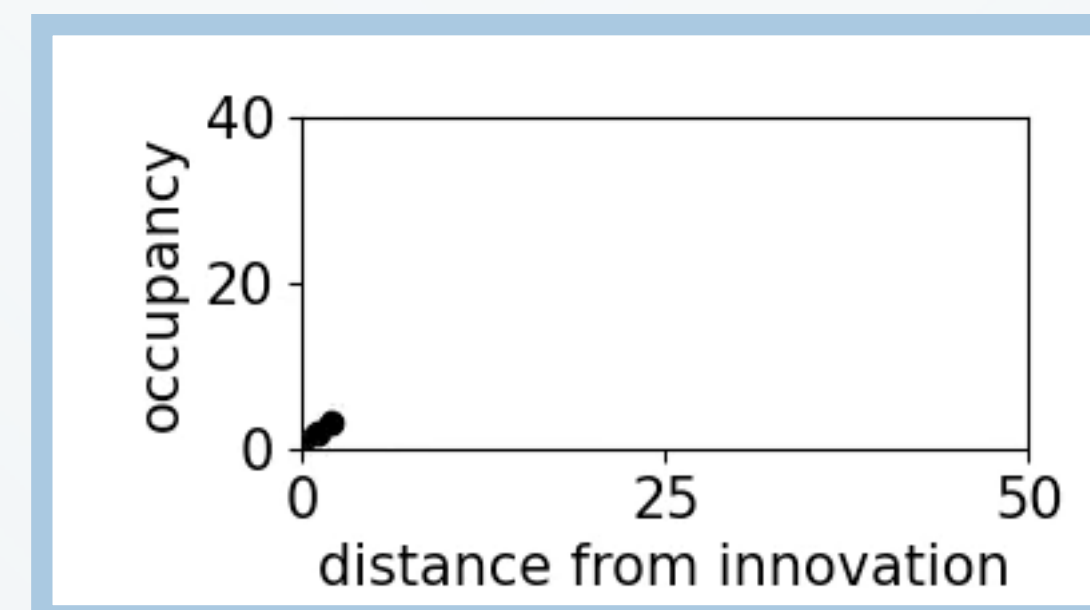
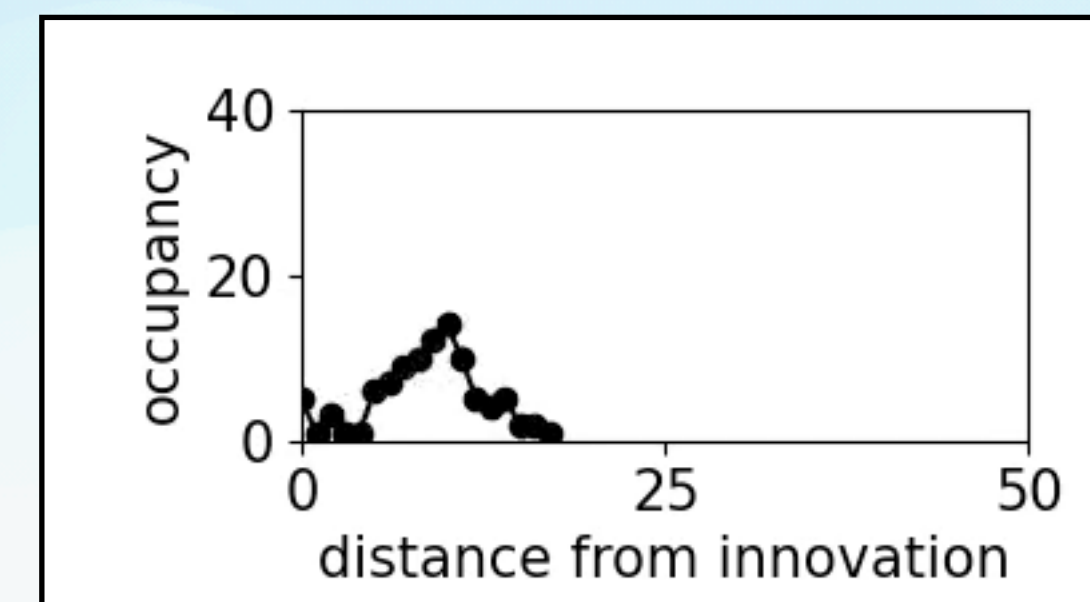
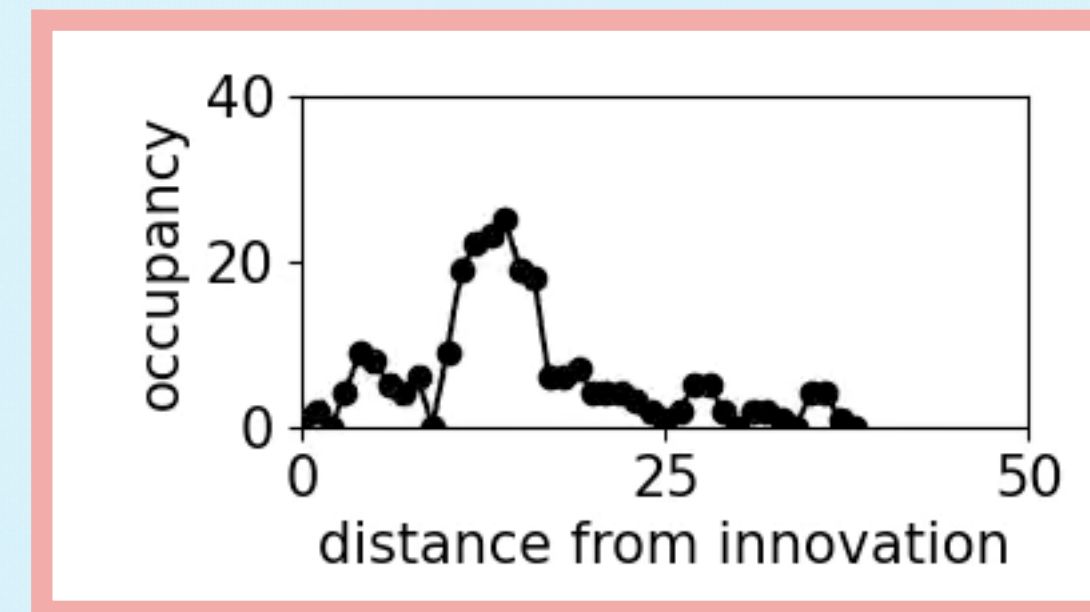
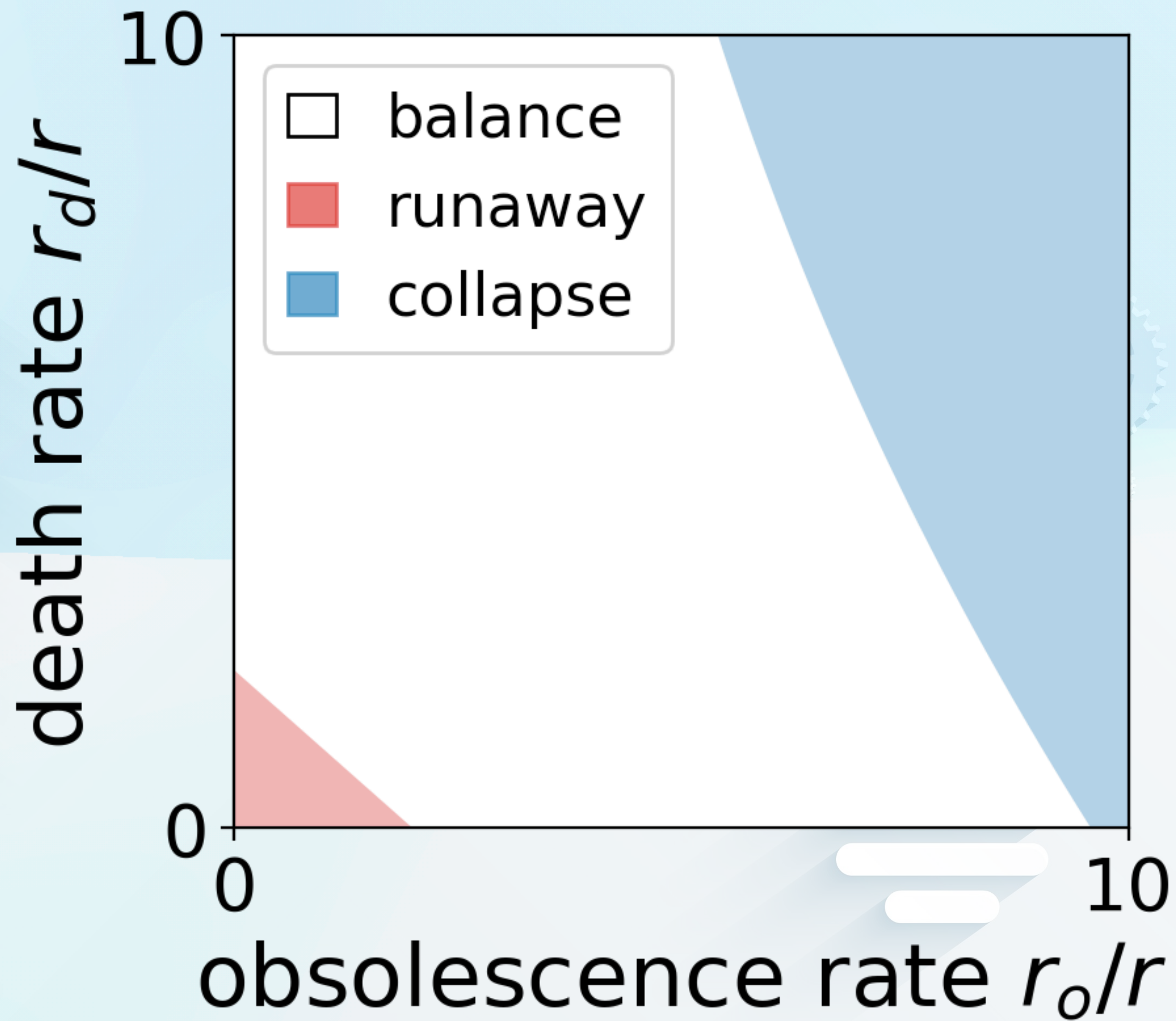
Silverberg & Verspagen (2005)

Valverde & Solé (2007)

Hidalgo et al. (2007)

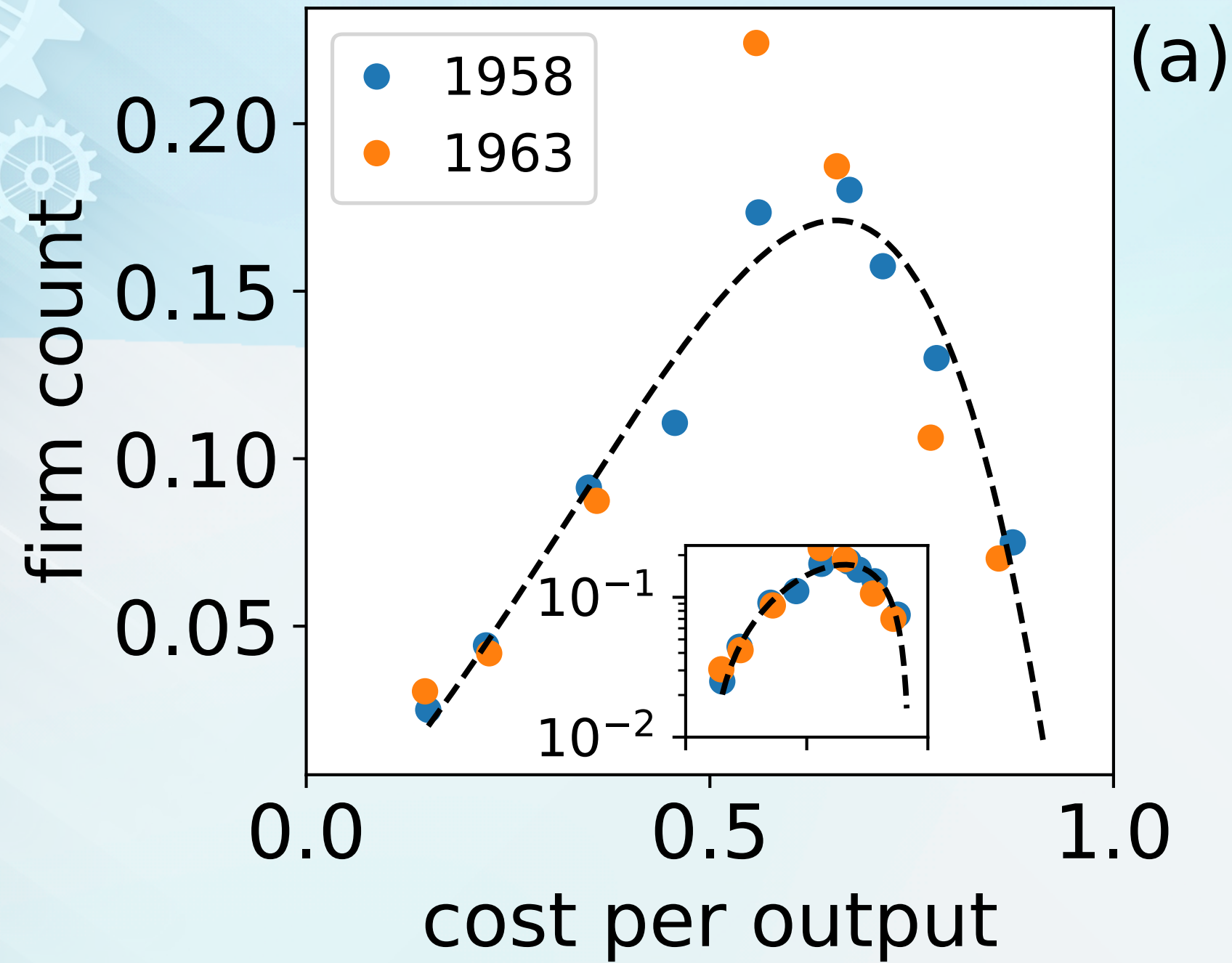
Sharma et al. (2023)



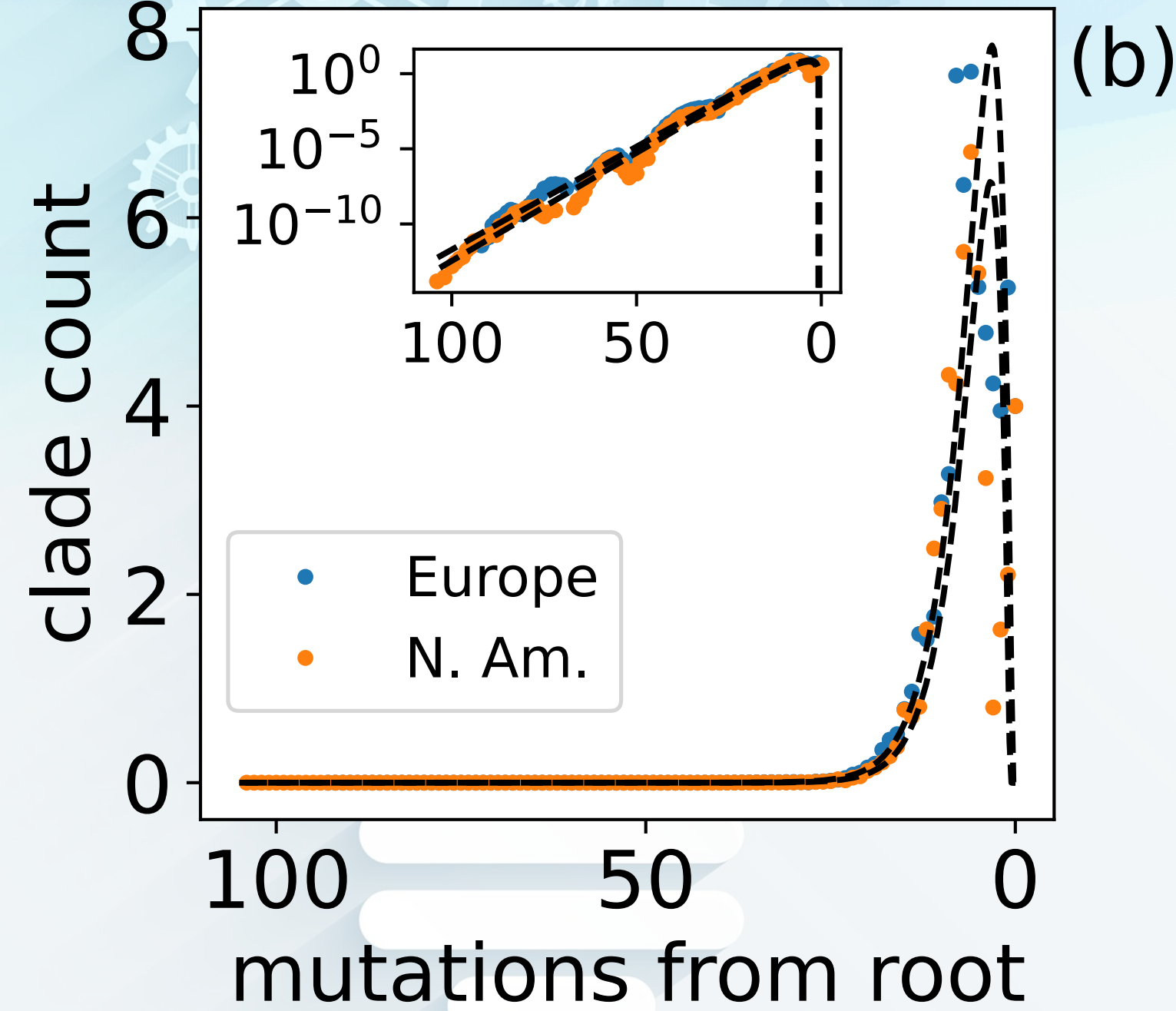


Predicting frequency of innovation agents across systems

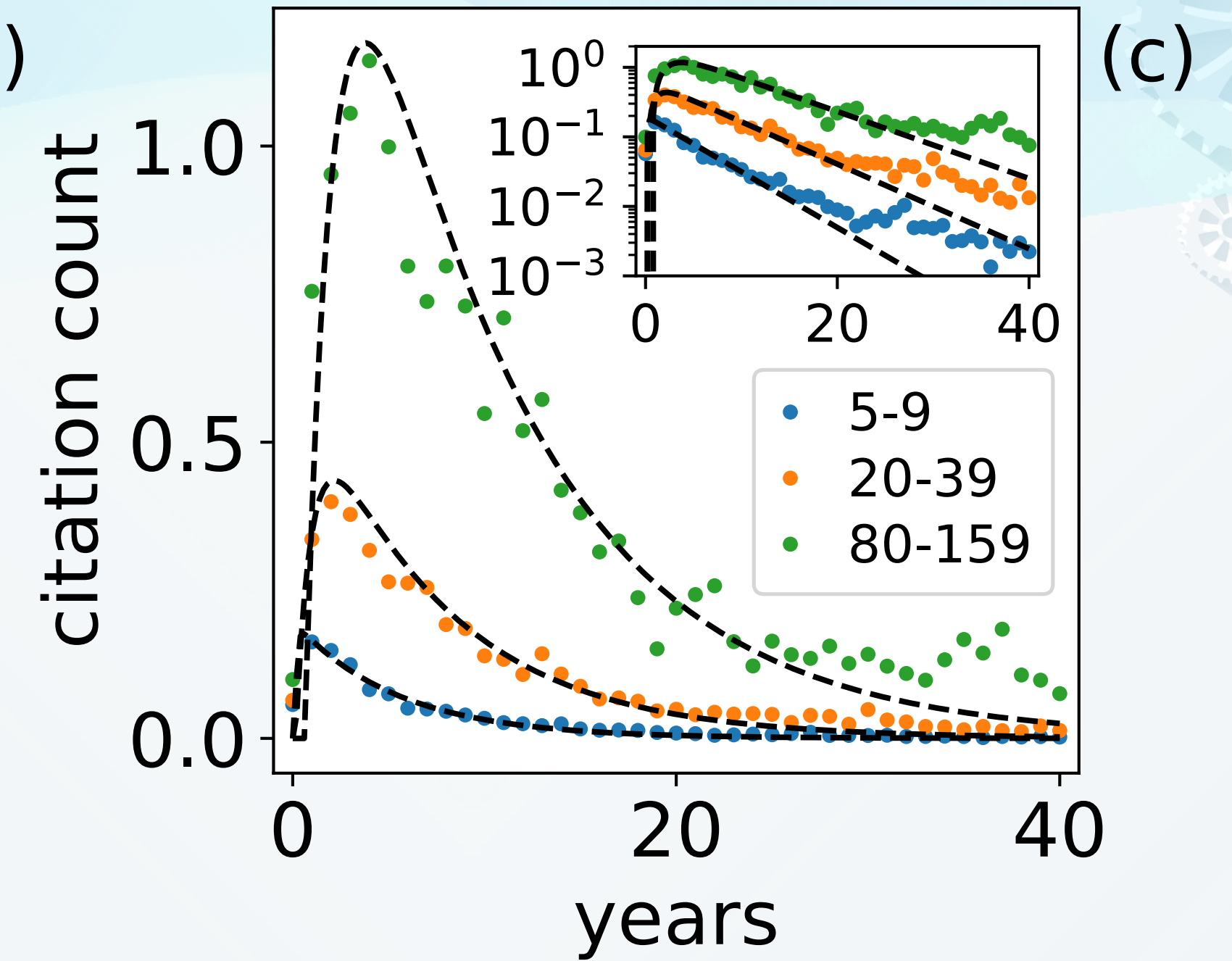
Schumpeterian



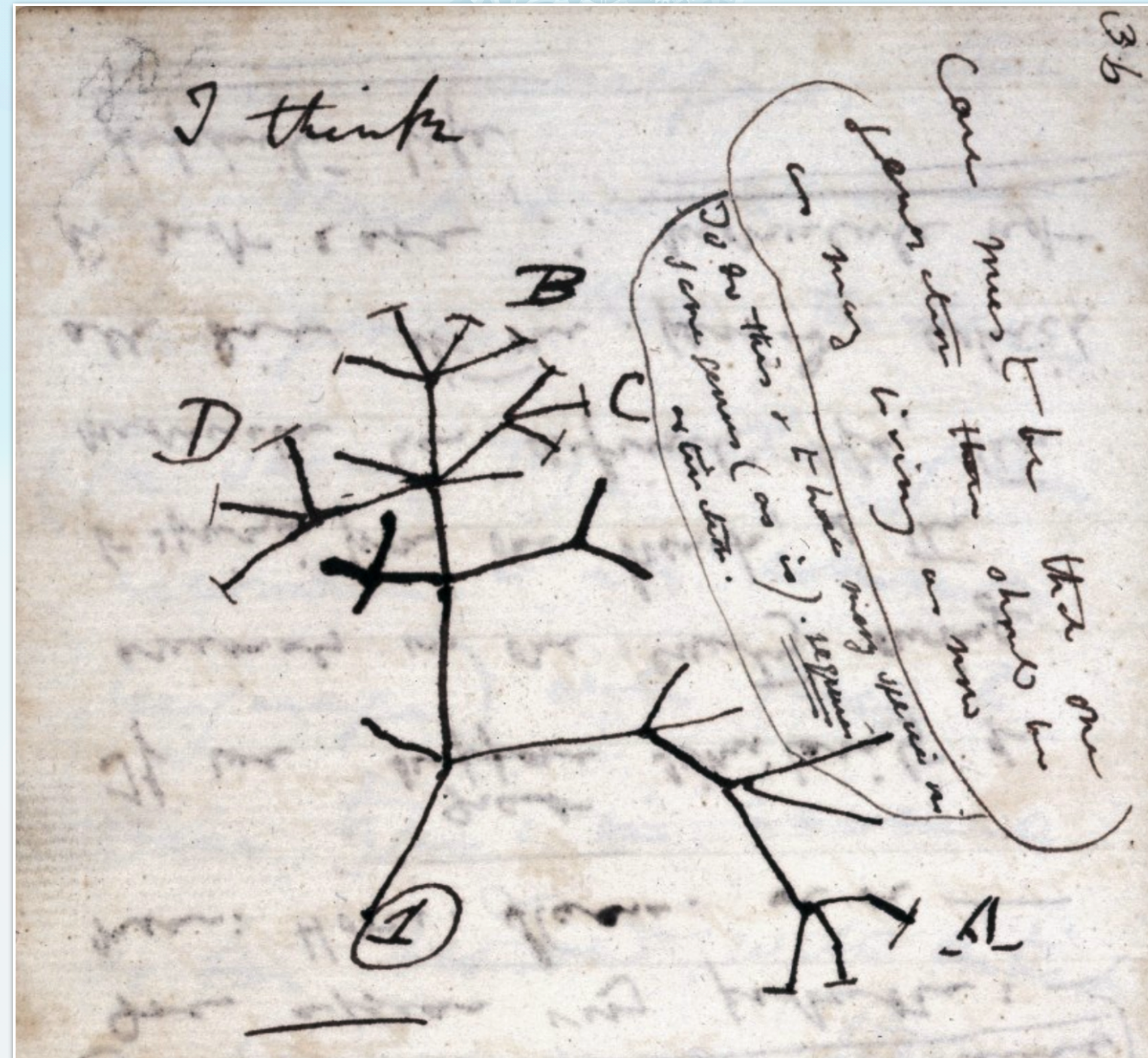
Darwinian



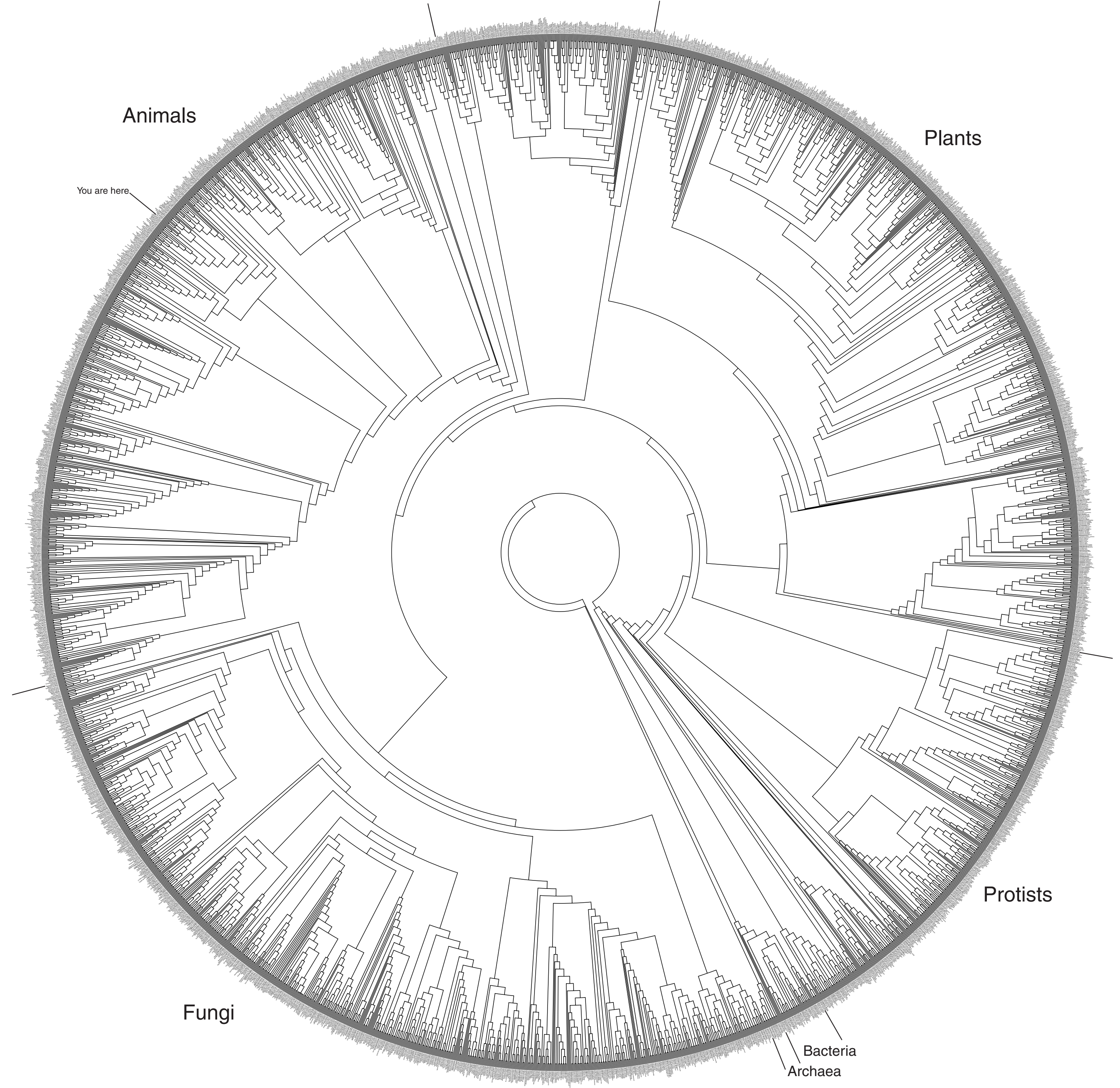
Socratic



Space of the possible

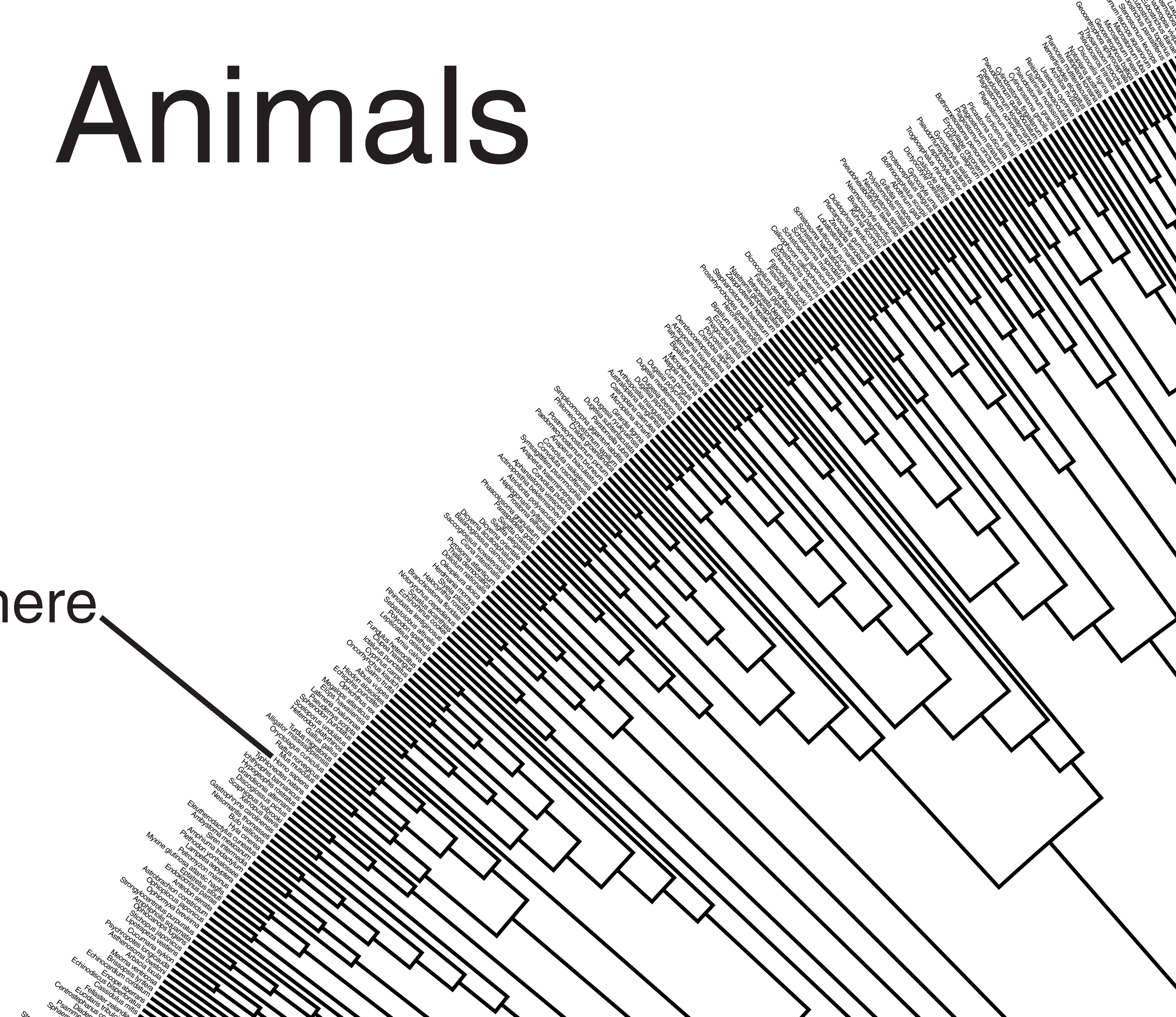


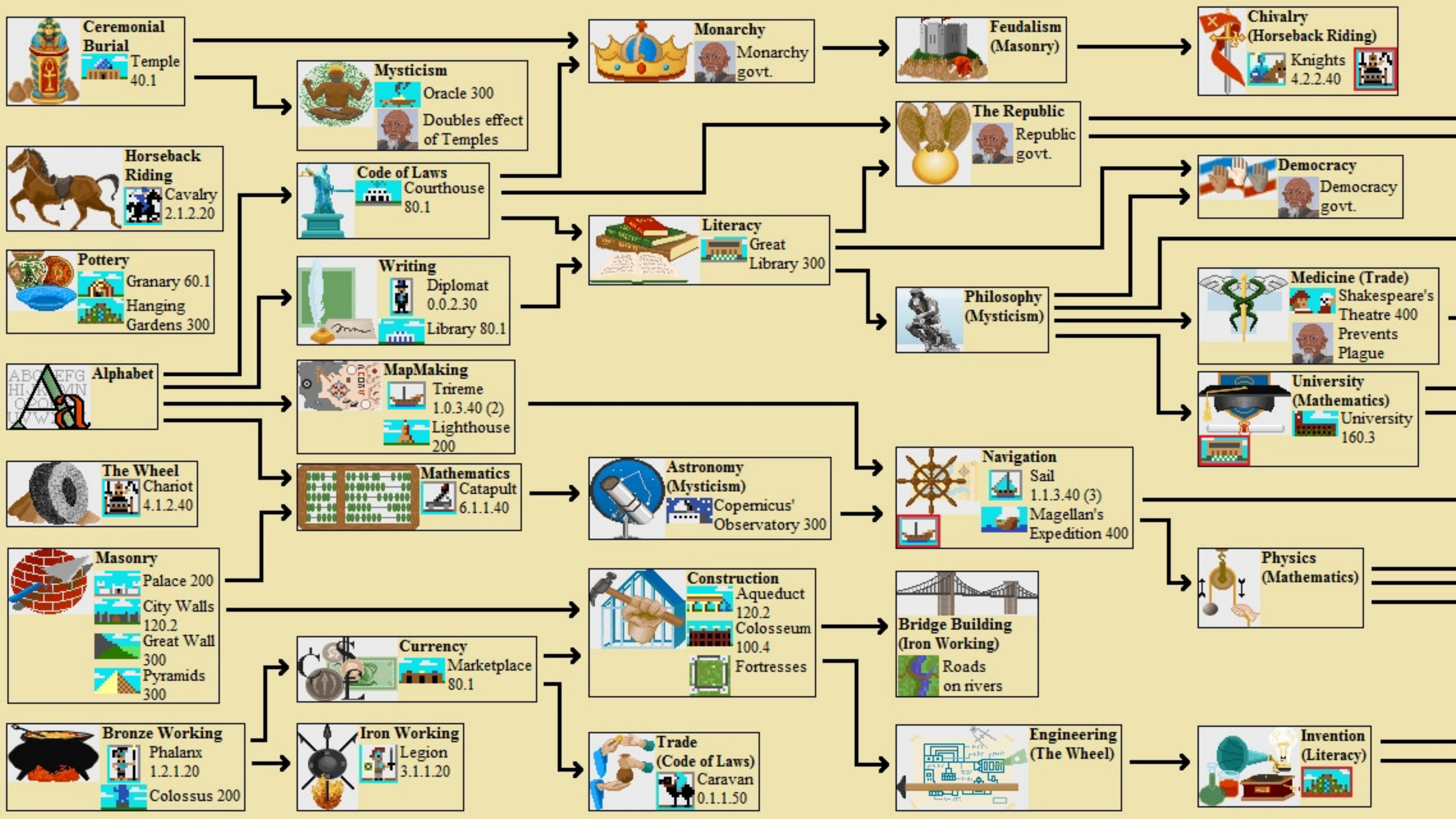
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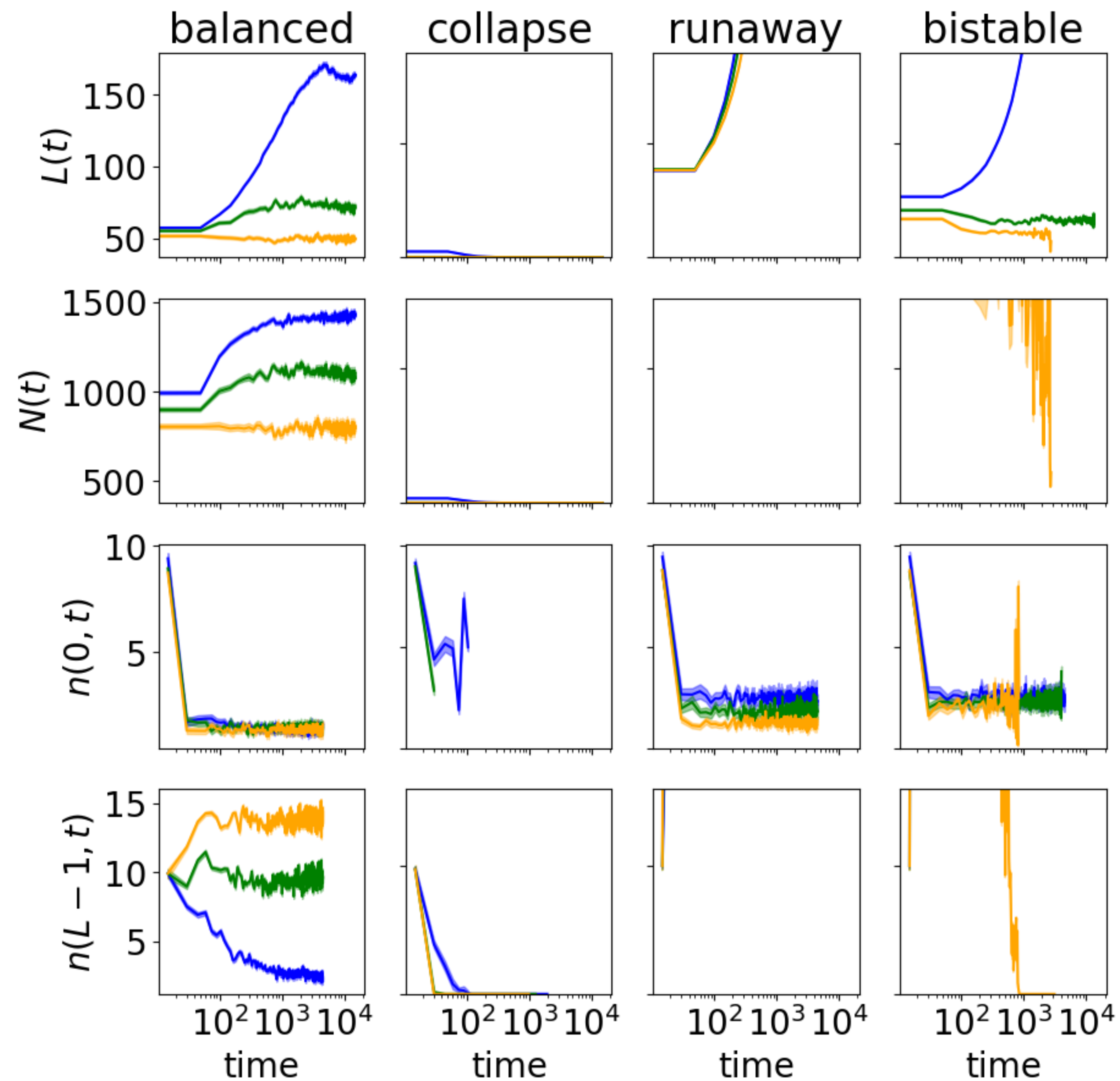
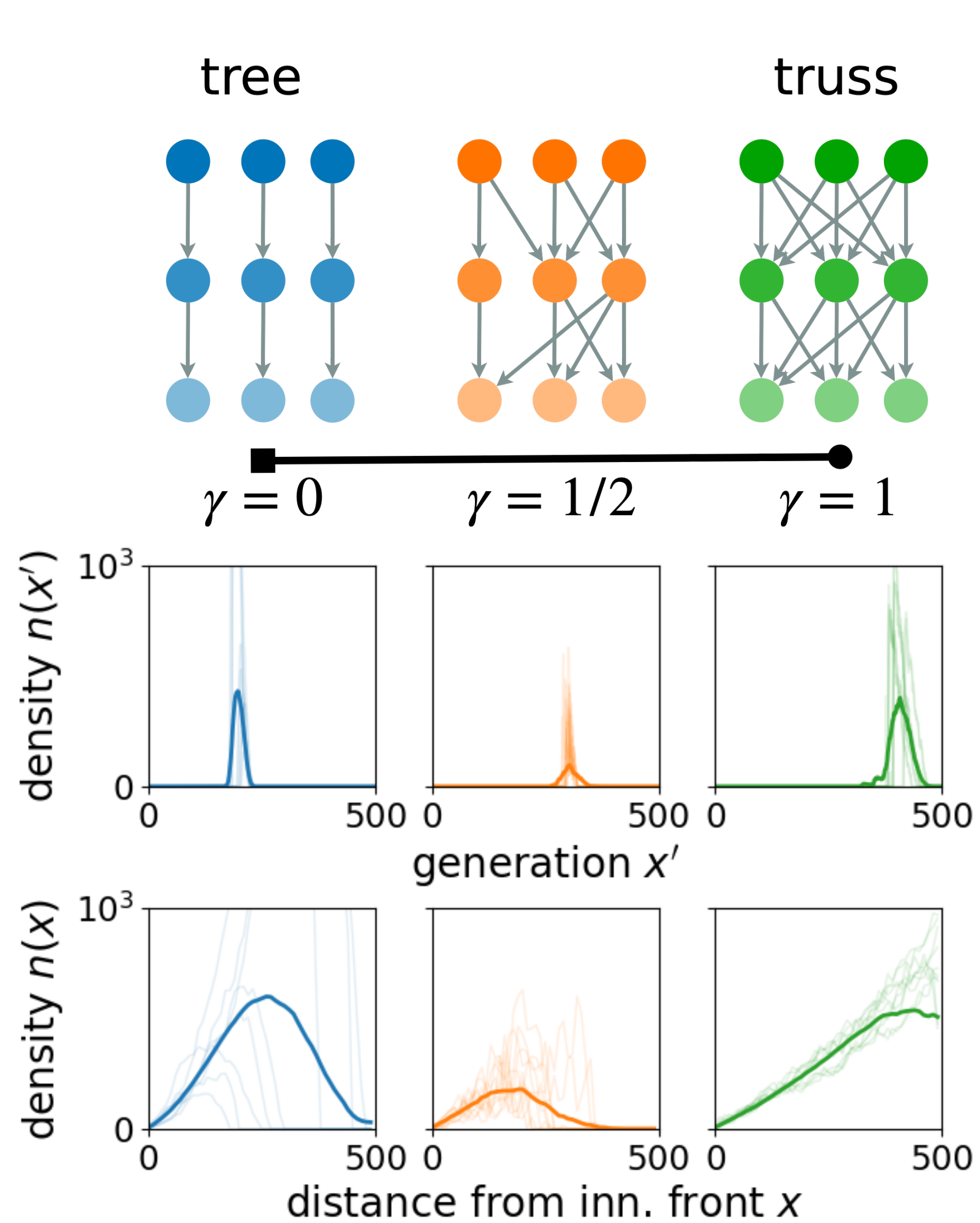
Animals

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Numerics and analytics



Questions?

More at <https://eddielee.co>

- 1. inference**
- 2. multiscale data analysis**
- 3. computation**
- 4. innovation & obsolescence**



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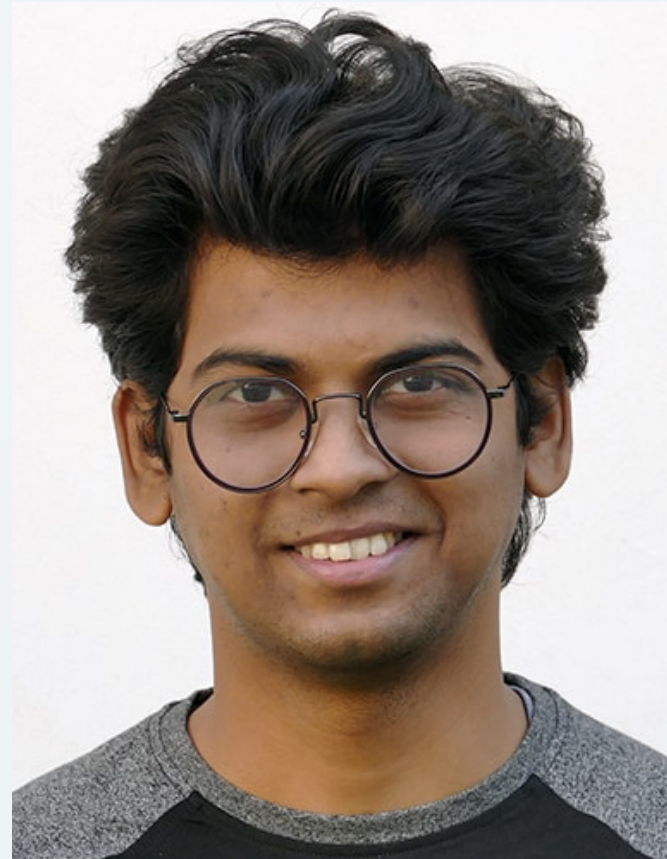
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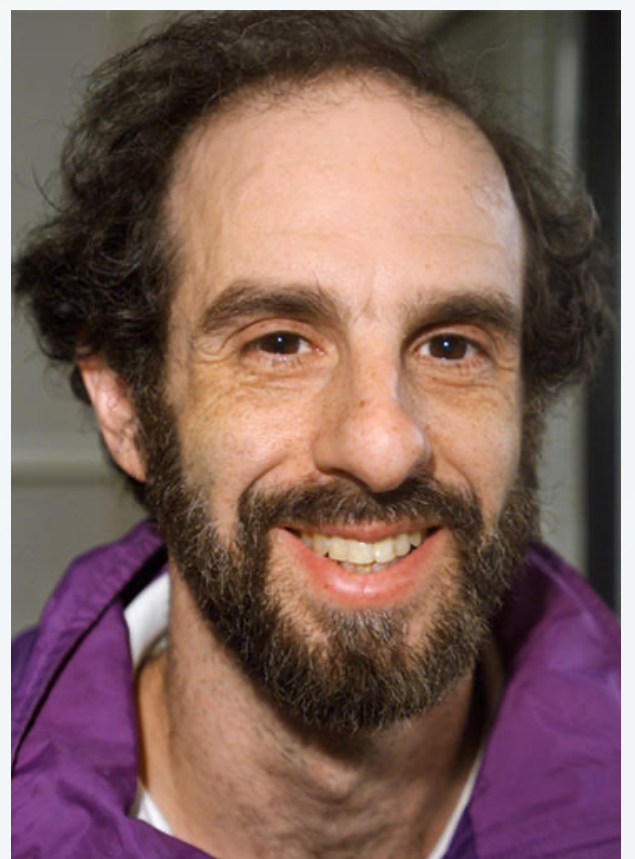
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