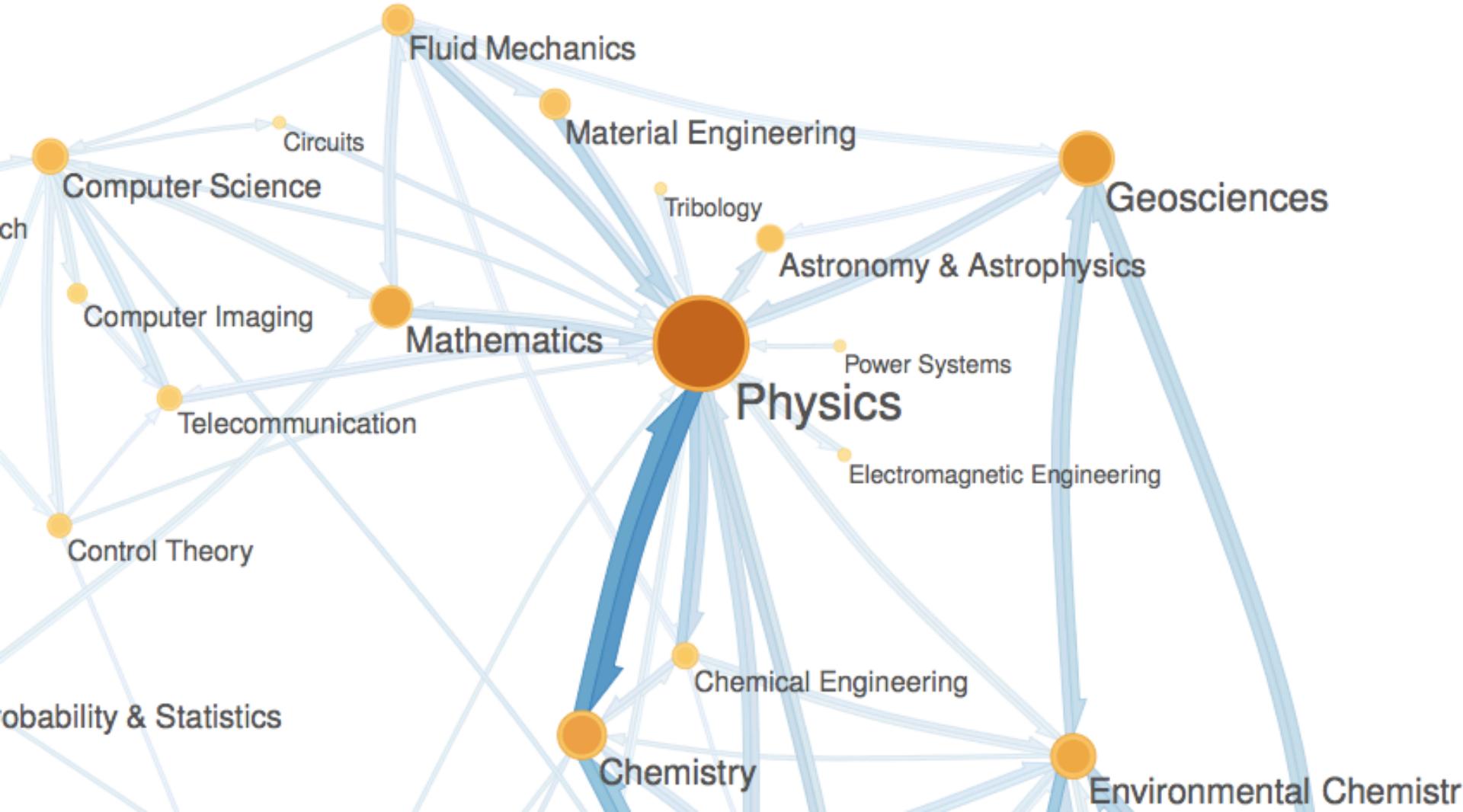


Mapping the HRA Community

Jevin West, Information School, University of Washington



What is my impact on science?

\$7,933,670,366





22,756 awards

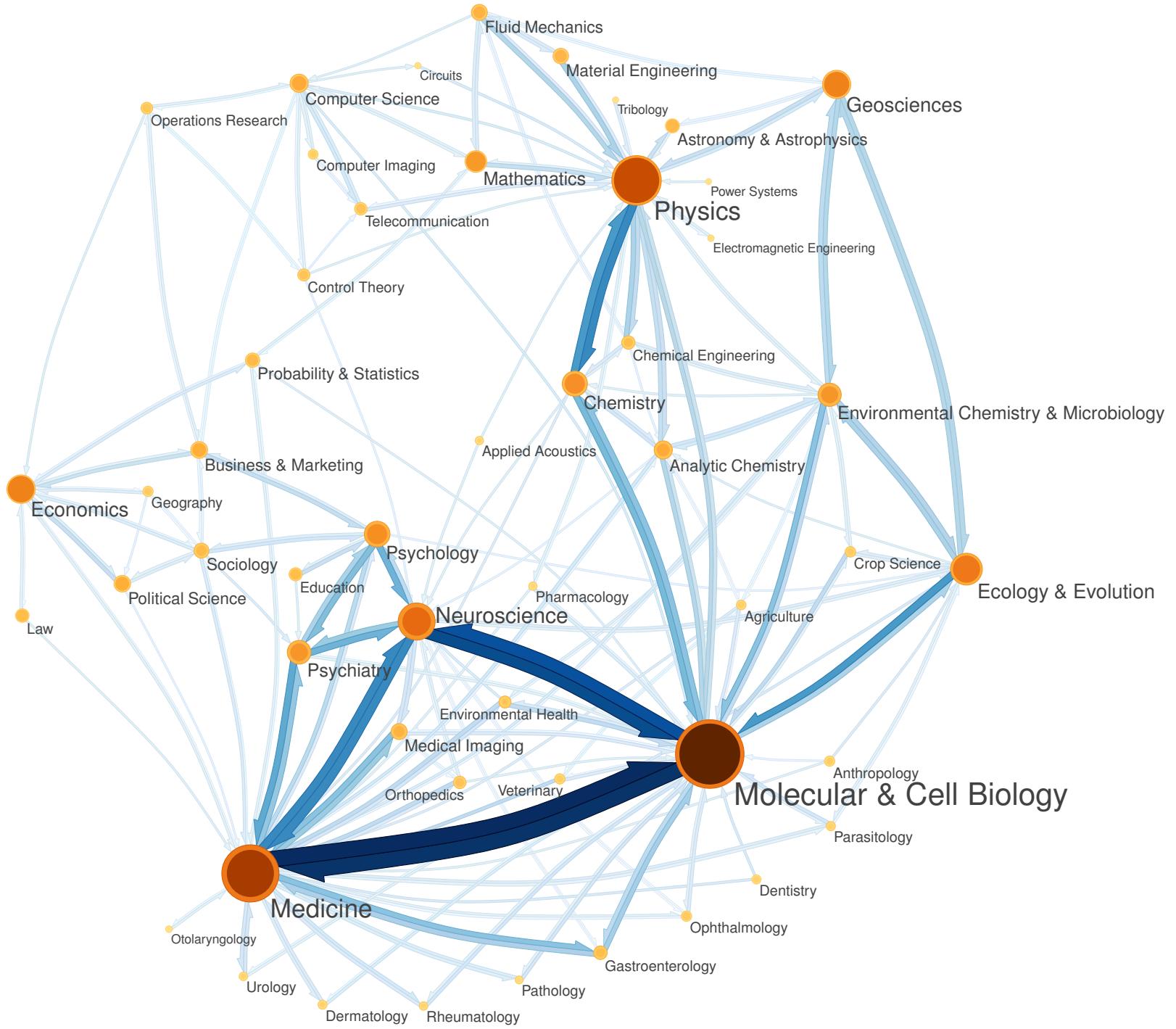
17,849 researchers

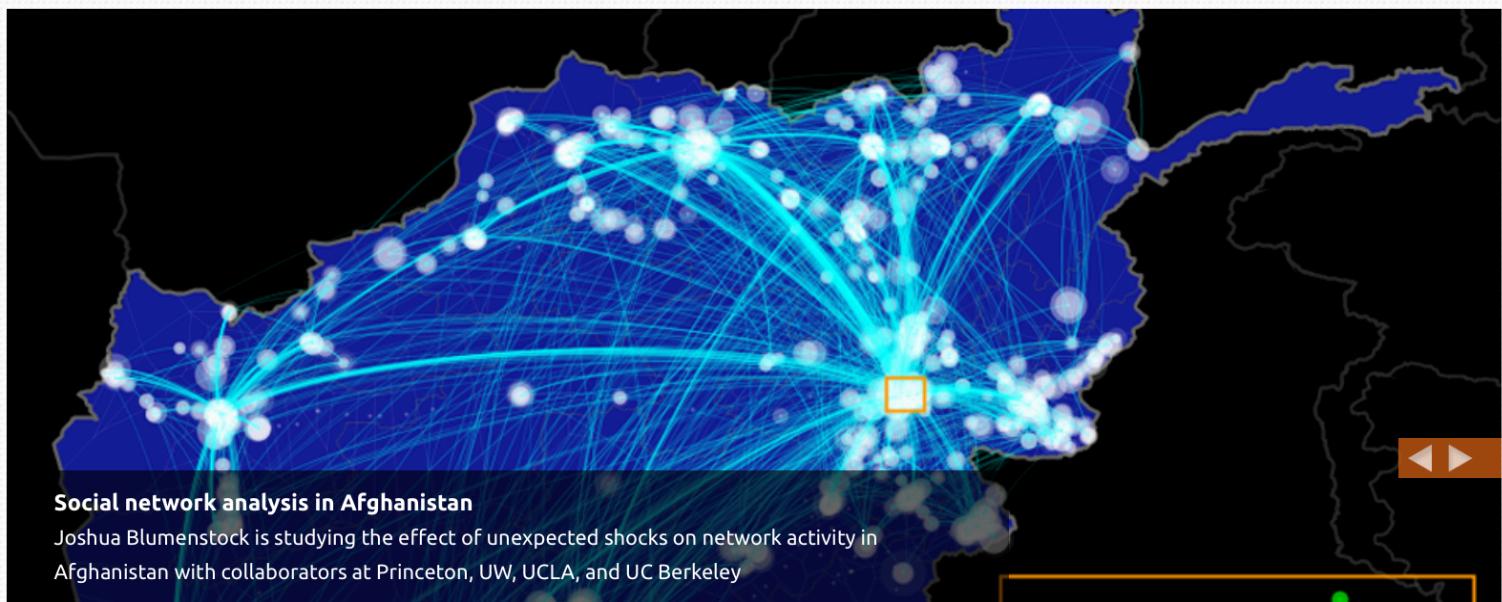
344,917 papers

8,174,533 citations

23.7 citations/paper

2006 - 2015

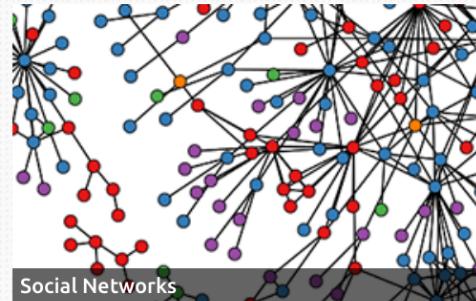




Research Focus Areas



Data for Development



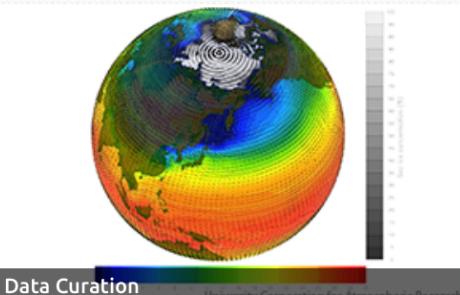
Social Networks



Data Visualization



Computational Social Science

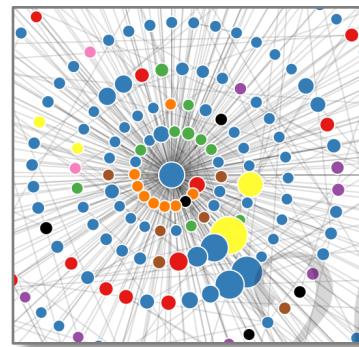


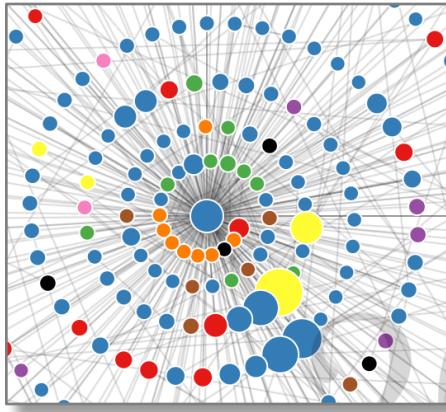
Data Curation



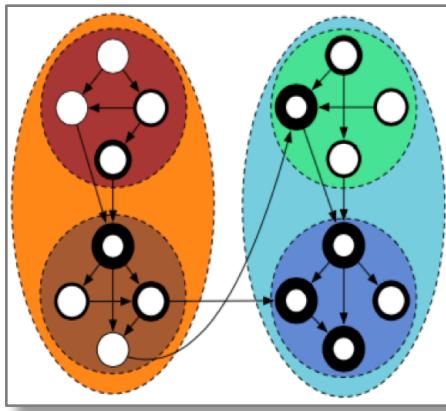
Science of Science

Science of Science

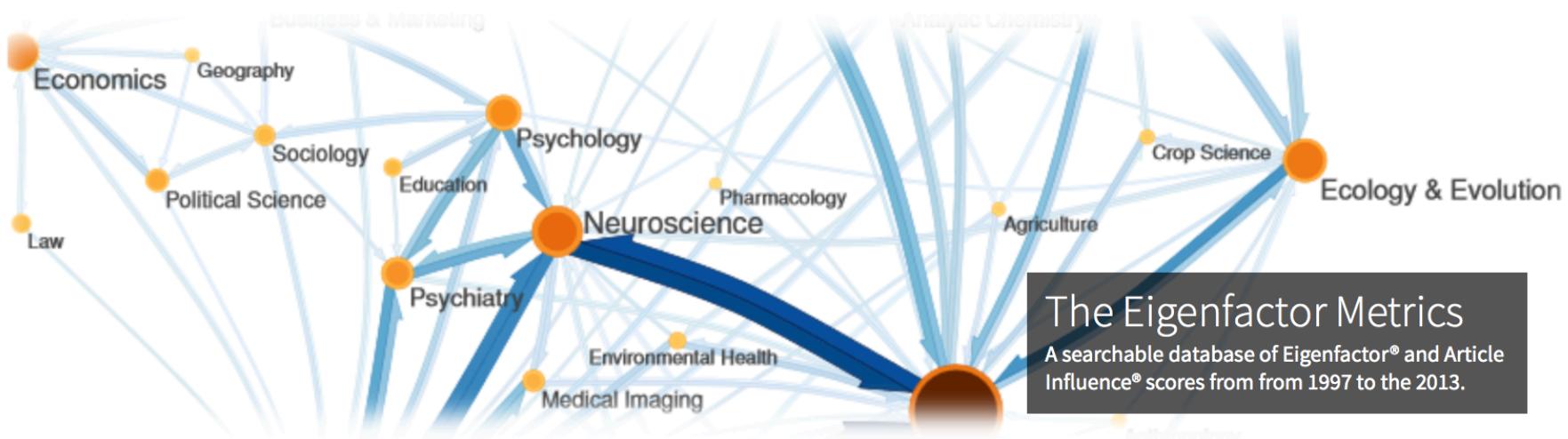




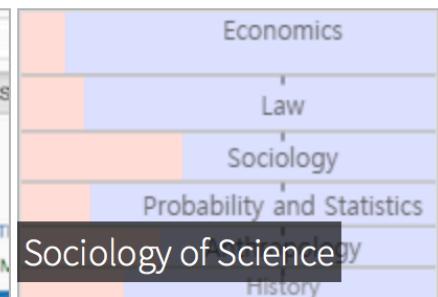
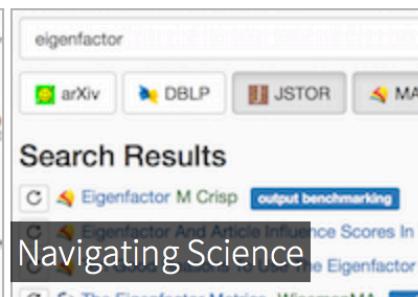
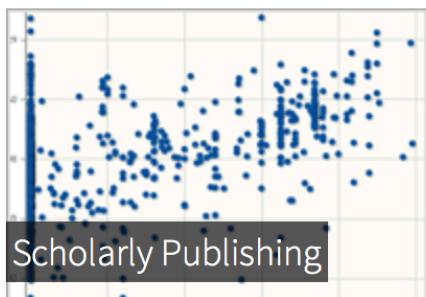
Knowledge Science



Knowledge Engineering



RESEARCH AREAS

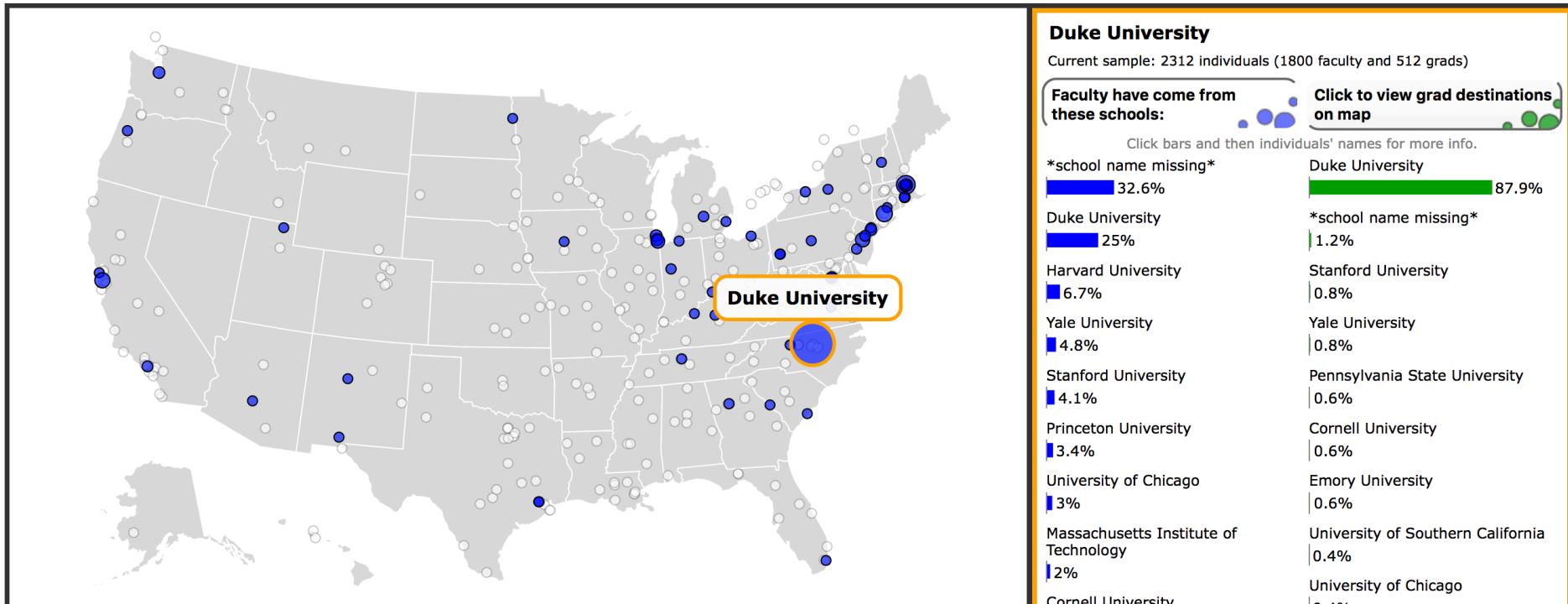


Faculty Migration

U.S. ACADEMIC MIGRATION MAP

Where do people who pursue academic careers in the U.S. go on to land faculty positions after earning their advanced degrees?

Where do faculty come from? Click on a school to explore.



Eigenfactor.org/projects/migration/

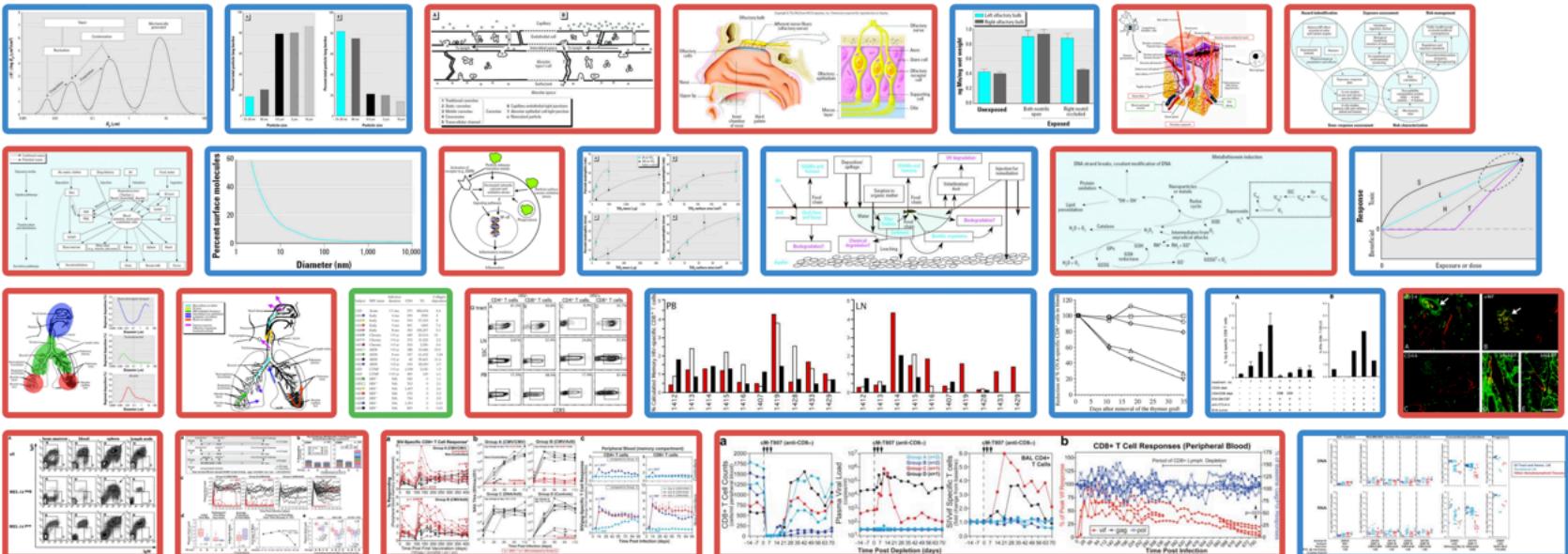
Figure-Centric Search Engine

 viziometrics.org

VizioMetrix About **Search** Crowdsourcing

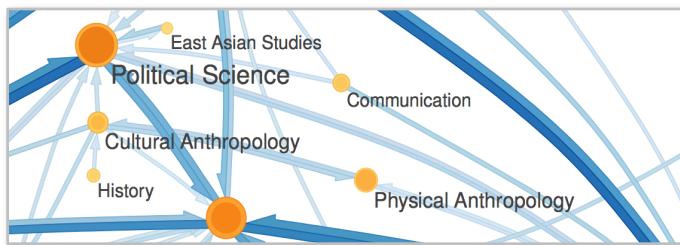
Impact blood lymph

Composite Equation Diagram Photo Plot Table

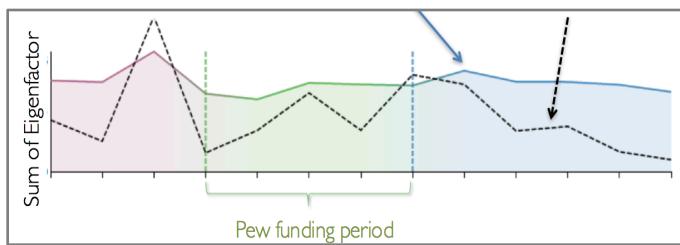


A project of the eScience Institute at the University of Washington

Talk Outline



Visualizing scholarly influence



Measuring funding impact

The H-index impact on science...



Jure Leskovec

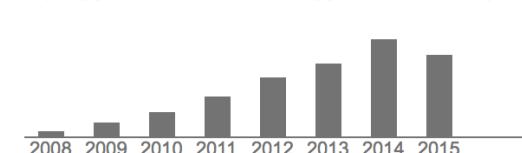
Follow ▾

Professor of Computer Science, Stanford University
Data mining, Social Network Analysis, Information Networks
Verified email at cs.stanford.edu - [Homepage](#)

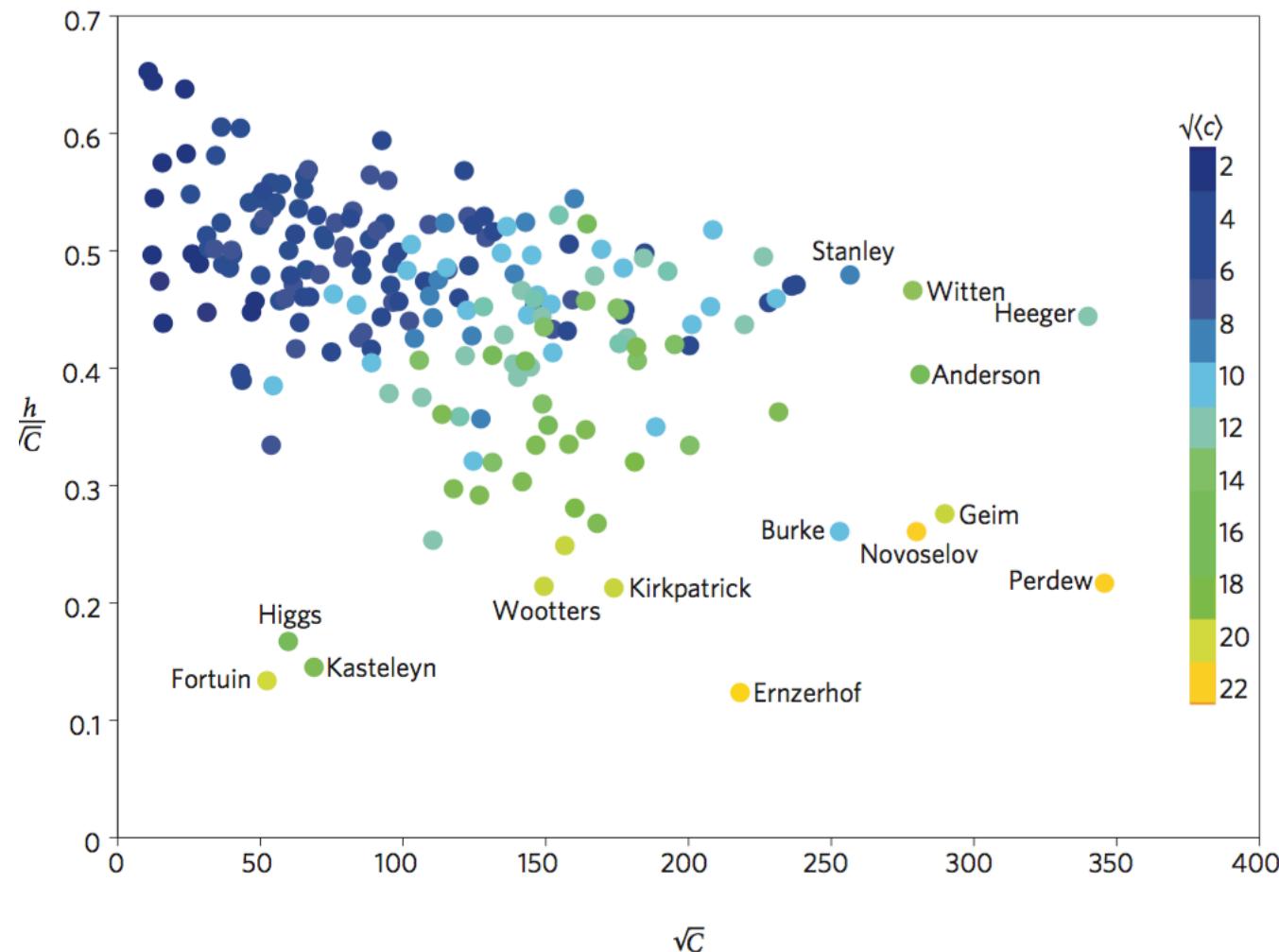
Title	1–20	Cited by	Year
Graphs over time: densification laws, shrinking diameters and possible explanations		1373	2005
J Leskovec, J Kleinberg, C Faloutsos Proceedings of the eleventh ACM SIGKDD international conference on Knowledge ...			
The dynamics of viral marketing		1338	2007
J Leskovec, LA Adamic, BA Huberman ACM Transactions on the Web (TWEB) 1 (1), 5			
Cost-effective outbreak detection in networks		887	2007
J Leskovec, A Krause, C Guestrin, C Faloutsos, J VanBriesen, N Glance Proceedings of the 13th ACM SIGKDD international conference on Knowledge ...			
Meme-tracking and the dynamics of the news cycle		885	2009
J Leskovec, L Backstrom, J Kleinberg Proceedings of the 15th ACM SIGKDD international conference on Knowledge ...			
Graph evolution: Densification and shrinking diameters		853	2007
J Leskovec, J Kleinberg, C Faloutsos ACM Transactions on Knowledge Discovery from Data (TKDD) 1 (1), 2			
Friendship and mobility: user movement in location-based social networks		728	2011
E Cho, SA Myers, J Leskovec Proceedings of the 17th ACM SIGKDD international conference on Knowledge ...			
Community structure in large networks: Natural cluster sizes and the absence of a small-world phenomenon			

Google Scholar

Citation indices	All	Since 2010
Citations	19409	17853
h-index	59	56
i10-index	103	101

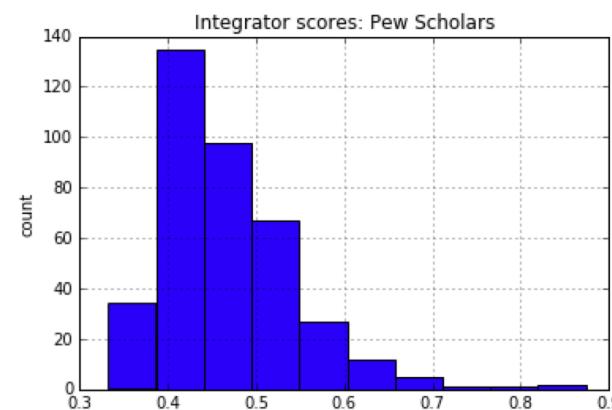
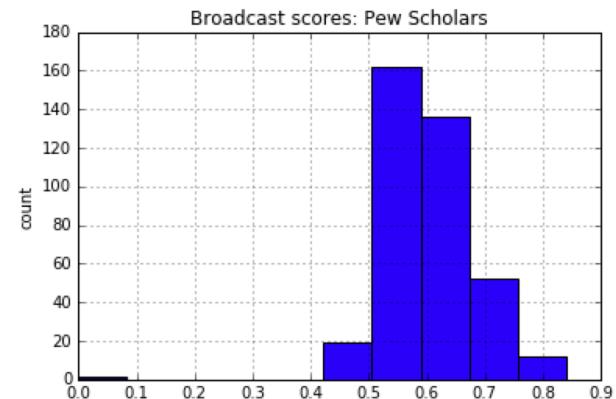
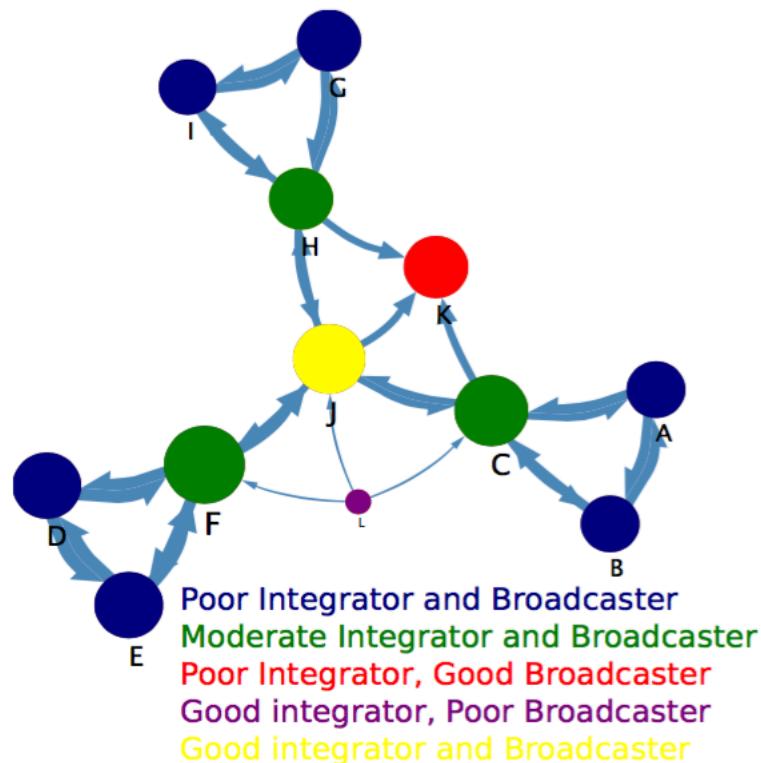


An evisceration of the H-index...



S. N. Dorogovtsev and J. F. F. Mendes (2015) Nature Physics

Measuring Interdisciplinarity



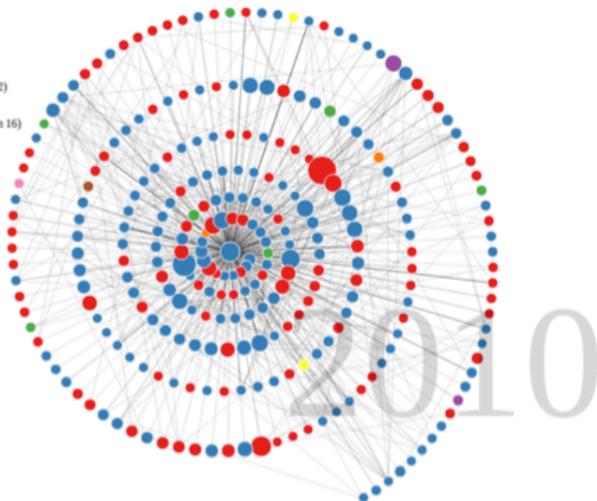
Bergstrom, CT, Foster, J, Portenoy, J, A. Misra, West, JD. (2016). Measuring interdisciplinarity without subject categories. (in prep)

Visualizing Interdisciplinarity



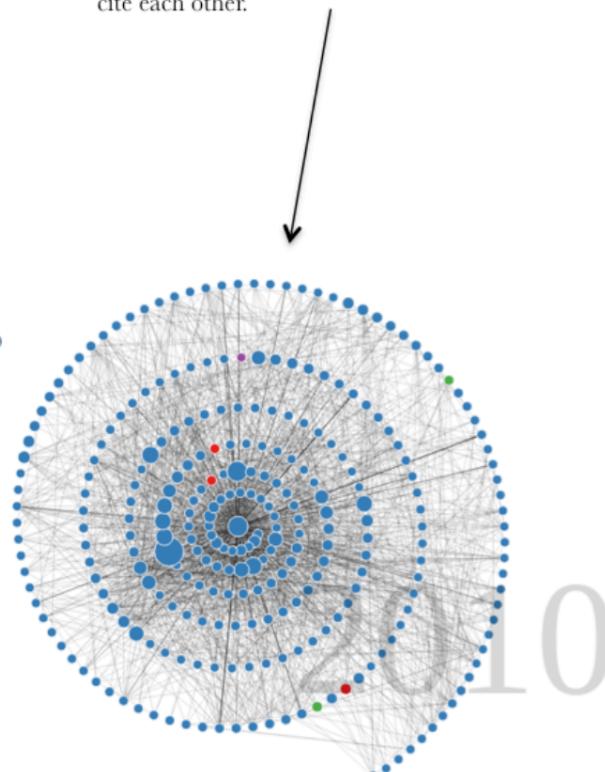
Jason Portenoy

- Papers in category "Medicine" (domain 6)
- Papers in category "Biology" (domain 4)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Engineering" (domain 8)
- Papers in category "Material Science" (domain 12)
- Papers in category "Physics" (domain 19)
- Papers in category "Agriculture Science" (domain 16)
- Papers in category "Social Science" (domain 22)



A more sparse network indicates fewer citations between papers shown in the network. This could be a result of the central scholar having impact across a wider set of academic communities.

- Papers in category "Biology" (domain 4)
- Papers in category "Medicine" (domain 6)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Social Science" (domain 22)



Visualizing Scholarly Influence Over Time

Influence of Pew Scholars

Roberta A. Gottlieb

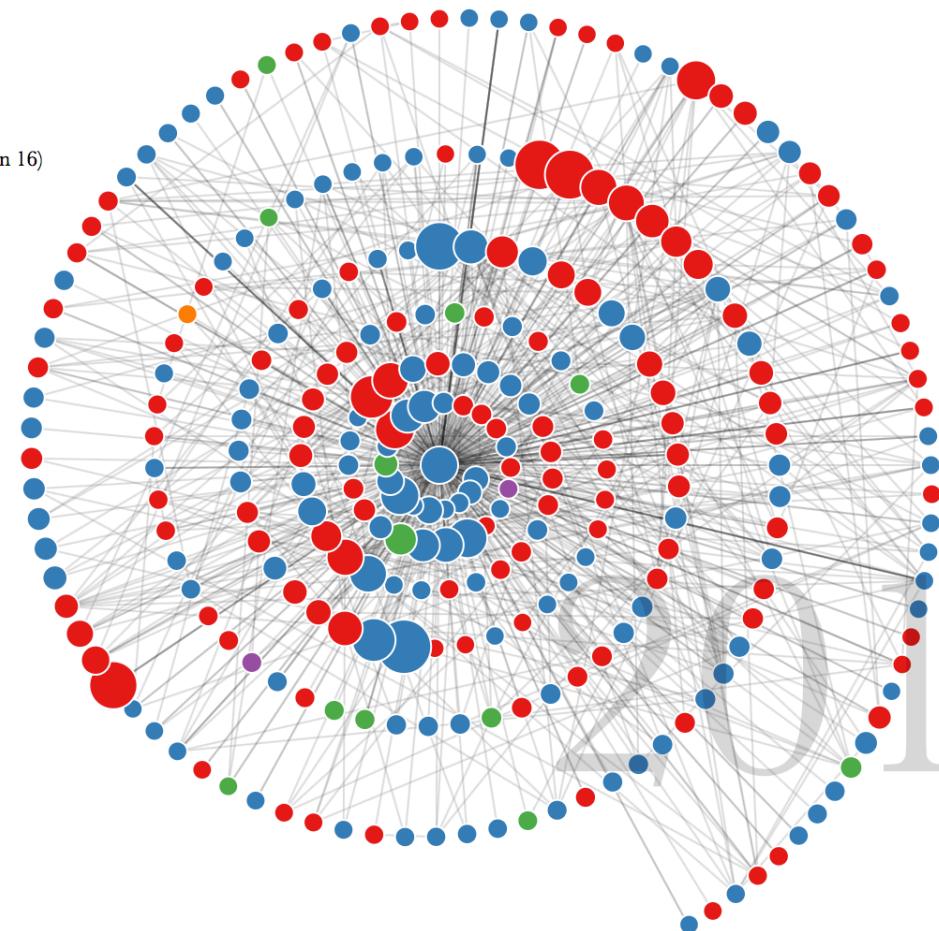
[Learn More](#)

- █ Papers in category "Medicine" (domain 6)
- █ Papers in category "Biology" (domain 4)
- █ Papers in category "Chemistry" (domain 5)
- █ Papers in category "Unknown" (domain 0)
- █ Papers in category "Agriculture Science" (domain 16)

Roberta A.
Gottlieb



Pew Scholar
1997

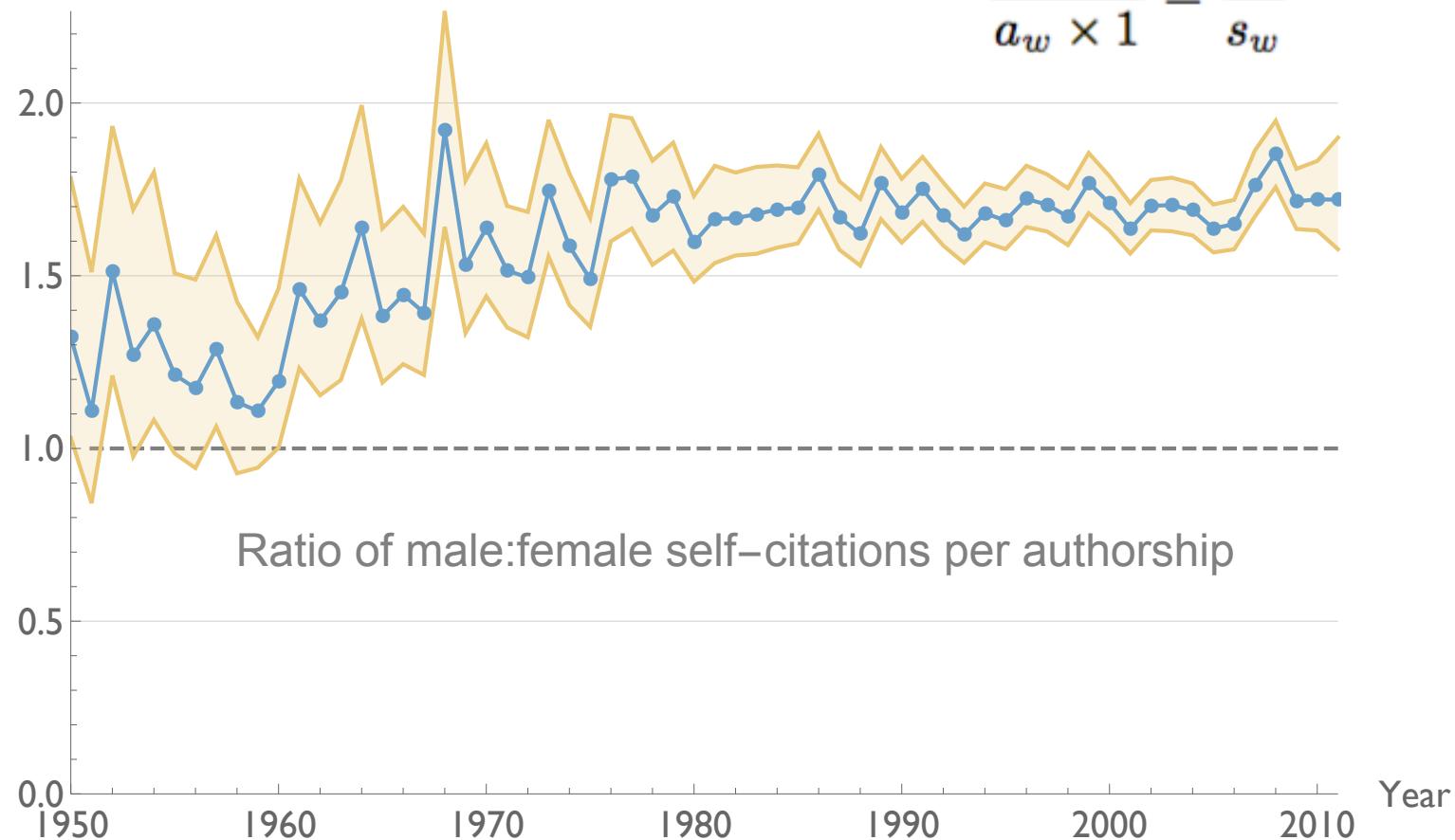


1997

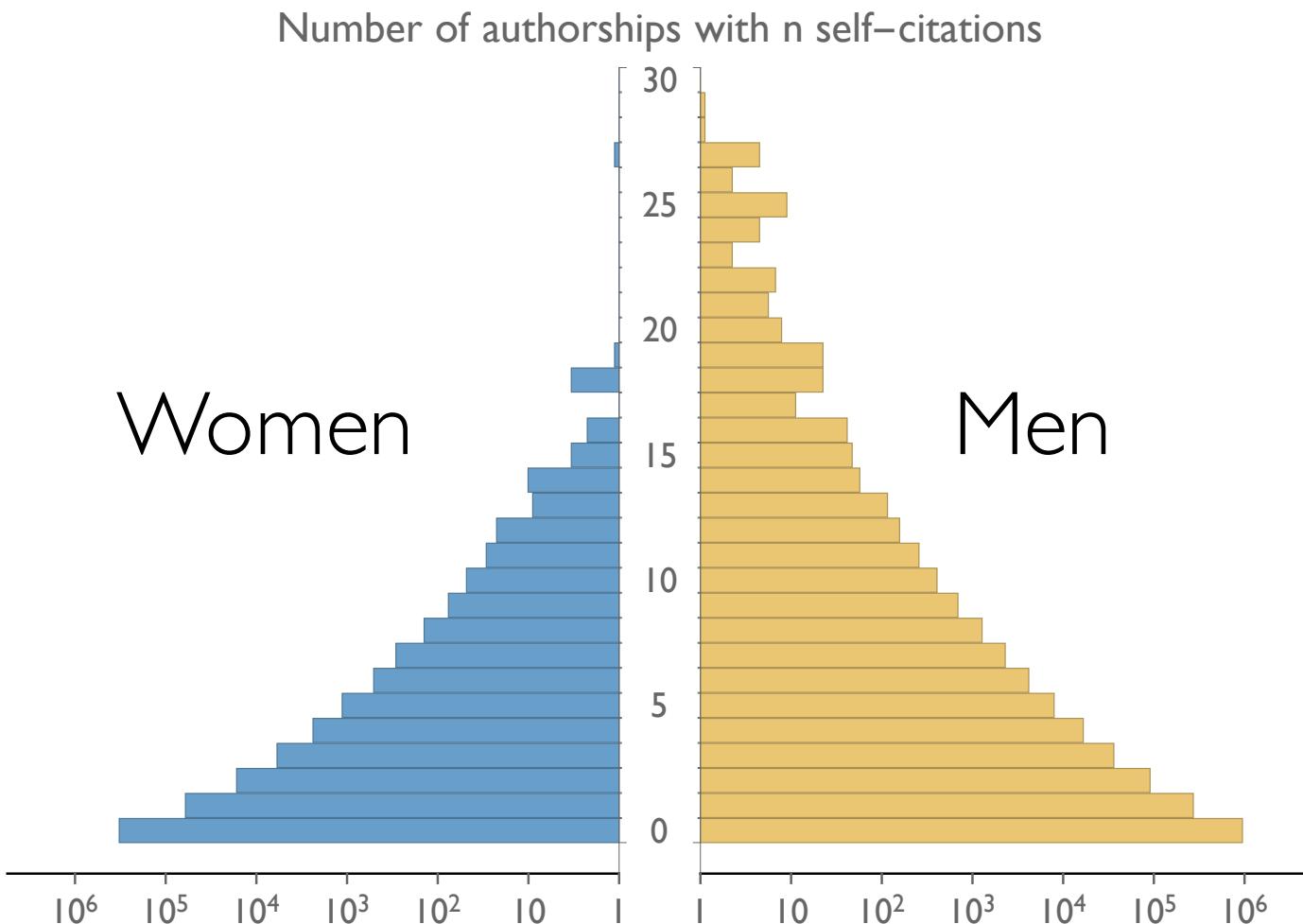
Self-citation over time

Ratio M:F

$$\frac{a_m \times k}{a_w \times 1} = \frac{s_m}{s_w}$$

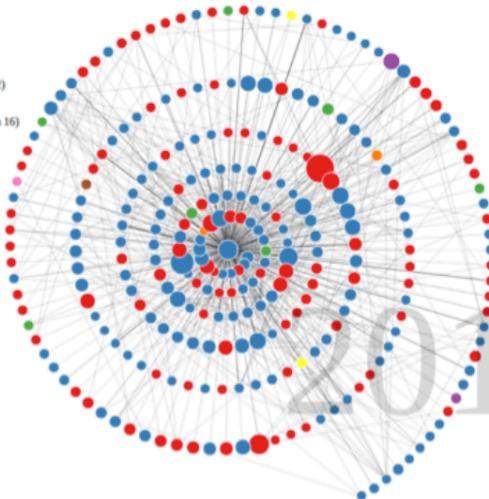


Gender and Self Citation



Comparing Authors

- Papers in category "Medicine" (domain 6)
- Papers in category "Biology" (domain 4)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Engineering" (domain 8)
- Papers in category "Material Science" (domain 12)
- Papers in category "Physics" (domain 19)
- Papers in category "Agriculture Science" (domain 16)
- Papers in category "Social Science" (domain 22)

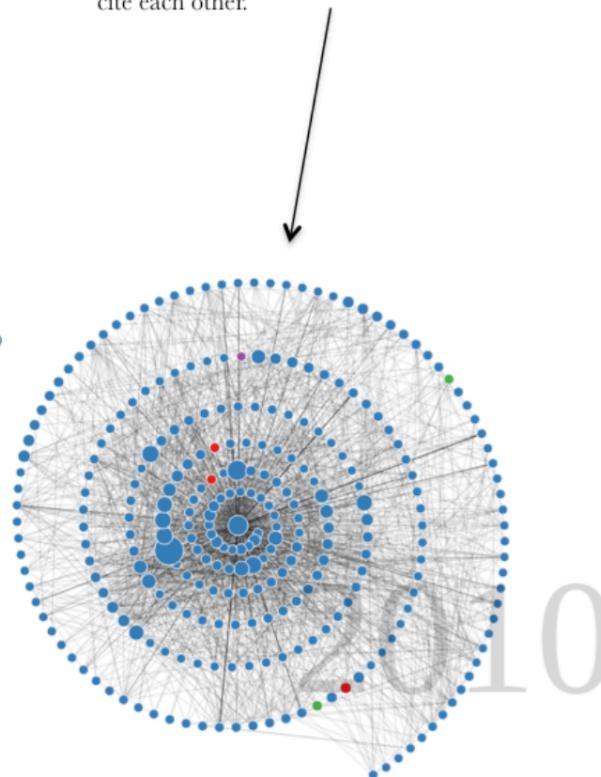


A denser network means that the papers that cite the central author also tend to cite each other.



A more sparse network indicates fewer citations between papers shown in the network. This could be a result of the central scholar having impact across a wider set of academic communities.

- Papers in category "Biology" (domain 4)
- Papers in category "Medicine" (domain 6)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Social Science" (domain 22)



Visualizing Scholarly Influence Over Time

Influence of Pew Scholars

Mark W. Grinstaff

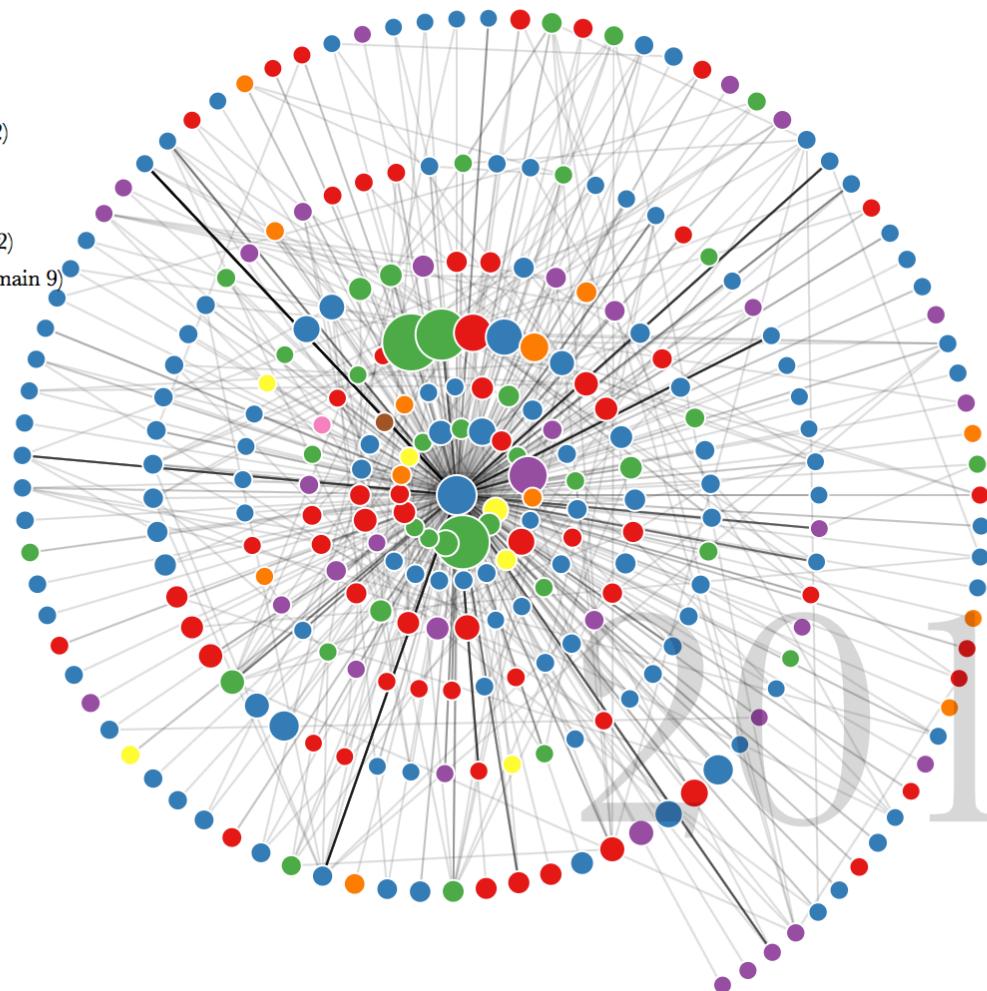
[Learn More](#)

- Papers in category "Chemistry" (domain 5)
- Papers in category "Medicine" (domain 6)
- Papers in category "Biology" (domain 4)
- Papers in category "Material Science" (domain 12)
- Papers in category "Engineering" (domain 8)
- Papers in category "Physics" (domain 19)
- Papers in category "Computer Science" (domain 2)
- Papers in category "Environmental Sciences" (domain 9)

Mark W.
Grinstaff



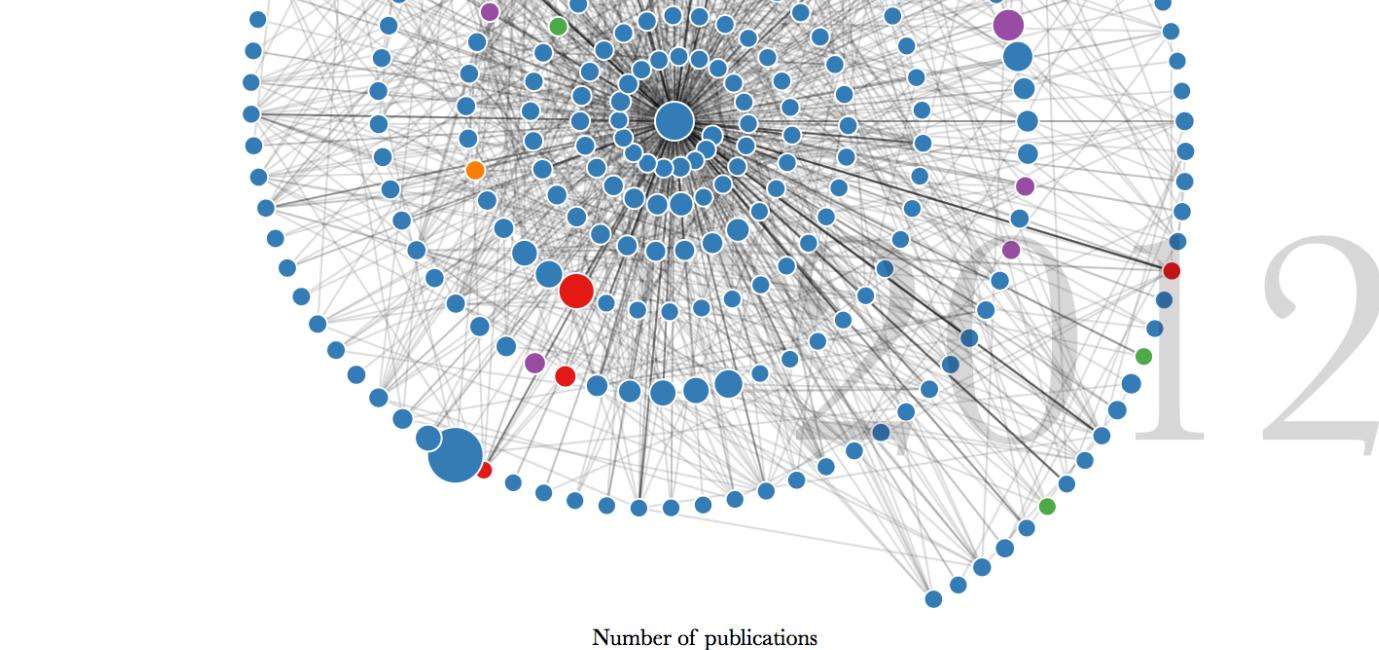
Pew Scholar
1999



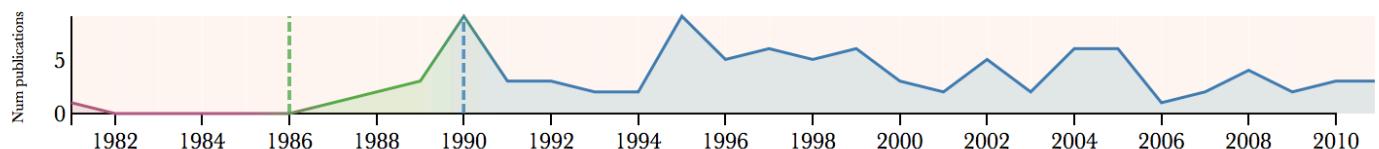
Philip A.
Hieter



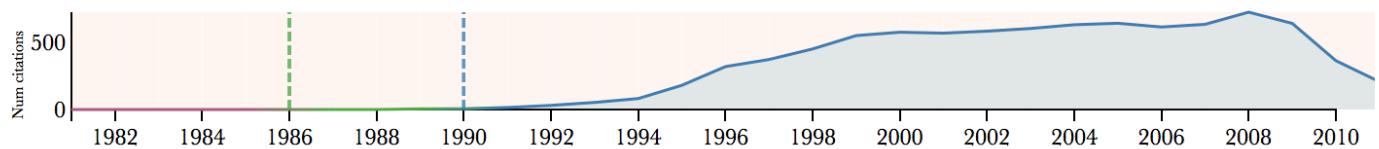
Pew Scholar
1986



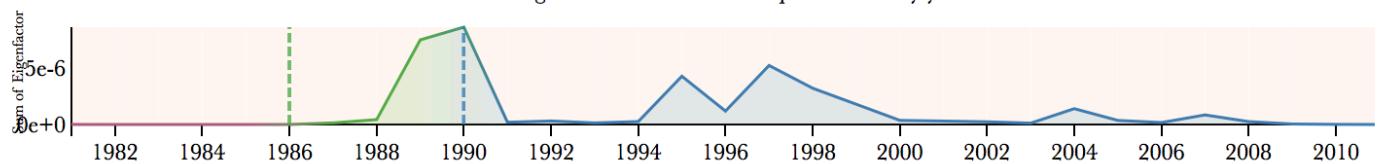
Number of publications



Number of citations received



Sum of eigenfactor for this author's publications by year



Funding

Citation Data



49 million scholarly publications

260 million citations

354 Pew Scholars

22,000 publications

62 publications/scholar

Pew EF 3 times the average EF

field classification



Scholars Program
in the Biomedical Sciences



Science

~ 37 citations/paper

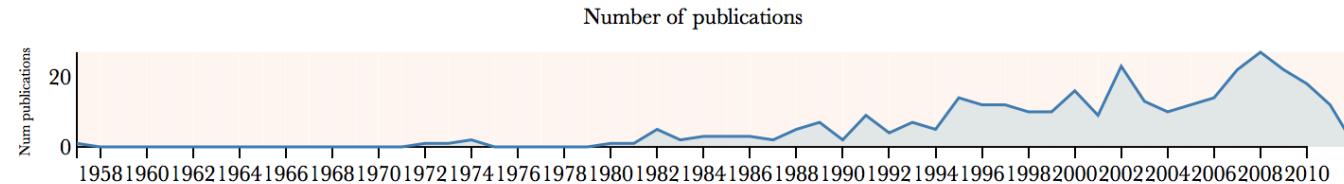
median citations = 11

~ 5 citations/paper

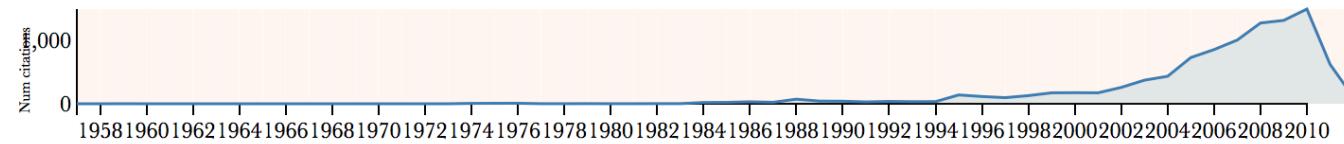
median citations = 0

12

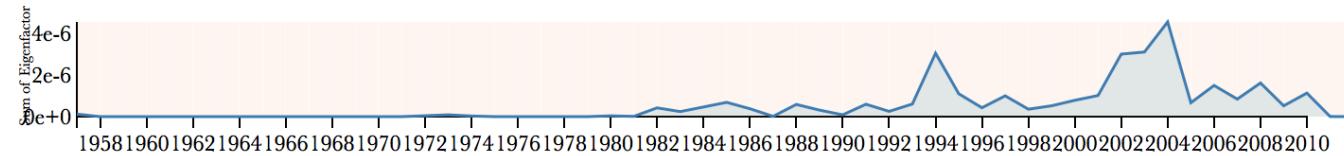
Institution as Center



Number of citations received



Sum of eigenfactor for this author's publications by year



scholar.eigenfactor.org

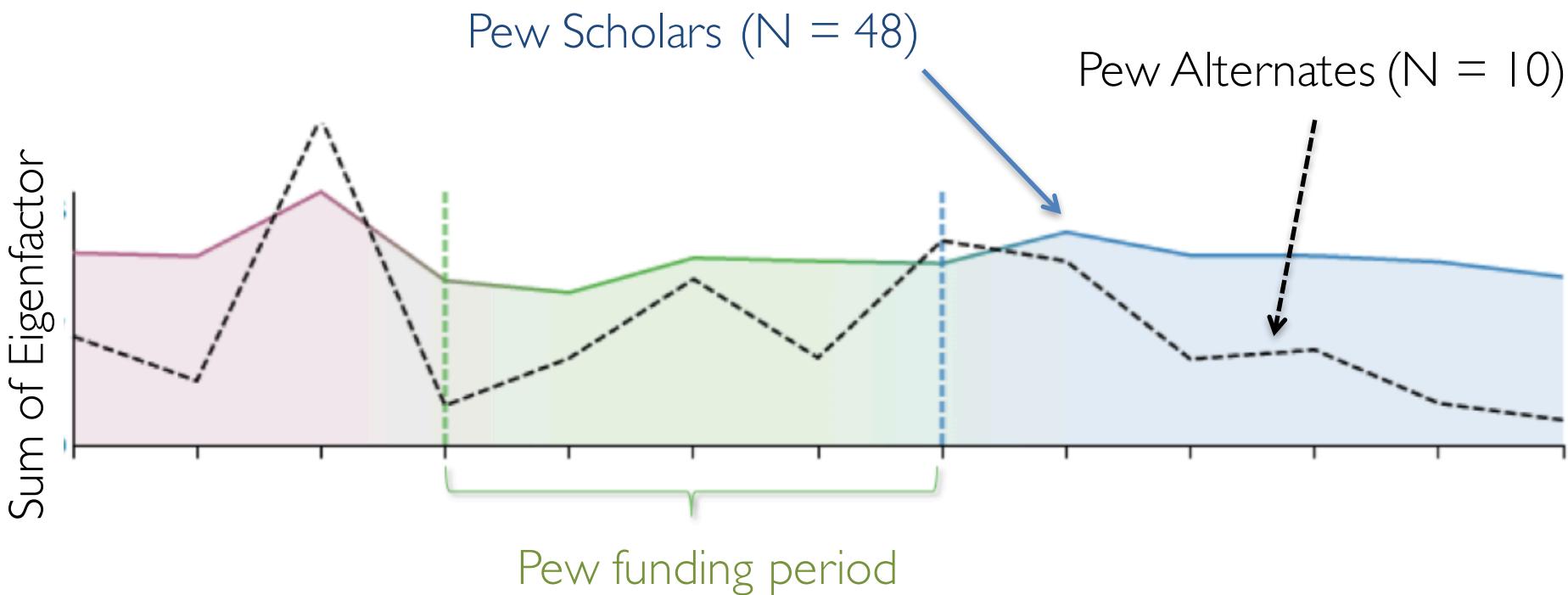
username: PewScholar

password: 1N!kdG

Jevin West, jevinw@uw.edu

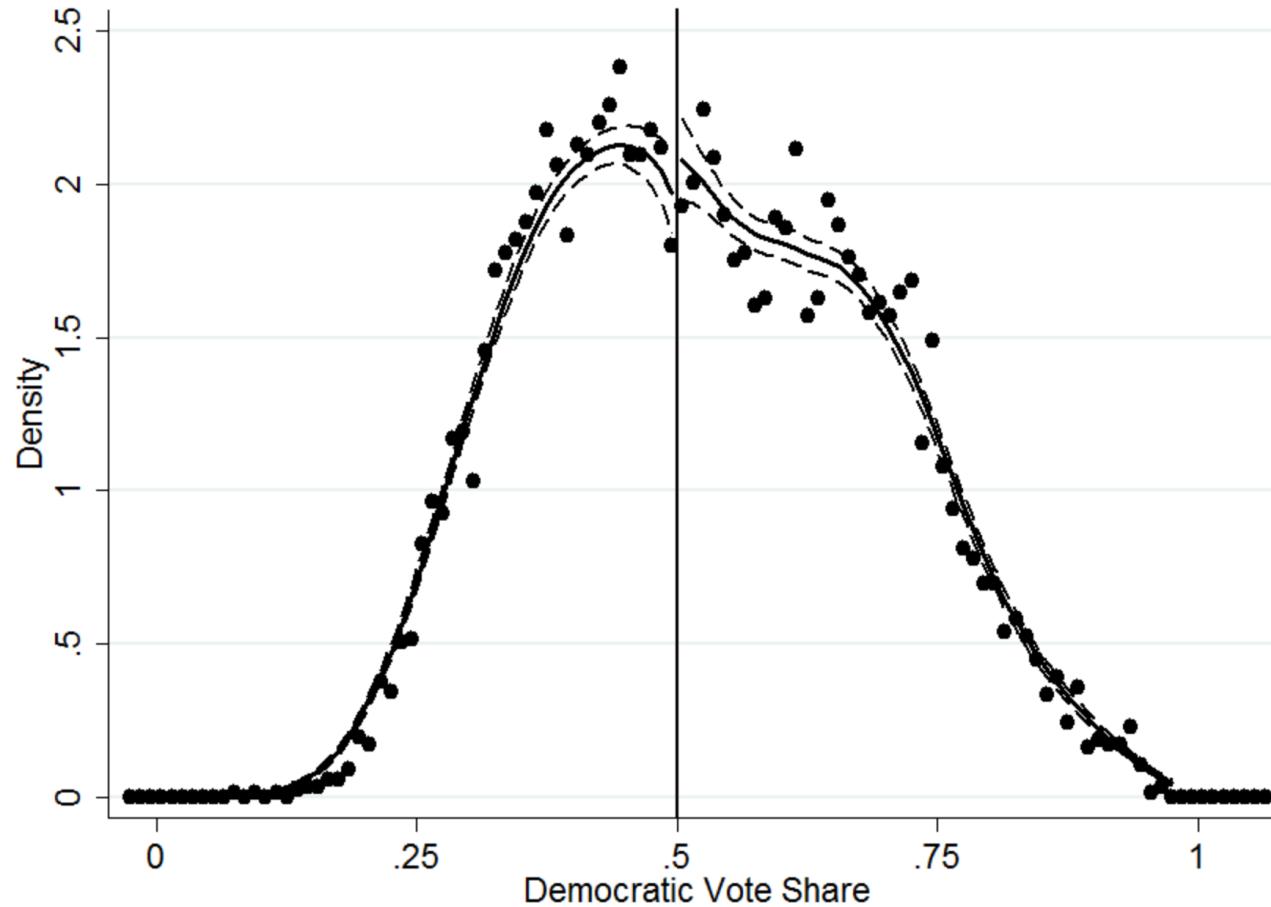
The Pew Impact...

Comparing Alternates



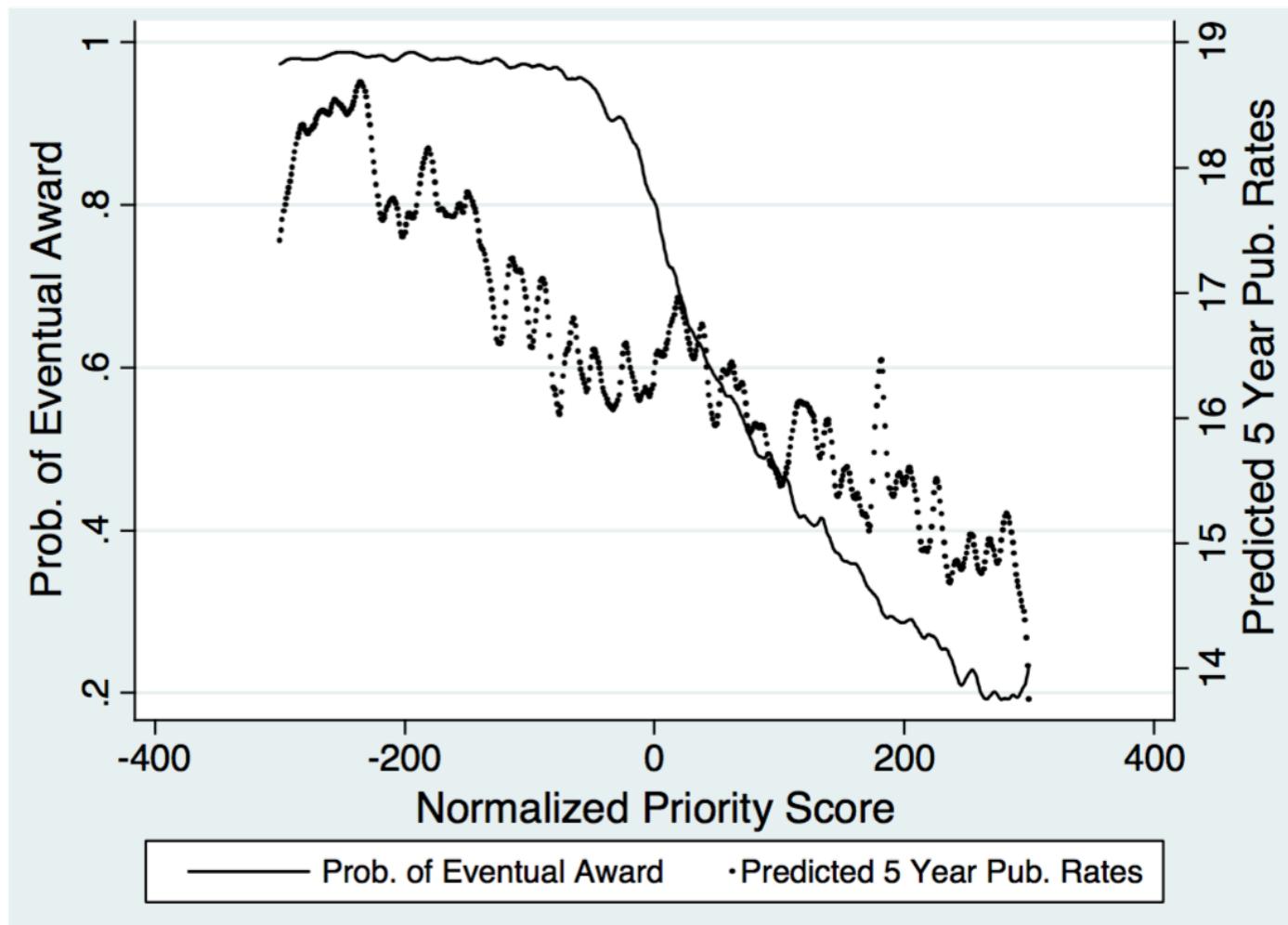
* Includes scholars and alternates from cohort years: 1997, 1999, 2000, 2001, 2002

Regression discontinuity design

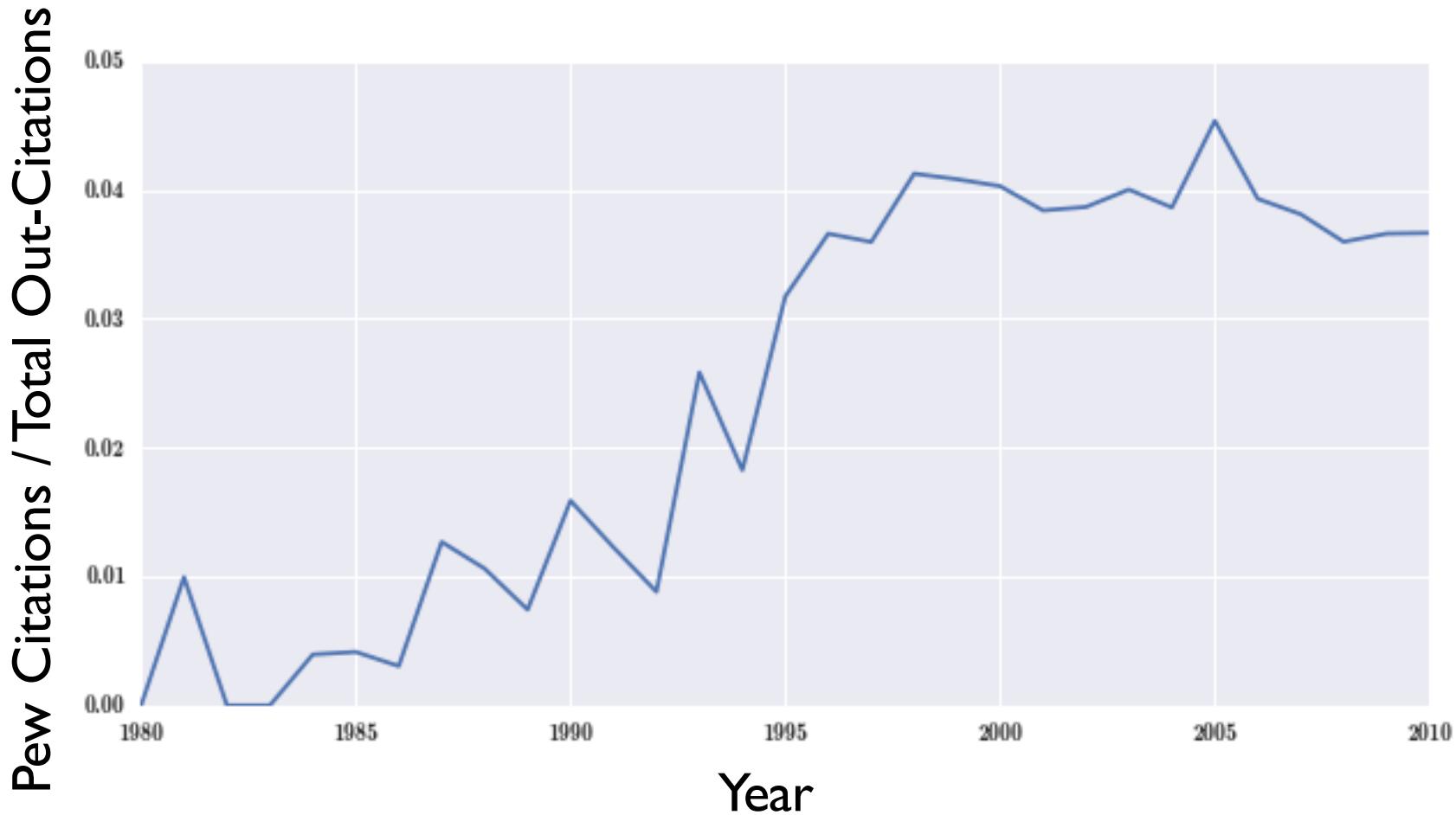


McCrary (2008)^[10] density test on data from Lee, Moretti, and Butler (2004).

The impact of research grant funding on scientific productivity



Community Effect



Capturing the Value of Biomedical Research

Stefano Bertuzzi^{2,3} and Zeina Jamaleddine^{1,*}

¹Sidra Medical and Research Center, Out-Patient Clinic, PO Box 26999, Al Luqta Street, Doha, Qatar

²American Society for Cell Biology, 8120 Woodmont Avenue, Suite 750 Bethesda, MD 20814-2762, USA

³Present address: American Society for Microbiology, 1752 North Street, N.W. Washington, DC 20036-2904, USA

*Correspondence: zjamaleddine@sidra.org

<http://dx.doi.org/10.1016/j.cell.2016.03.004>

Assessing the real-world impact of biomedical research is notoriously difficult. Here, we present the framework for building a prospective science-centered information system from scratch that has been afforded by the Sidra Medical and Research Center in Qatar. This experiment is part of the global conversation on maximizing returns on research investment.

The Complex Case of Assessing the Value of Research Investments

Assessing the impact of research funding on scientific, economic, and health outcomes is a complex endeavor, and quantifying it is made even more difficult by the unpredictable nature of basic science. The question of what society receives back in improved health, cost effectiveness, or cost savings from investments in research seems simple at first glance, but it often leads to an analytical quagmire. It is difficult to measure relevant parameters in a holistic manner that provides meaningful answers beyond mere anecdotes. Too often, analysts fall back on appealing summary statistics, some of which, unfortunately, are based on flawed metrics for assessing scientific productivity (Colantonio, 1999; Leno, 2000;

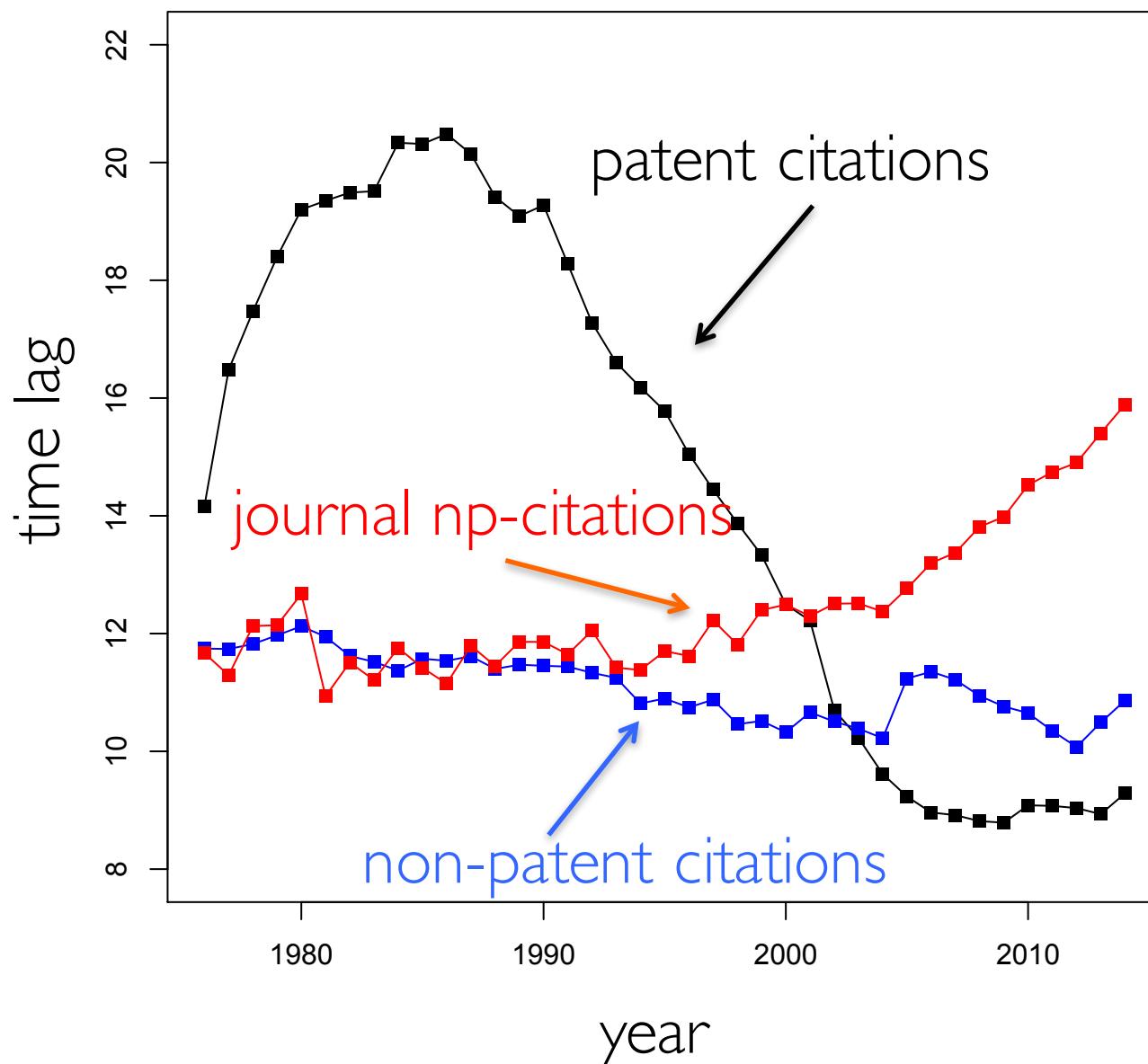
the systems currently in place to track science funding were developed primarily for administrative purposes. Relevant data needed to understand the effects of science funding often reside in different and disparate systems that are not interoperable, even within the same institution. Most importantly, these systems were not designed to capture scientific content and outcomes. This is evident in the fact that nearly all of them use grants as the fundamental unit of analysis. This is the wrong predicate for understanding innovation since the driver is the scientist, while the grant is simply the vehicle. In blunt terms, you can have a big research grant that goes nowhere; you can have an innovator on a tiny grant who upends an entire field. Grant-based analysis cannot tell them apart.

Better tools that are specifically designed to assess the impact of research must be in place (Macilwain, 2010). This was our starting point. We believe this could be achieved by constructing a tracking system for a nation or region's universities, hospitals, funding agencies, regulatory agencies, intellectual property, advocacy, industry, public policies, and strategic plans. Moreover, we could keep track of deliverables in programs that support talented individuals and foster their free creativity through appropriate systems that allow them to easily network and access relevant and useful opportunities.

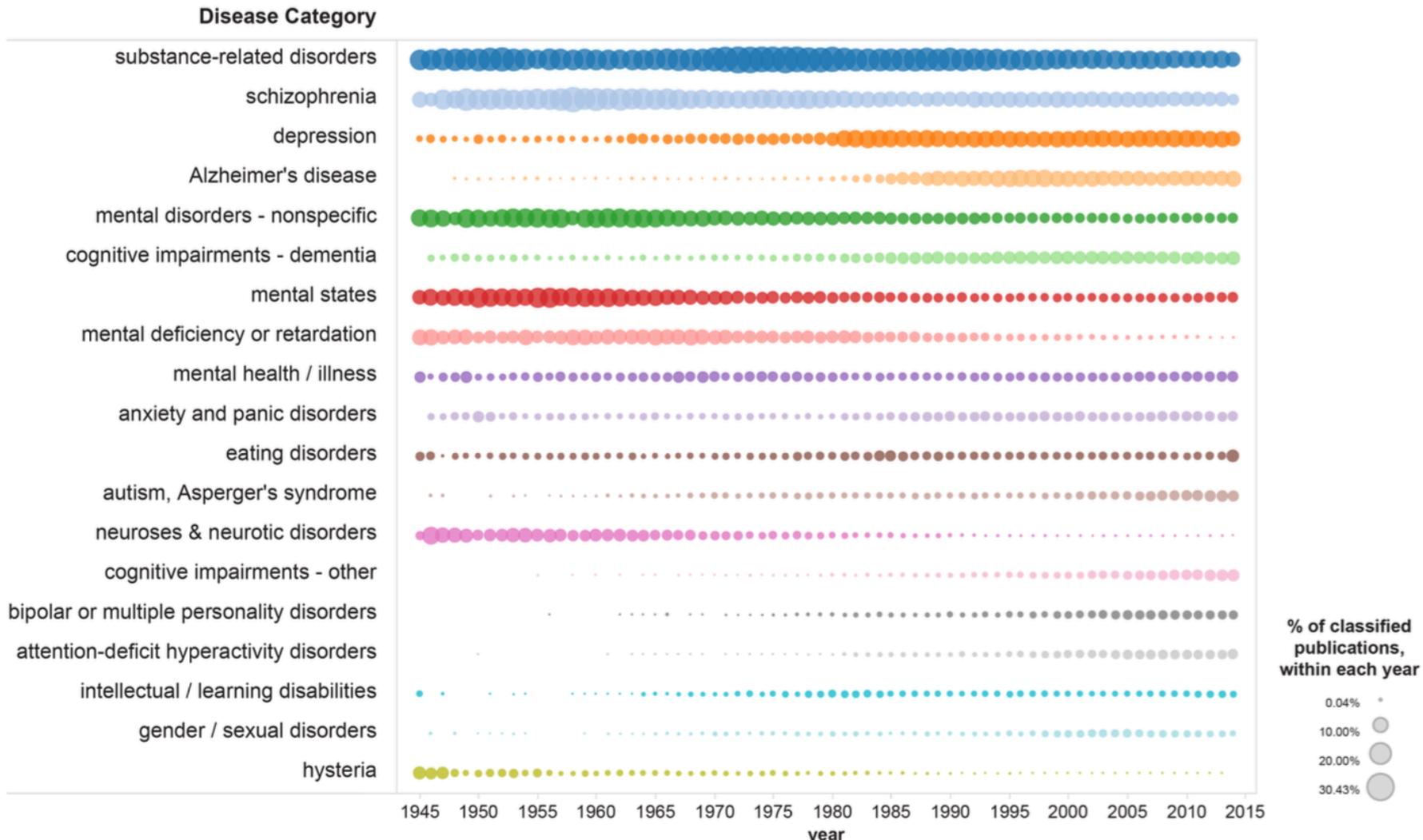
Why Qatar?

Qatar, a small Arabian Gulf country undergoing rapid technological and economic transformation, provides a rare opportunity

Translational Lags

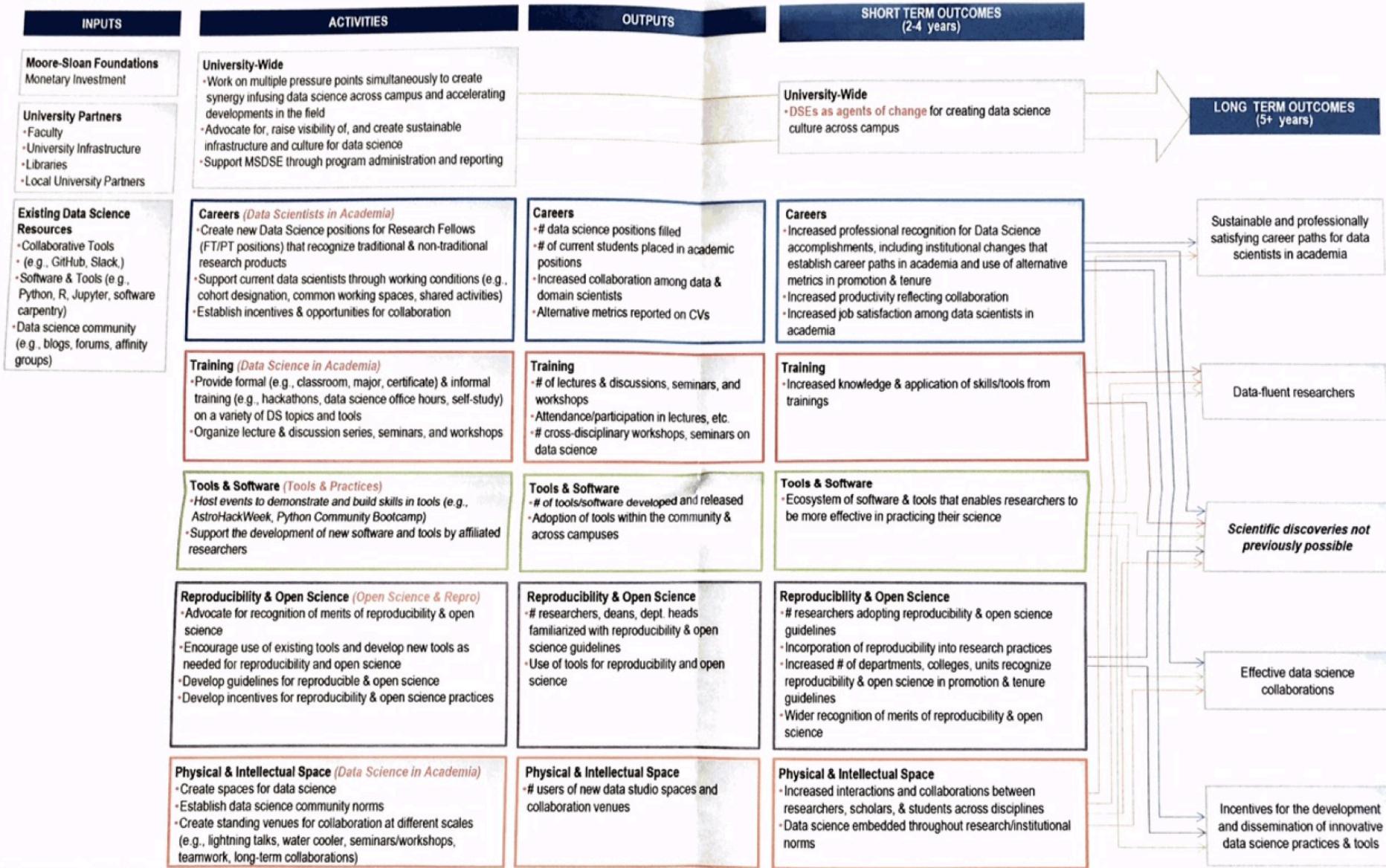


PubMed Publications by Disease Category



Qualitative Methods

Exhibit 1: MSDSE Logic Model



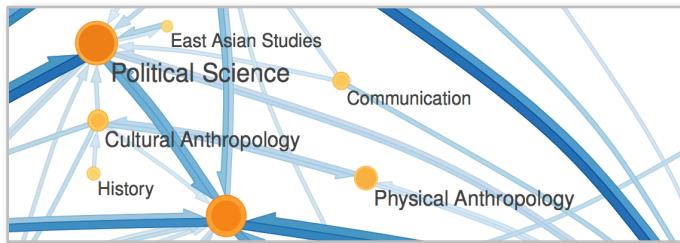
Qualitative Methods



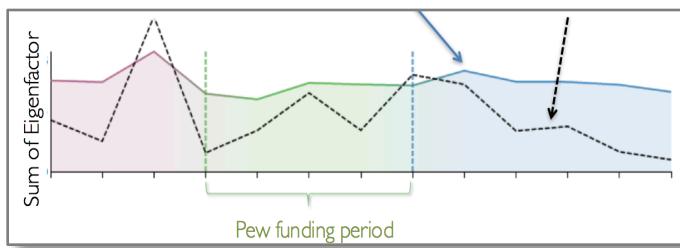
Jody Roberts



Talk Outline

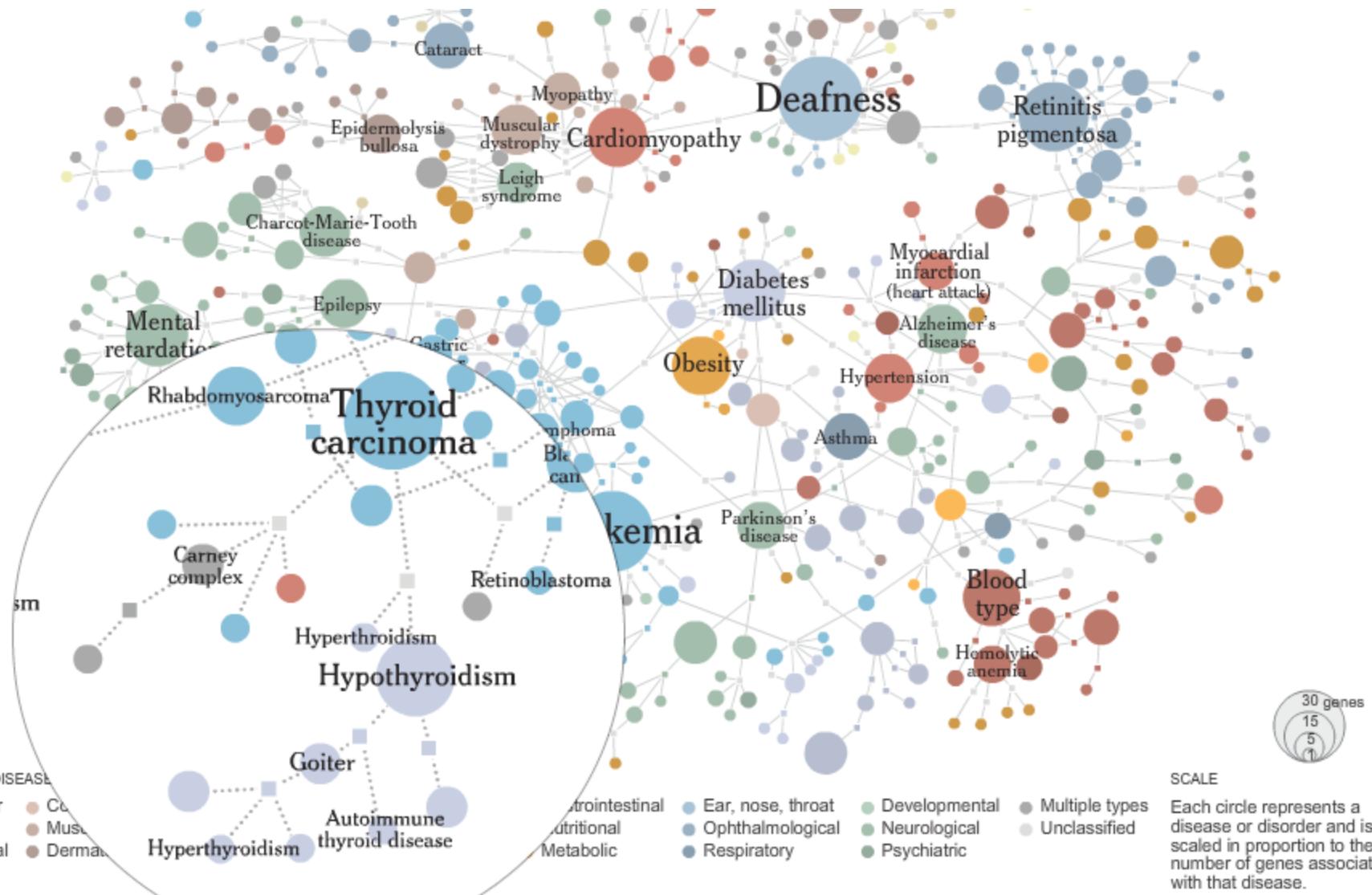


Visualizing scholarly influence

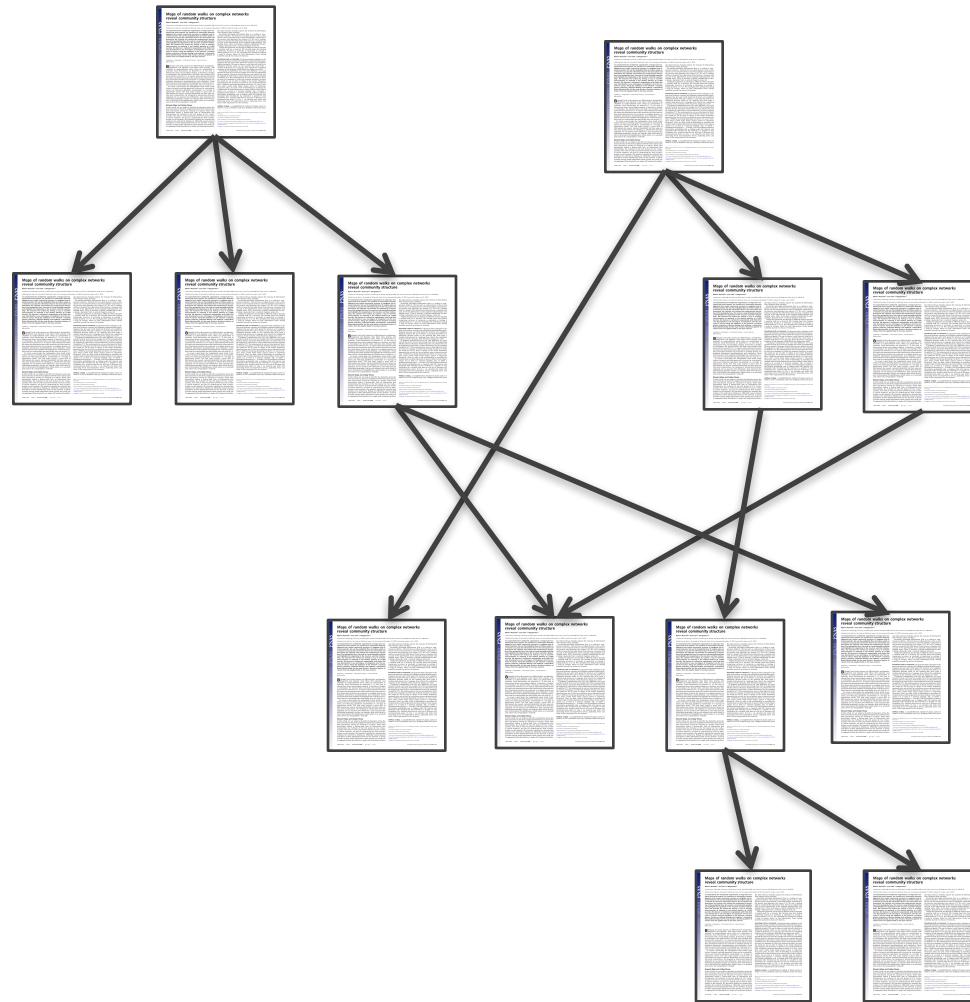


Measuring funding impact

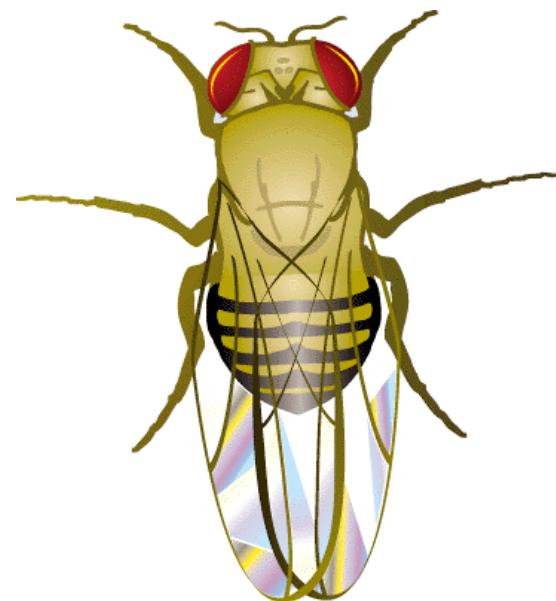
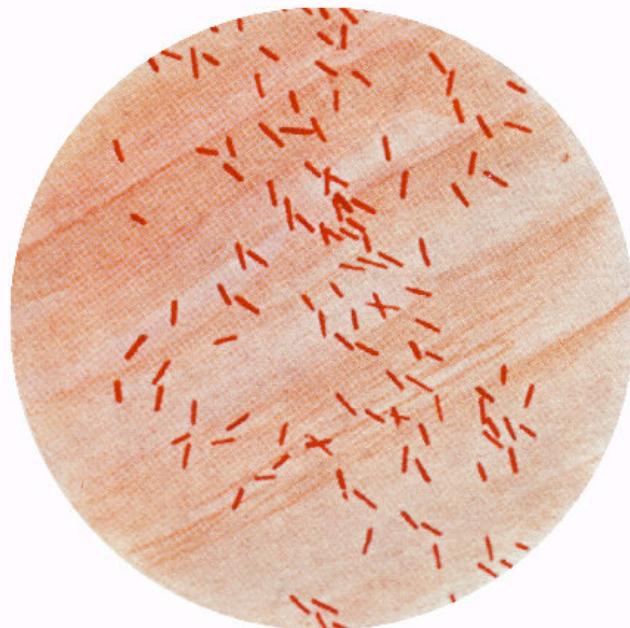
Disease association network

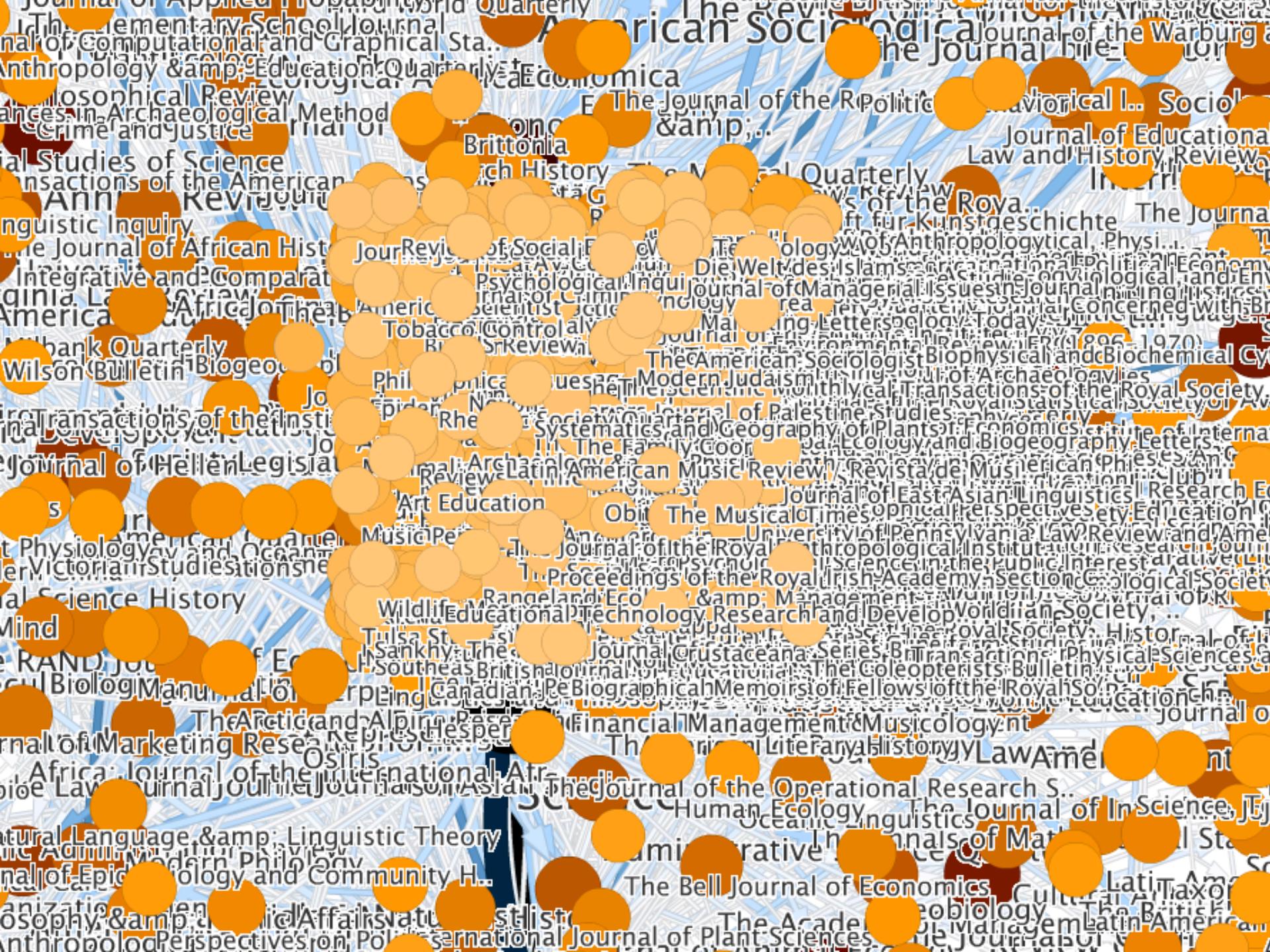


Citations form a vast network



de Solla Price, Science (1965)





Eigenfactor algorithm

$$P = \alpha H + (1 - \alpha) a.e^T$$

Matrix representing the random walk over citations

Probability of not teleporting

Cross-citation Matrix dictating the structure of the citation network

Probability of teleporting to completely new journal weighted by the number of articles in that journal

$$EF = 100 \frac{H\pi}{\sum_i [H\pi]_i}$$

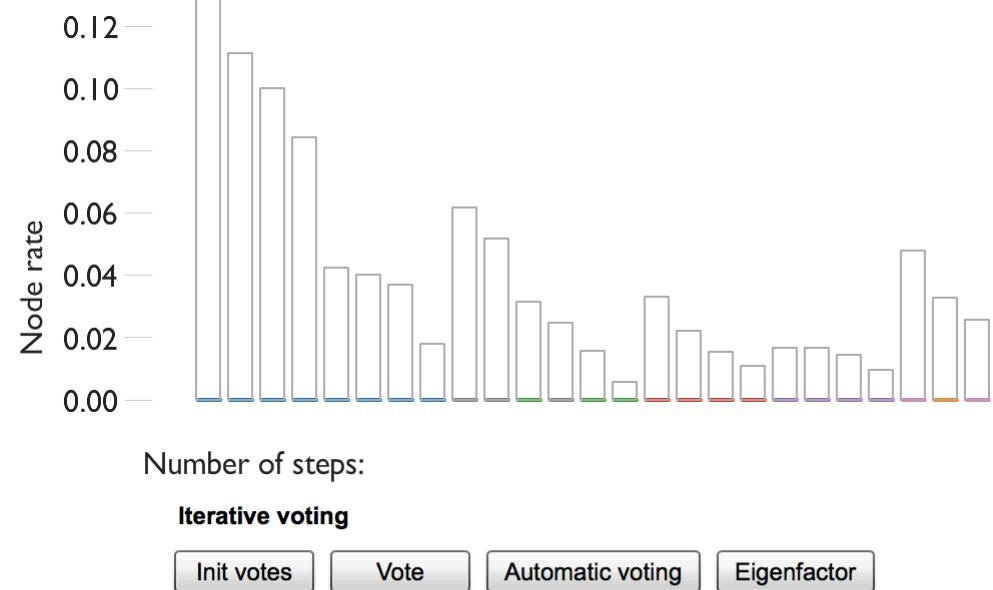
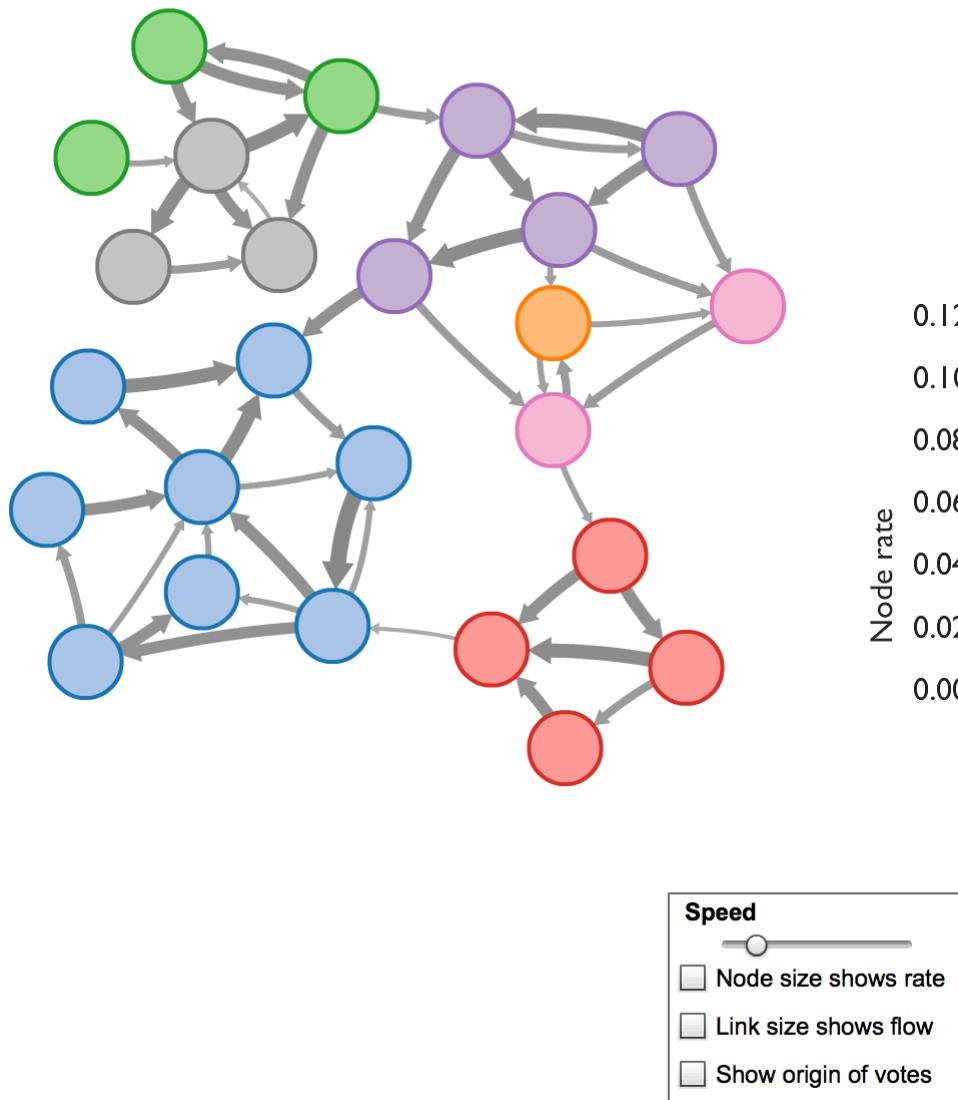
Leading eigenvector of the random walk matrix P .

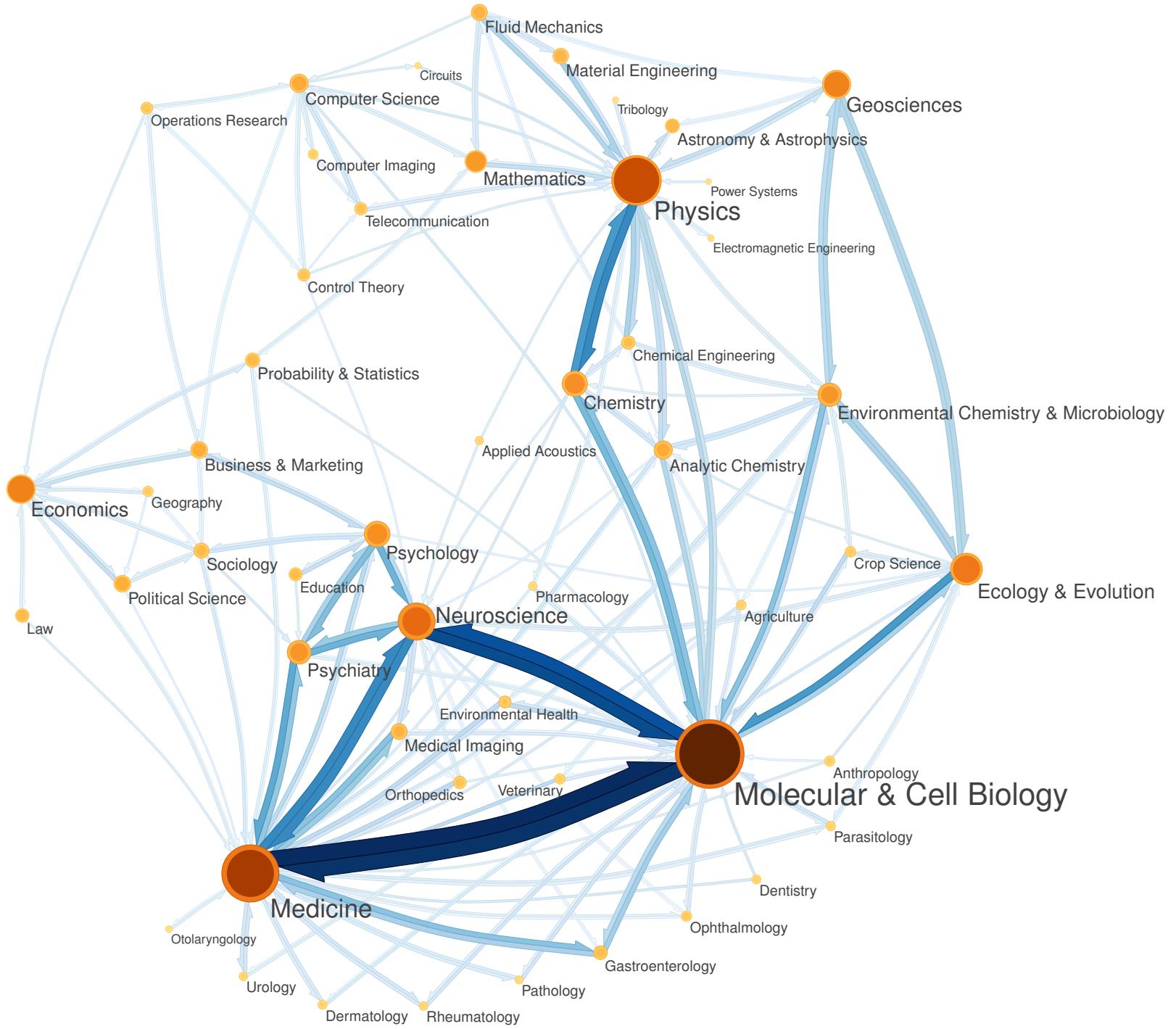
Normalization

The map equation

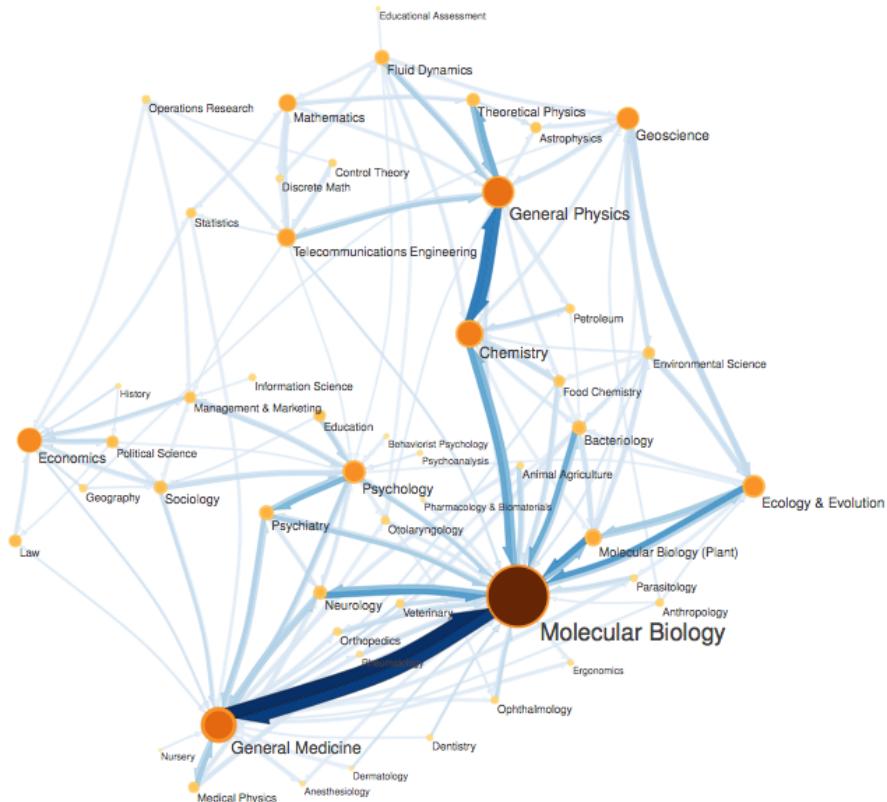
$$L(M) = q_{\curvearrowright} H(Q) + \sum_{i=1}^m p_{\circlearrowleft}^i H(\mathcal{P}^i)$$

Community Detection

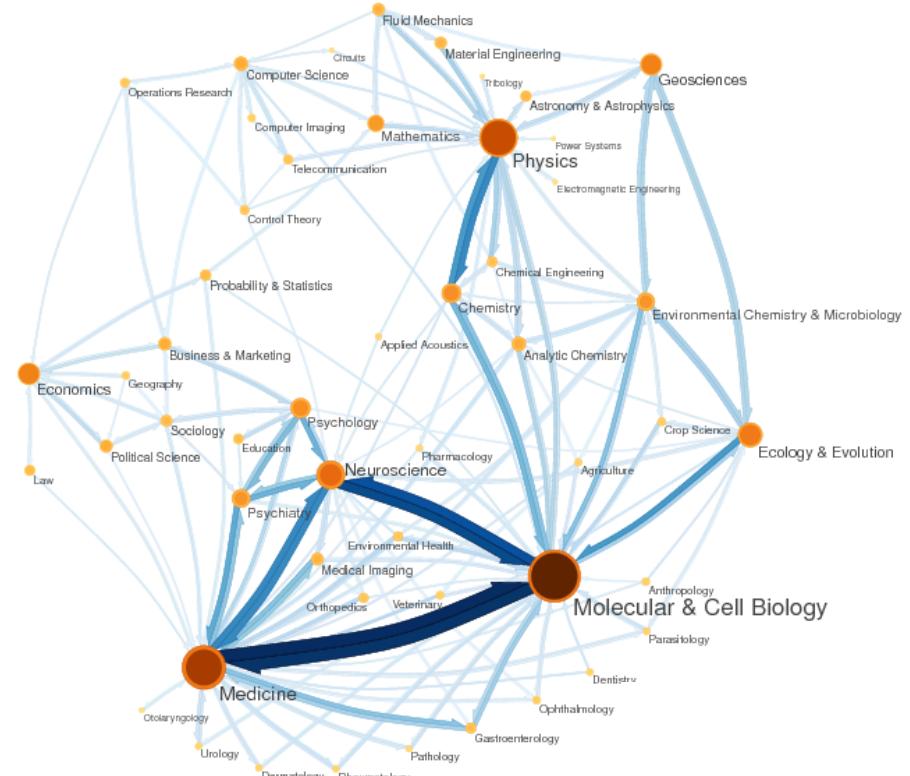




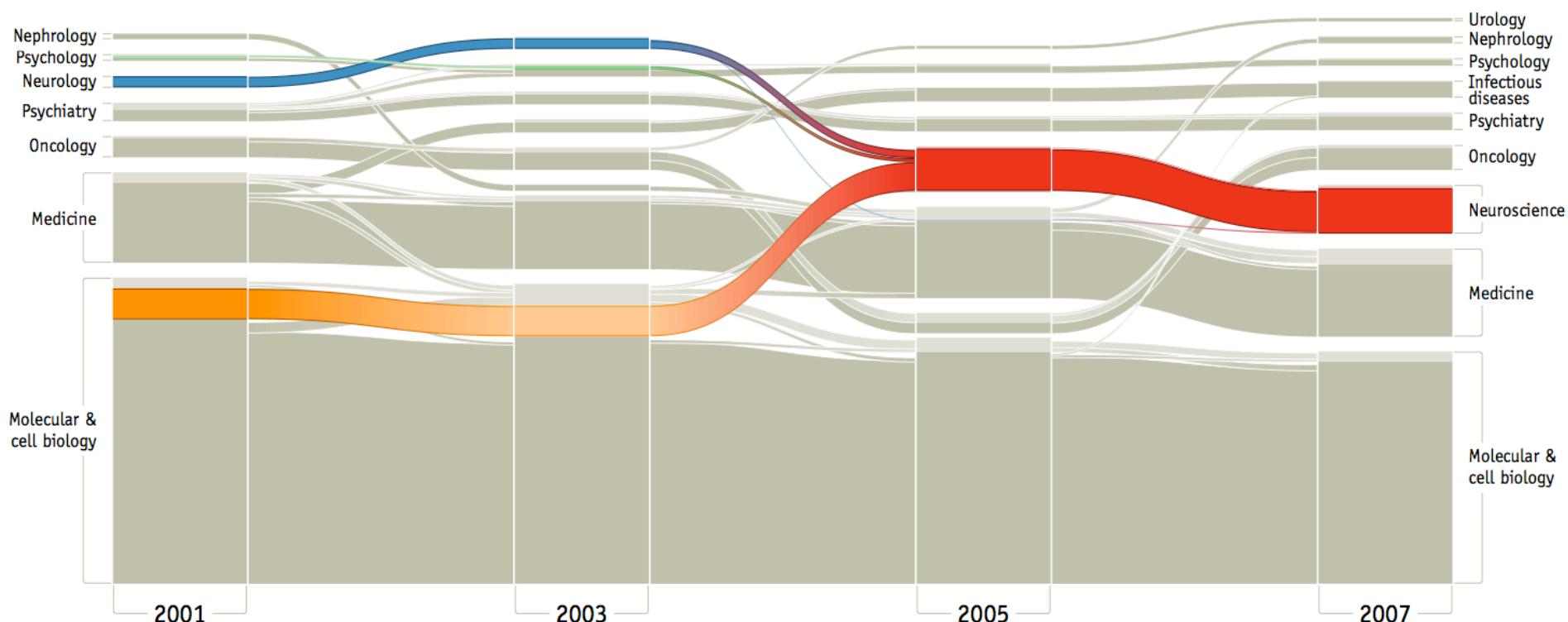
1995

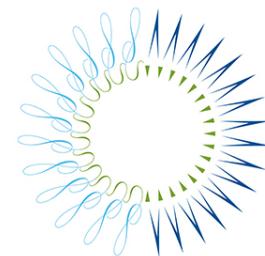


2004

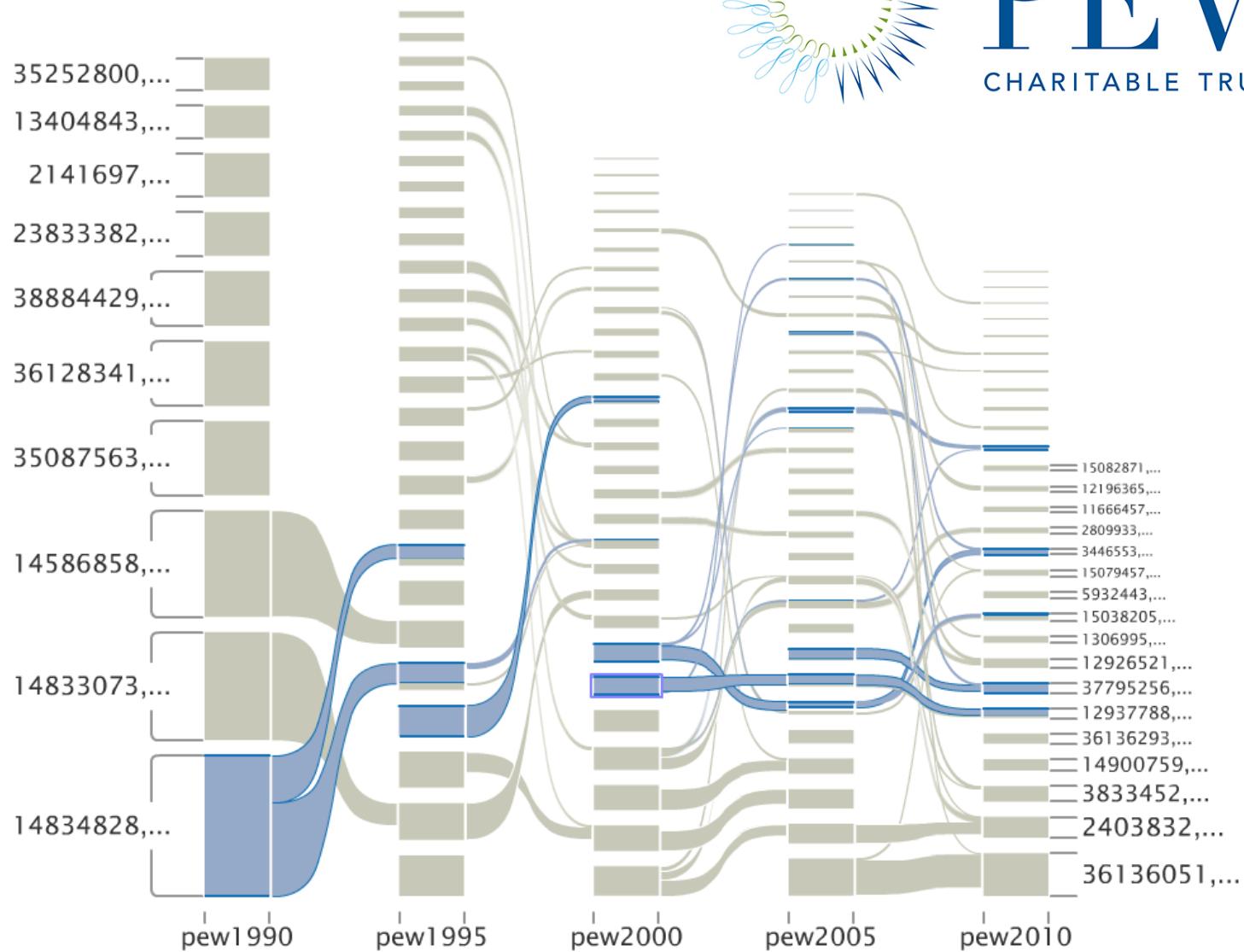


Citation networks over time

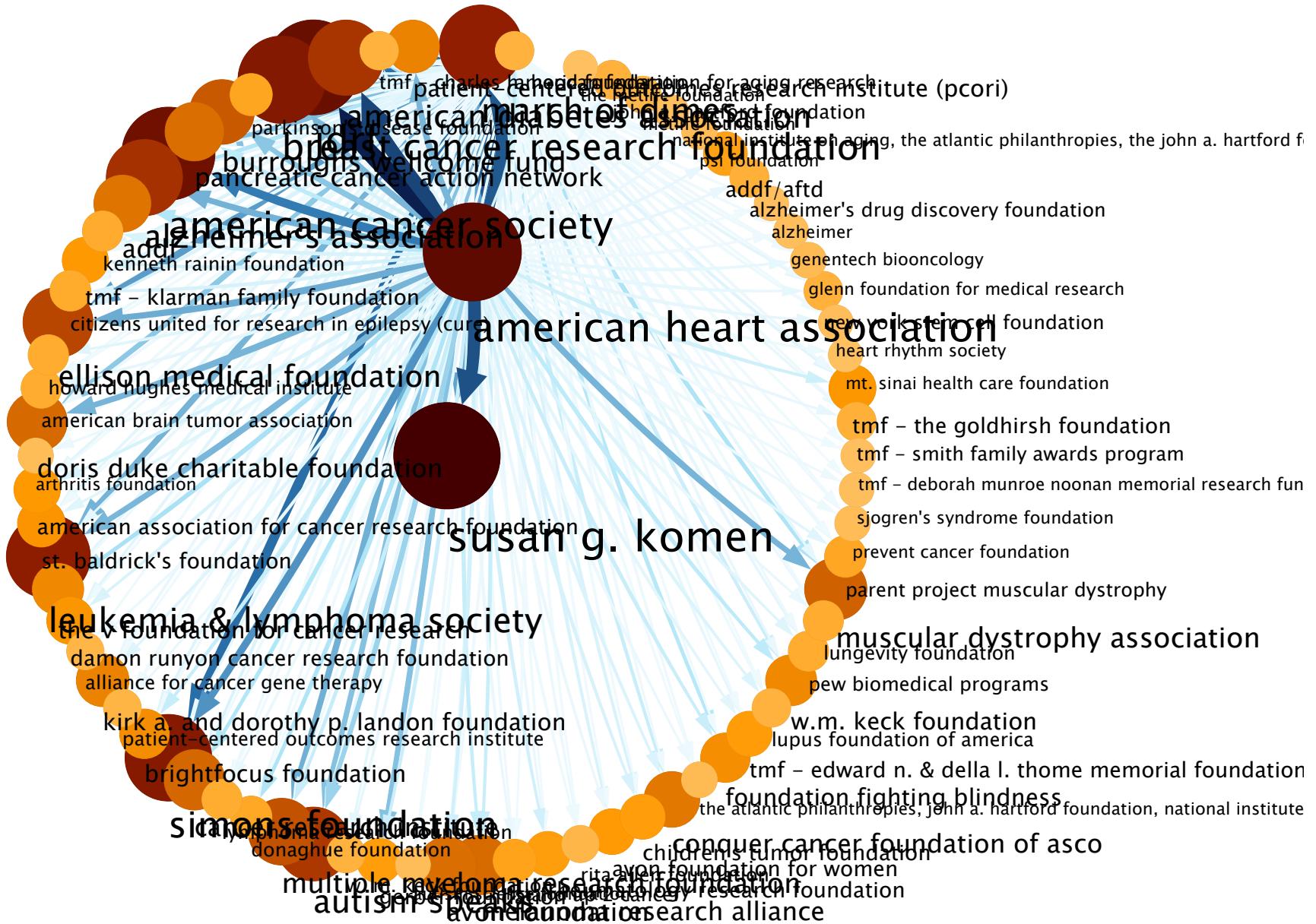




THE
PEW
CHARITABLE TRUSTS



Health Research Alliance



Hierarchical Mapping



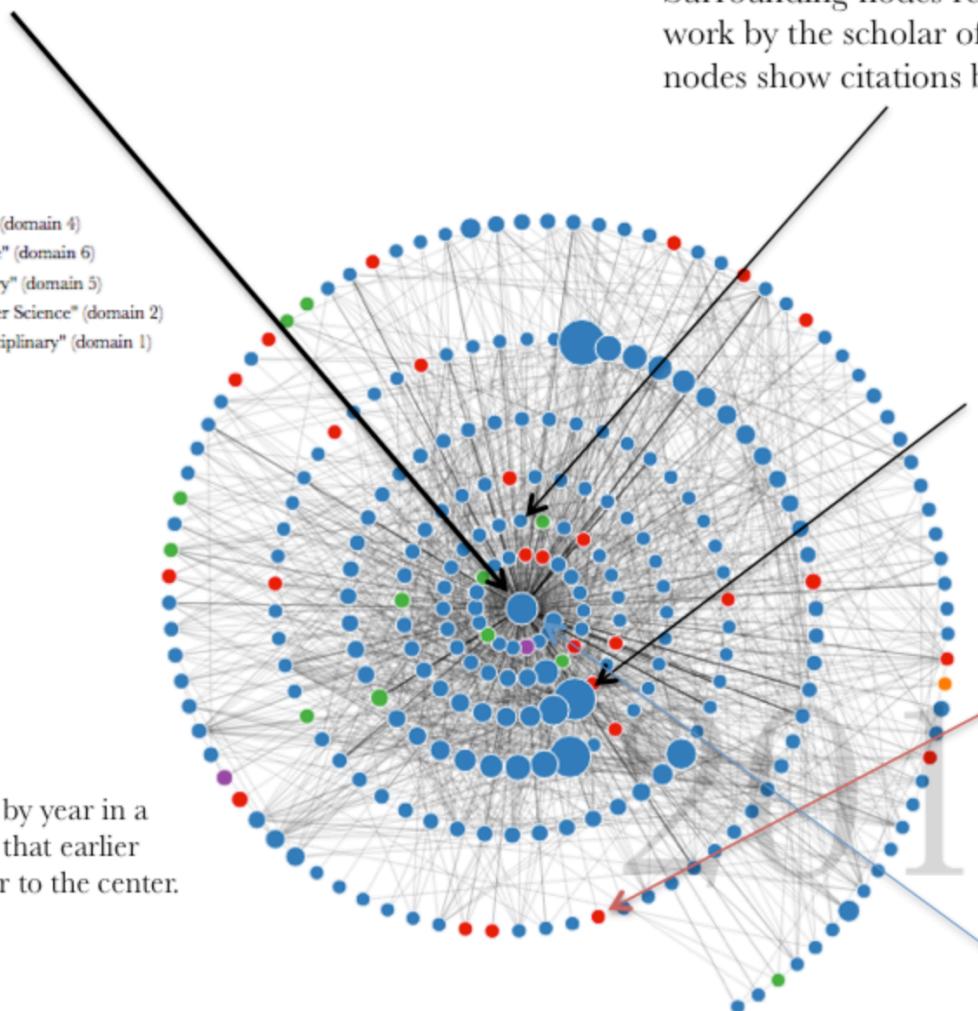
Pew Influence

The **center node** represents all of the papers authored by the scholar of interest.

Surrounding nodes represent papers that have cited work by the scholar of interest. Lines between the nodes show citations between papers.

- Papers in category "Biology" (domain 4)
- Papers in category "Medicine" (domain 6)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Computer Science" (domain 2)
- Papers in category "Multidisciplinary" (domain 1)

Papers are revealed by year in a spiral formation, so that earlier papers appear closer to the center.



Showing a scholar's influence

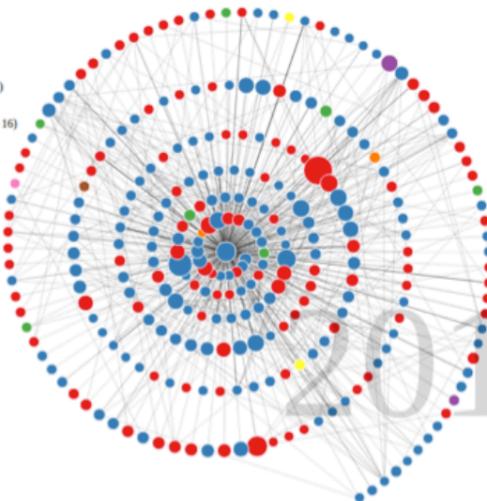
The size of each node is scaled by the *Eigenfactor score* of that paper—a metric of influence that takes into account its position in the total citation network. Bigger nodes represent the most influential papers that have cited the central scholar.

The color of each node shows the academic discipline of the paper. A more colorful network means that the impact of the central scholar's work has extended out to a wider range of fields.

The **color** of the **center node** represents the dominant field of the central scholar—the most common field of all the scholar's publications.

Comparing Authors

- Papers in category "Medicine" (domain 6)
- Papers in category "Biology" (domain 4)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Engineering" (domain 8)
- Papers in category "Material Science" (domain 12)
- Papers in category "Physics" (domain 19)
- Papers in category "Agriculture Science" (domain 16)
- Papers in category "Social Science" (domain 22)

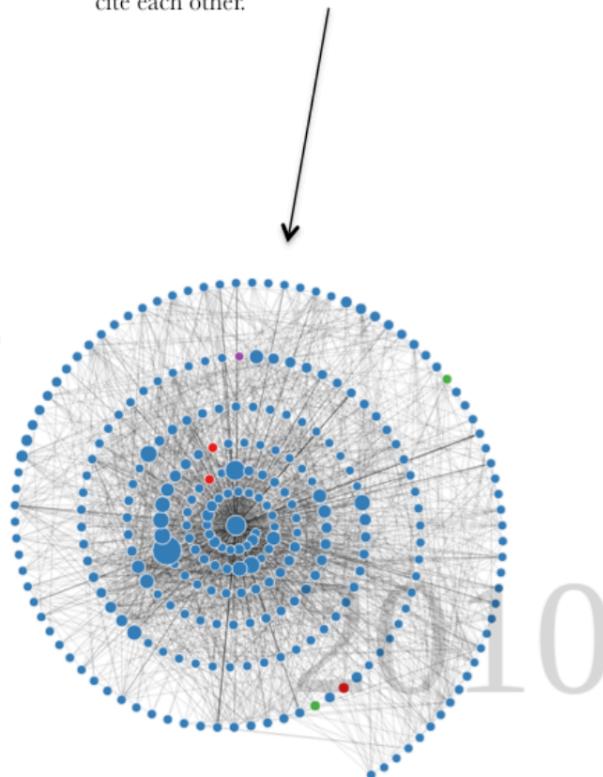


A denser network means that the papers that cite the central author also tend to cite each other.

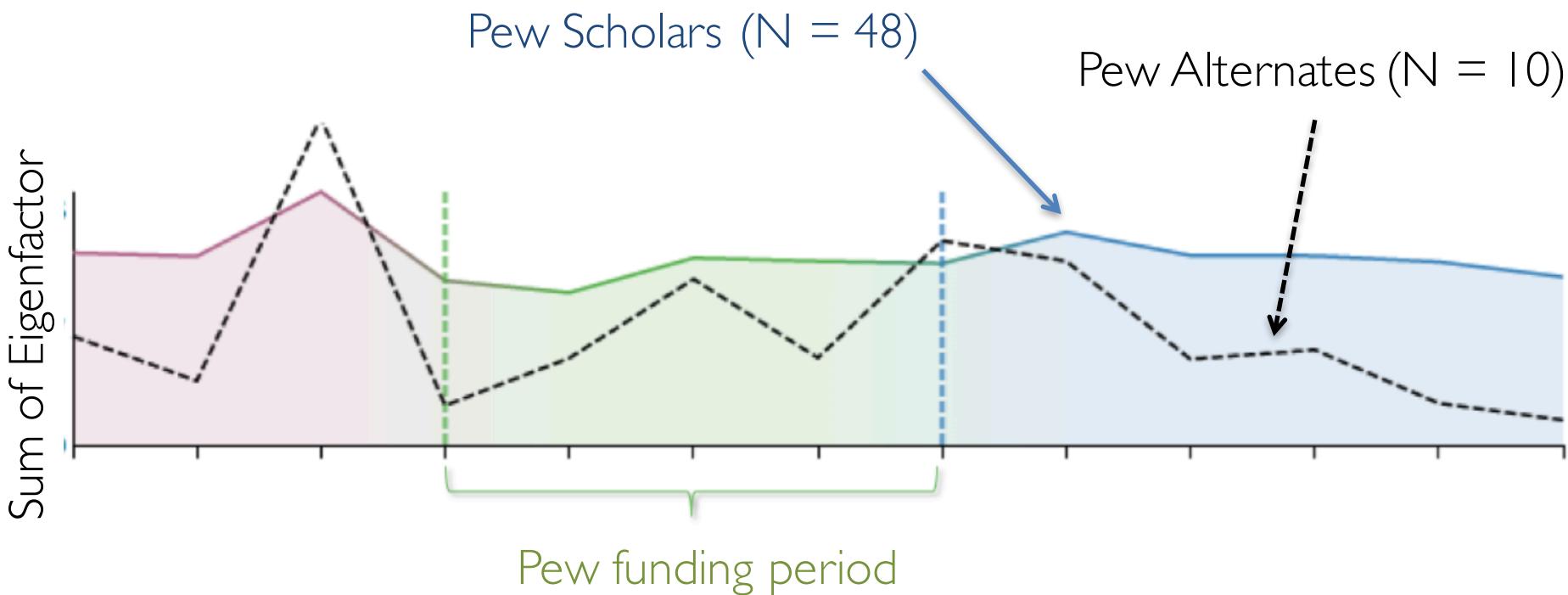


A more sparse network indicates fewer citations between papers shown in the network. This could be a result of the central scholar having impact across a wider set of academic communities.

- Papers in category "Biology" (domain 4)
- Papers in category "Medicine" (domain 6)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Social Science" (domain 22)



Comparing Alternates



* Includes scholars and alternates from cohort years: 1997, 1999, 2000, 2001, 2002

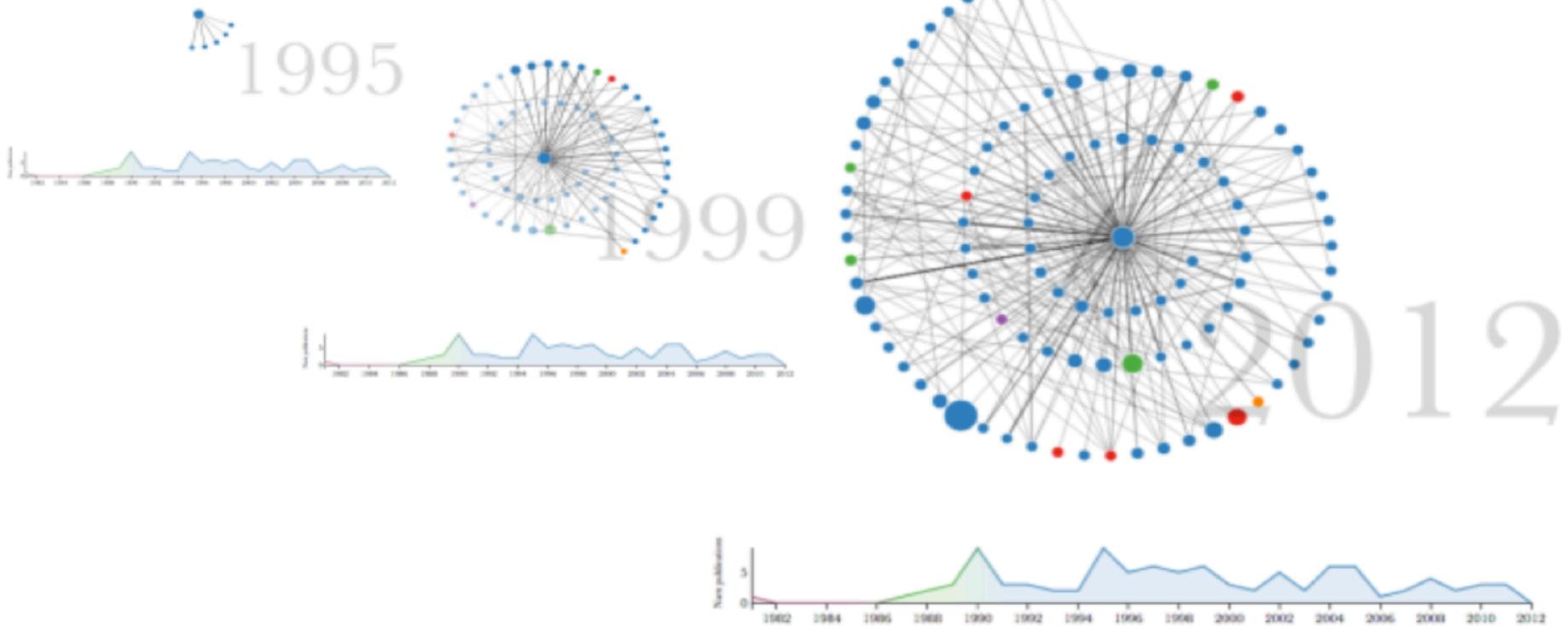
Data I need from you...

Future Directions

- Regression discontinuity design analysis (pew scholars versus alternates)
- Integrate interviews from Chemical Heritage Foundation
- Automated narration of visualization
- Author disambiguation and further data cleaning
- User studies for improved hypothesis generation
- Personalize visualization for different stakeholders (funders, researchers, donors)
- Distinguish reviews, model organisms, funding agency
- Basic research versus applied research
- Individual grants versus collaborative grants

Explore the data

scholar.eigenfactor.org



* Please use Chrome web browser for best results

Acknowledgements

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Jody Roberts , Chemical Heritage Foundation

Martin Rosvall, Department of Physics, Umea University

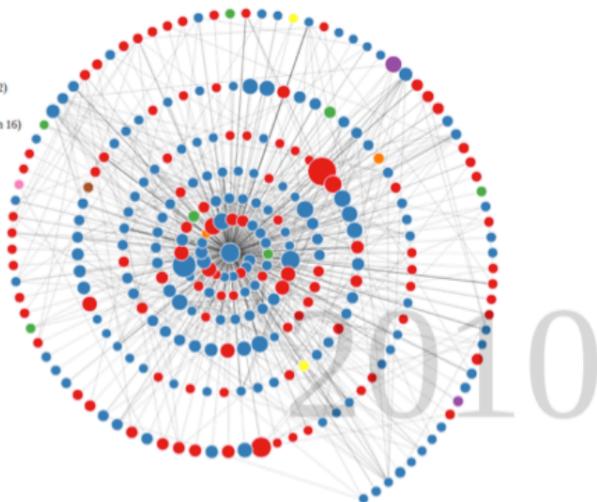
Carl Bergstrom, Department of Biology, University of Washington

Visualizing Interdisciplinarity



Jason Portenoy

- Papers in category "Medicine" (domain 6)
- Papers in category "Biology" (domain 4)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Engineering" (domain 8)
- Papers in category "Material Science" (domain 12)
- Papers in category "Physics" (domain 19)
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