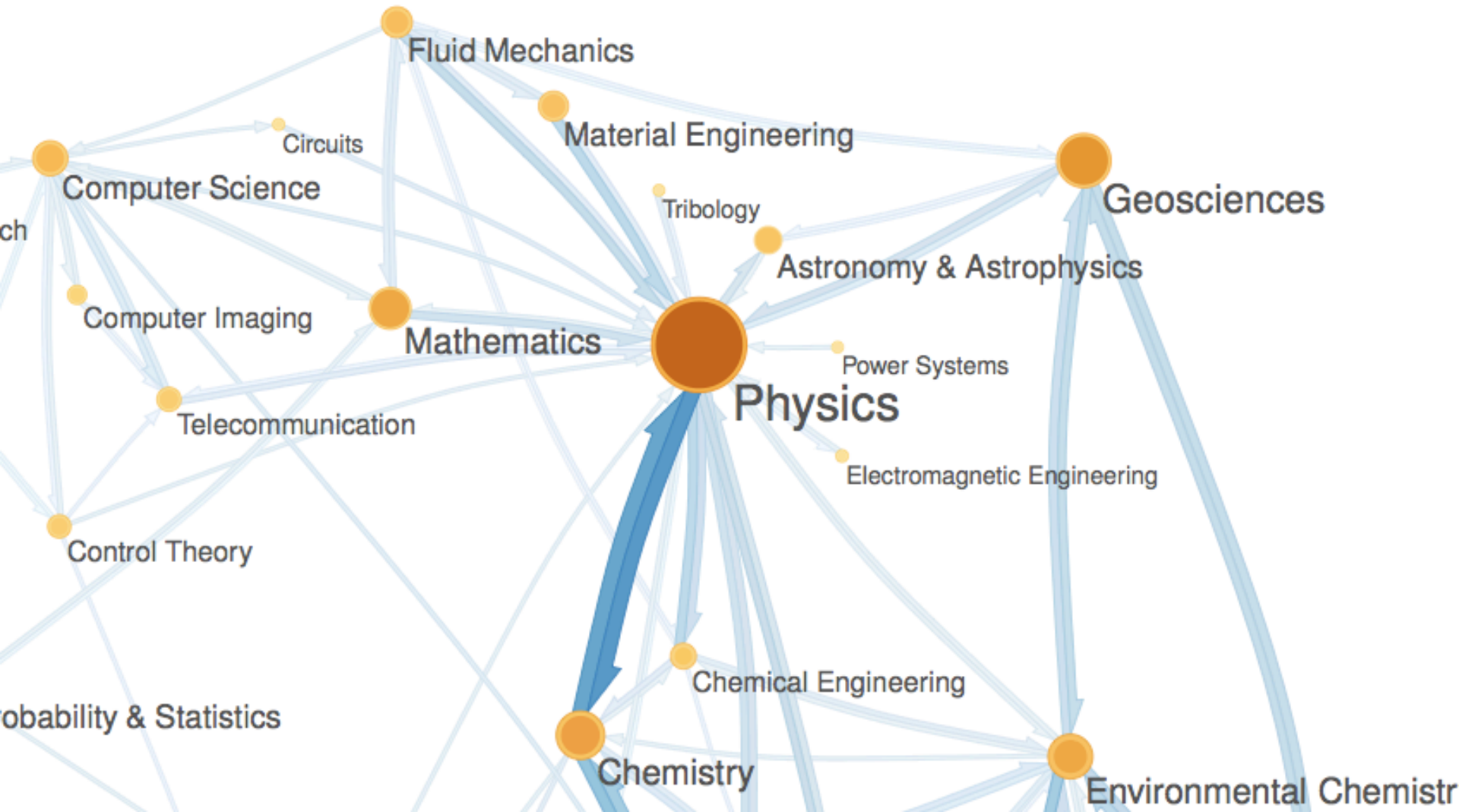
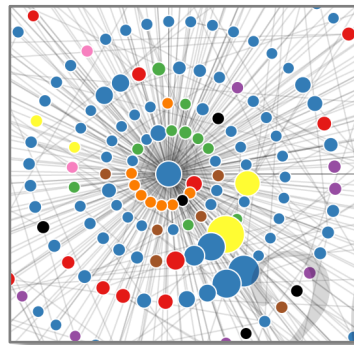


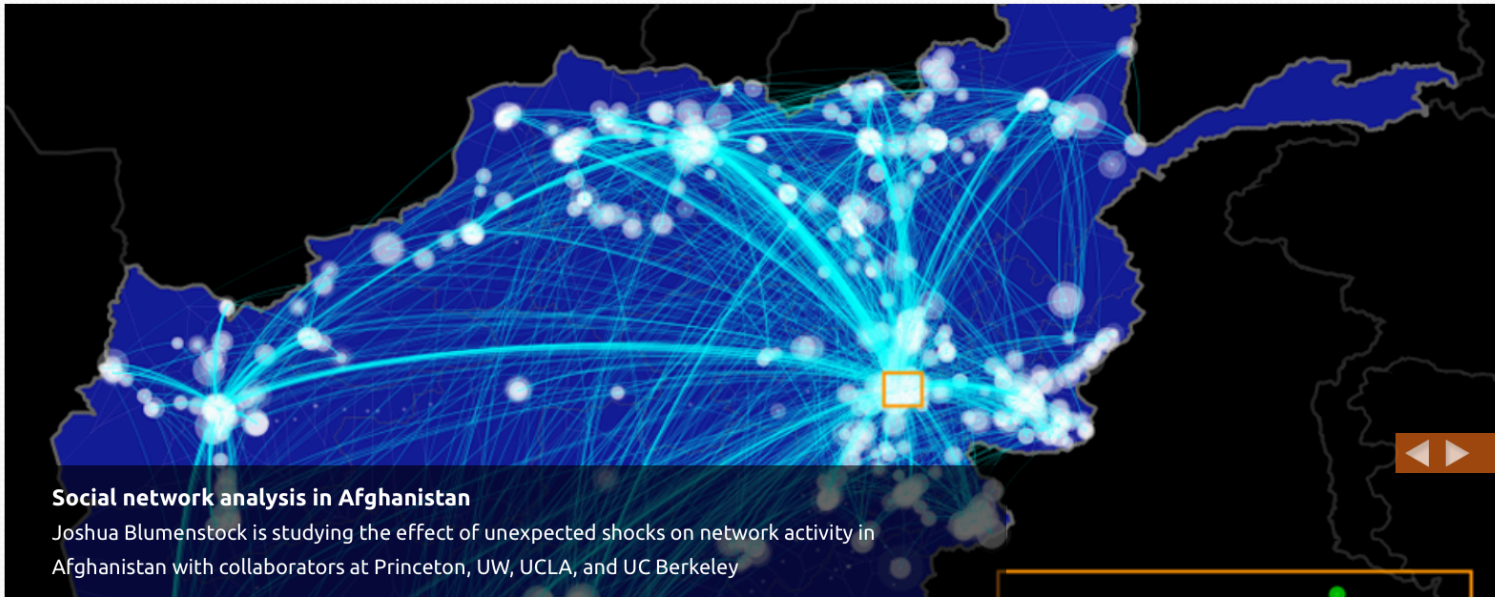
The Science of Science

Jevin West, Information School, University of Washington



Science of Science





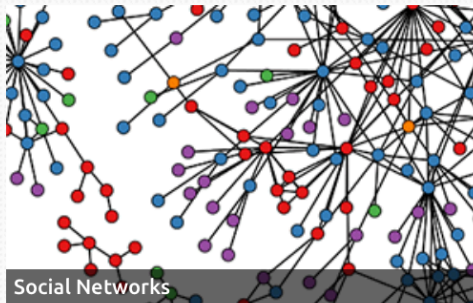
Social network analysis in Afghanistan

Joshua Blumenstock is studying the effect of unexpected shocks on network activity in Afghanistan with collaborators at Princeton, UW, UCLA, and UC Berkeley

Research Focus Areas



Data for Development



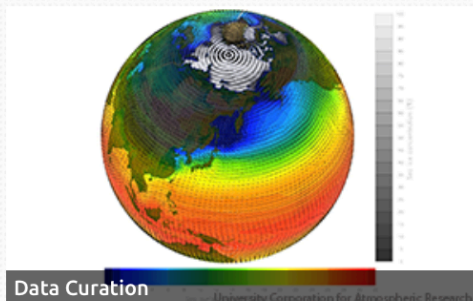
Social Networks



Data Visualization



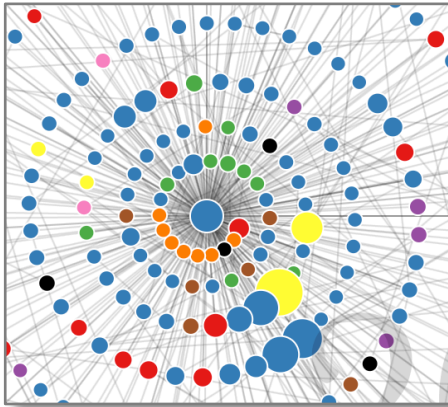
Computational Social Science



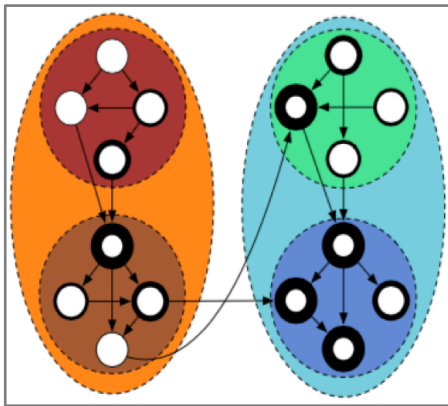
Data Curation



Science of Science



Knowledge Science

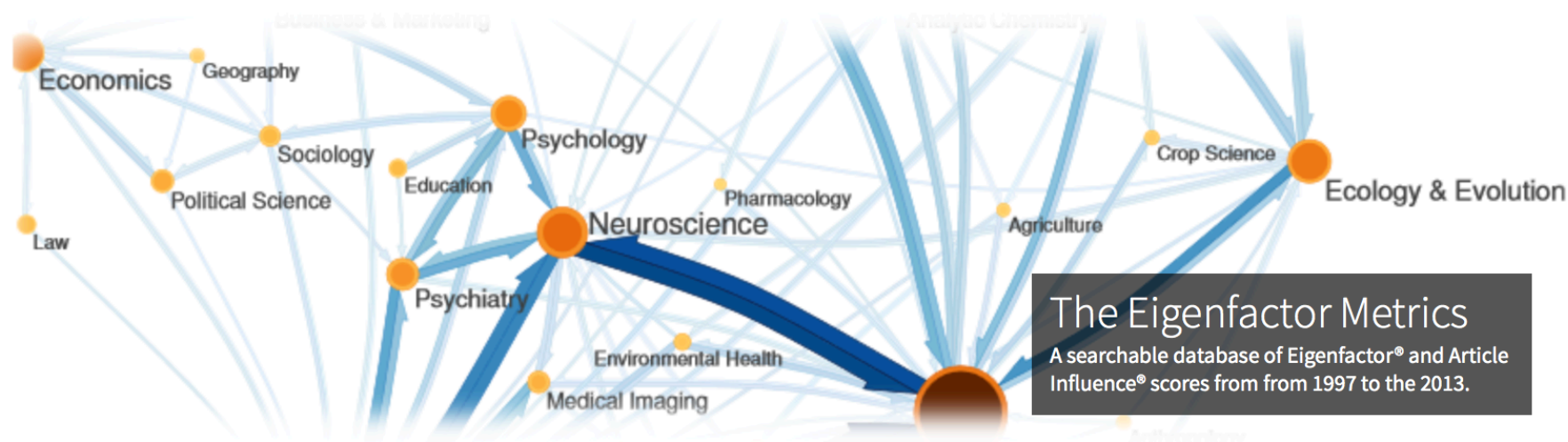


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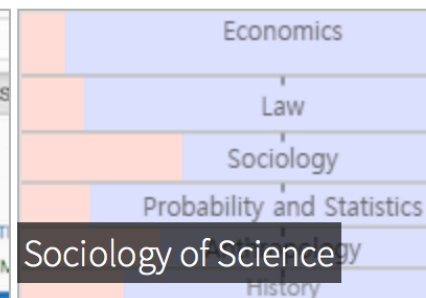
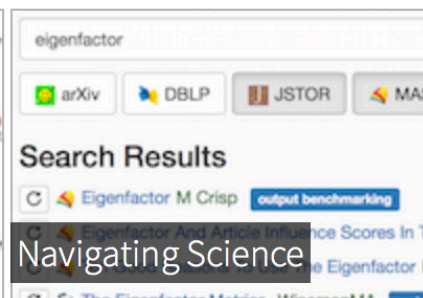
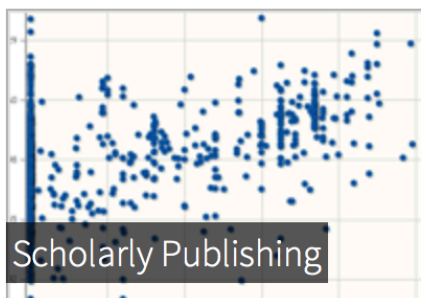
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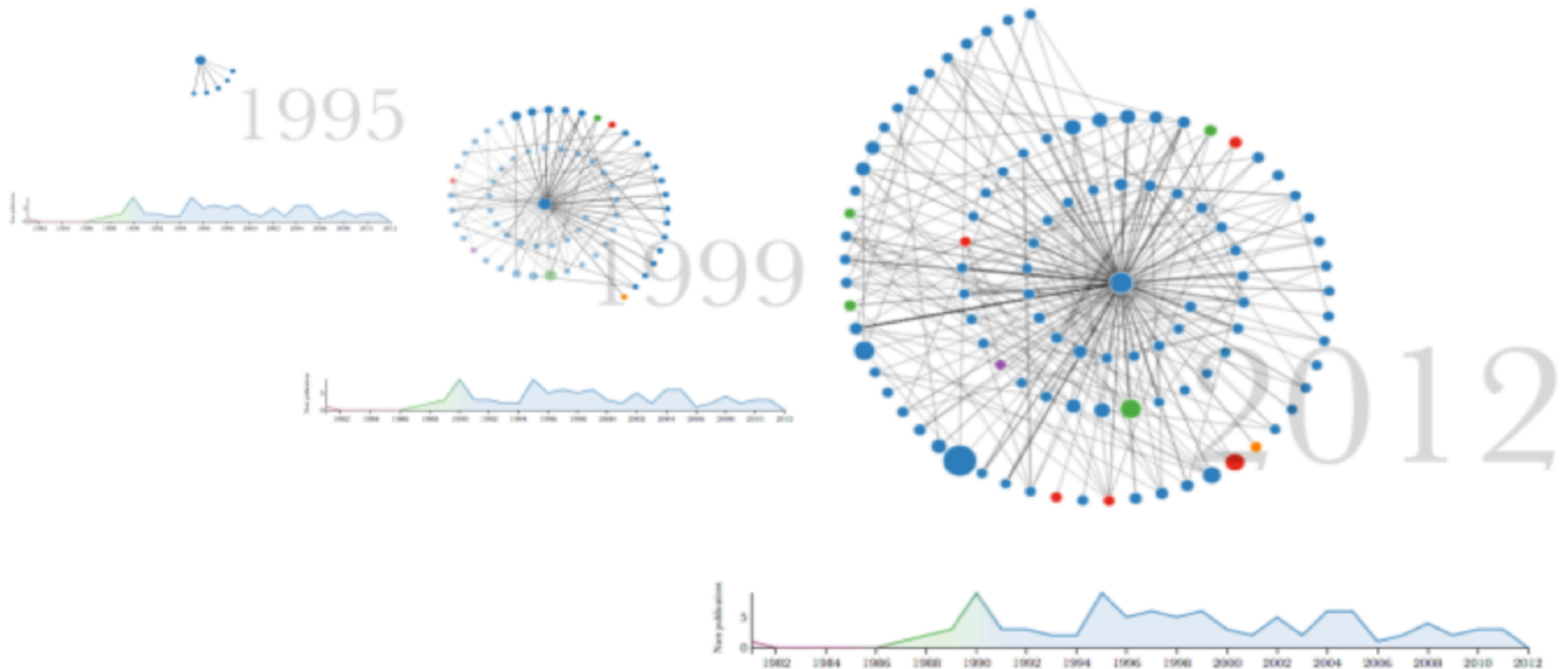
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Maladies of Science...

IS THERE A REPRODUCIBILITY CRISIS?



Essay

Why Most Published Research Findings Are False

John P. A. Ioannidis

Summary

There is increasing concern that most current published research findings are false. The probability that a research claim is true may depend on study power and bias, the number of other studies on the same question, and, importantly, the ratio of true to no relationships among the relationships probed in each scientific field. In this framework, a research finding is less likely to be true when the studies conducted in a field are smaller; when effect sizes are smaller; when there is a greater number and lesser preselection of tested relationships; where there is greater flexibility in designs, definitions, outcomes, and analytical modes; when there is greater financial and other interest and prejudice; and when more teams are involved in a scientific field in chase of statistical significance. Simulations show that for most study designs and settings, it is more likely for a research claim to be false than true. Moreover, for many current scientific fields, claimed research findings may often be simply accurate measures of the prevailing bias. In this essay, I discuss the implications of these problems for the conduct and interpretation of research.

factors that influence this problem and some corollaries thereof.

Modeling the Framework for False Positive Findings

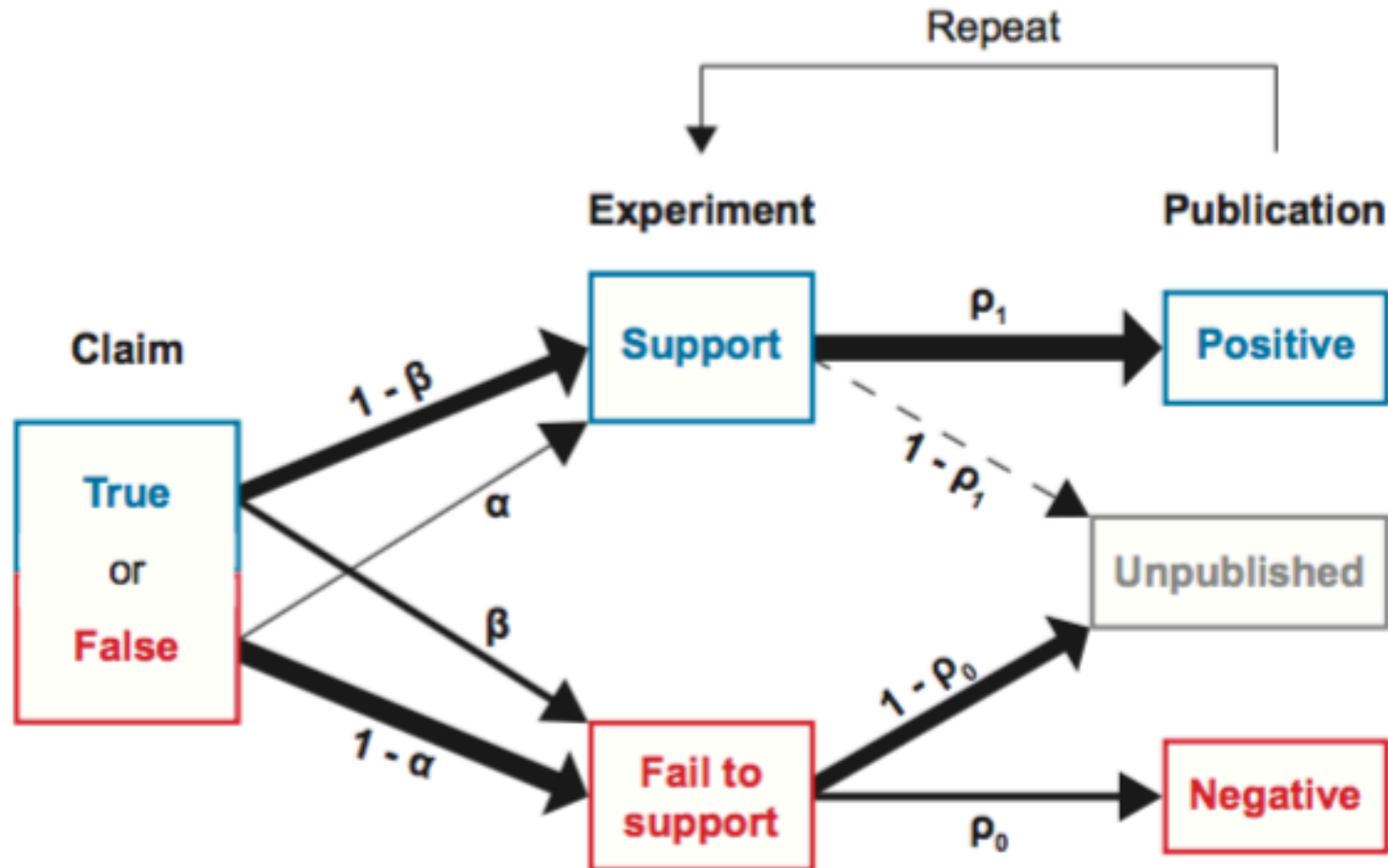
Several methodologists have pointed out [9–11] that the high rate of nonreplication (lack of confirmation) of research discoveries is a consequence of the convenient, yet ill-founded strategy of claiming conclusive research findings solely on the basis of a single study assessed by formal statistical significance, typically for a p -value less than 0.05. Research is not most appropriately represented and summarized by p -values, but, unfortunately, there is a widespread notion that medical research articles

It can be proven that most claimed research findings are false.

should be interpreted based only on p -values. Research findings are defined here as any relationship reaching formal statistical significance, e.g., effective interventions, informative predictors, risk factors, or associations. “Negative” research is also very useful.

is characteristic of the field and can vary a lot depending on whether the field targets highly likely relationships or searches for only one or a few true relationships among thousands and millions of hypotheses that may be postulated. Let us also consider, for computational simplicity, circumscribed fields where either there is only one true relationship (among many that can be hypothesized) or the power is similar to find any of the several existing true relationships. The pre-study probability of a relationship being true is $R/(R+1)$. The probability of a study finding a true relationship reflects the power $1 - \beta$ (one minus the Type II error rate). The probability of claiming a relationship when none truly exists reflects the Type I error rate, α . Assuming that c relationships are being probed in the field, the expected values of the 2×2 table are given in Table 1. After a research finding has been claimed based on achieving formal statistical significance, the post-study probability that it is true is the positive predictive value, PPV. The PPV is also the complementary probability of what Wacholder et al. have called the false positive report probability [10]. According to the 2×2 table, one gets $PPV = (1 - R)R/(R$

Publication bias and the canonization of false facts



Maladies of Science

Process

Incentives

The **H-index** impact on science



Jure Leskovec

Professor of Computer Science, [Stanford University](#)
[Data mining](#), [Social Network Analysis](#), [Information Networks](#)
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ACM Transactions on the Web (TWEB) 1 (1), 5

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[Cost-effective outbreak detection in networks](#)

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[Meme-tracking and the dynamics of the news cycle](#)

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[Graph evolution: Densification and shrinking diameters](#)

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[Friendship and mobility: user movement in location-based social networks](#)

E Cho, SA Myers, J Leskovec
Proceedings of the 17th ACM SIGKDD international conference on Knowledge ...

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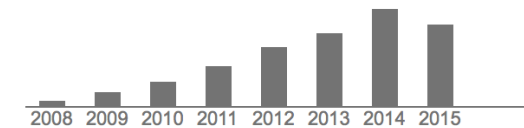
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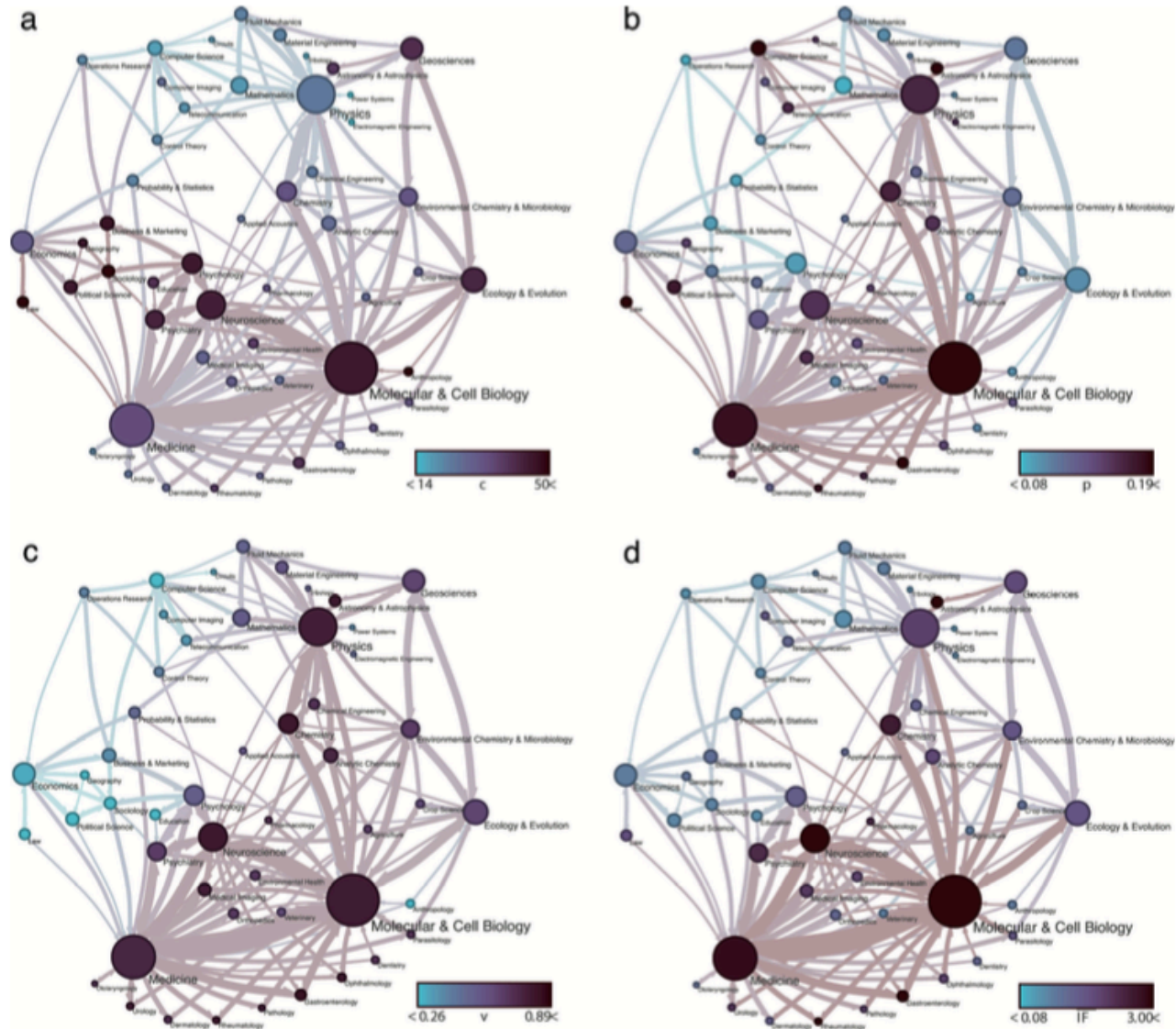
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Citations	19409	17853
h-index	59	56
i10-index	103	101



Evisceration of the H-index



Impact Factor Inflation



Impact factor drives...

The worse misuse of the H-index

You don't know me, but I am a fellow academic.....

I also noticed you have pretty good research prestige, as indicated by your **H-index** of 15. Even one paper of yours has been cited 170 times! This is great considering you are pretty new at academia.

My researcher prestige is not bad either, but I've been at it a bit longer: [LINK](#)

...

I know how hard it is for academics who are single to find a mate (I've been single for a long time). ...if there is an off-chance that you are single, please pick me. I like your work. It is very interesting and I think you are quite attractive. I would like to go out on a date with you, what do you say? Let's start a wonderful romance.

The San Francisco Declaration on Research Assessment (DORA), initiated by the American Society for Cell Biology (ASCB) together with a group of editors and publishers of scholarly journals, recognizes the need to improve the ways in which the outputs of scientific research are evaluated. The group met in December 2012 during the ASCB Annual Meeting in San Francisco and subsequently circulated a draft declaration among various stakeholders. DORA as it now stands has benefited from input by many of the original signers listed below. It is a worldwide initiative covering all scholarly disciplines. We encourage individuals and organizations who are concerned about the appropriate assessment of scientific research to sign DORA.

[Download the Declaration \(PDF\)](#)[Download the DORA Logo \(PDF\)](#)[Download the DORA Poster \(PDF\)](#)

San Francisco Declaration on Research Assessment

Putting science into the assessment of research

There is a pressing need to improve the ways in which the output of scientific research is evaluated by funding agencies, academic institutions, and other parties. To address this issue, a group of editors and publishers of scholarly journals met during the Annual Meeting of The American Society for Cell Biology (ASCB) in San Francisco, CA, on December 16, 2012. The group developed a set of recommendations, referred to as the San Francisco Declaration on

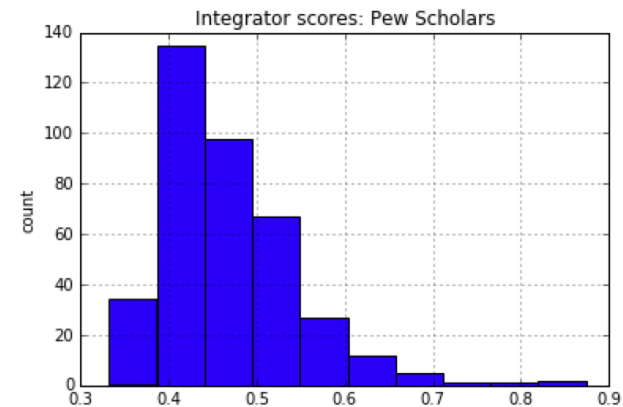
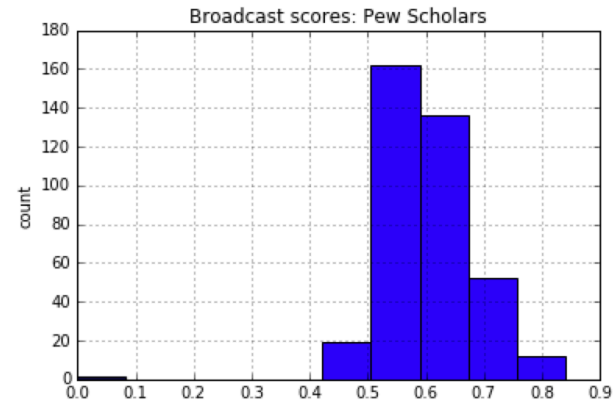
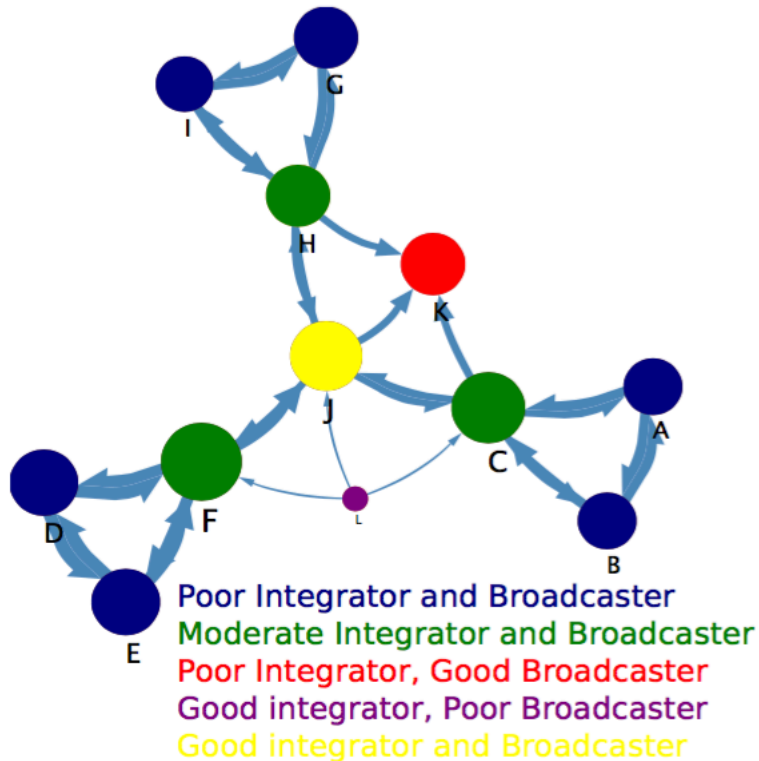
Stephen Curry on Why Universities Should Sign DORA

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Measuring Interdisciplinarity



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

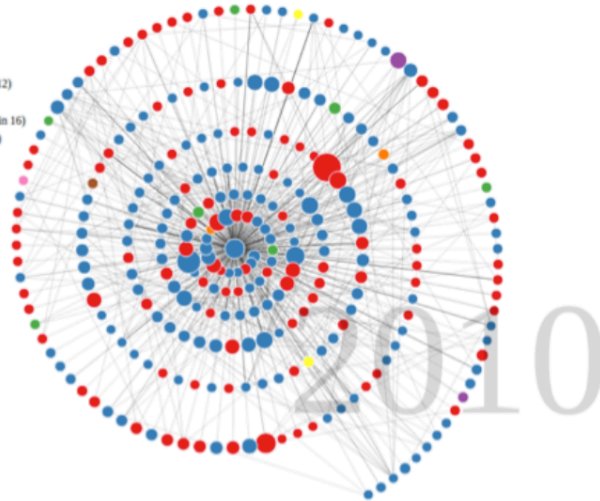
Moving from single metric summaries to interactive (hypothesis-driven) visualizations

Visualizing Influence



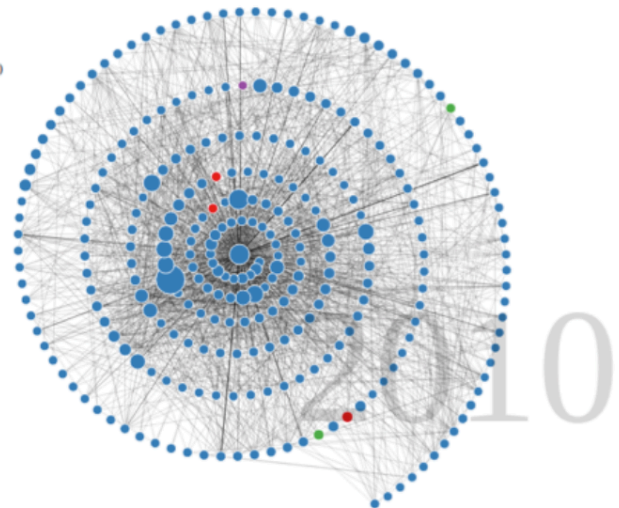
Jason Portenoy

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

- 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)



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.

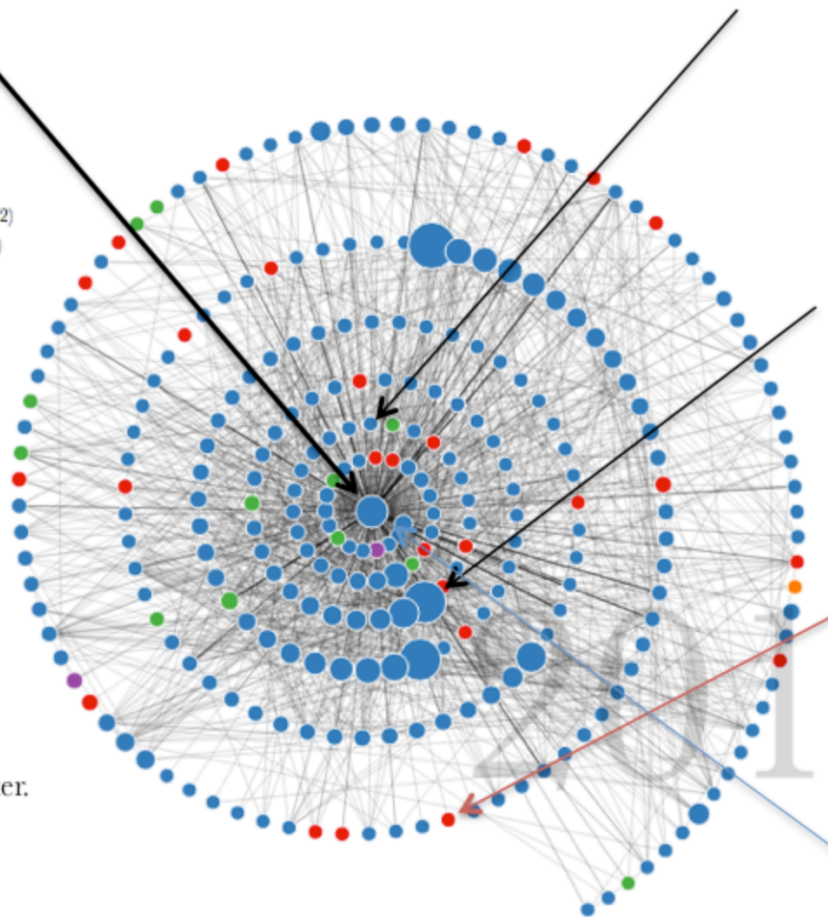
Pew Influence

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.

Visualizing Scholarly Influence Over Time

Influence of Pew Scholars

Roberta A. Gottlieb

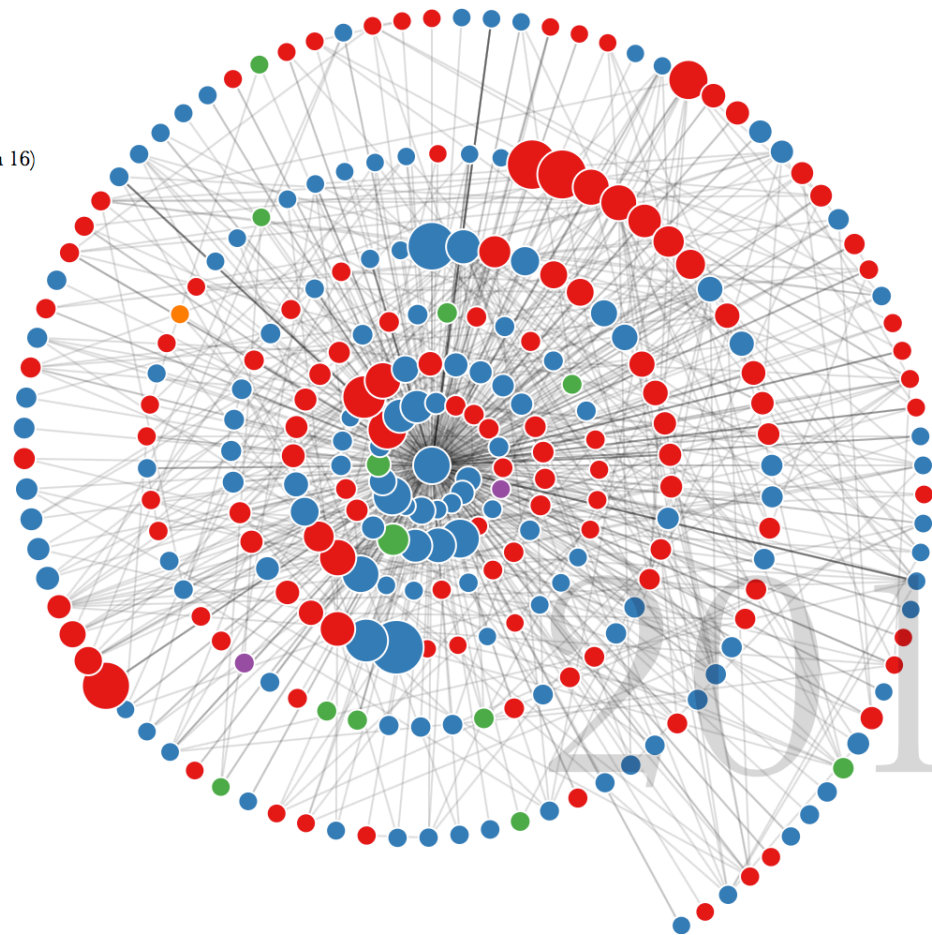
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- Papers in category "Medicine" (domain 6)
- Papers in category "Biology" (domain 4)
- Papers in category "Chemistry" (domain 5)
- Papers in category "Unknown" (domain 0)
- Papers in category "Agriculture Science" (domain 16)

Roberta A.
Gottlieb



Pew Scholar
1997



2012

Visualizing Scholarly Influence Over Time

Influence of Pew Scholars

Mark W. Grinstaff

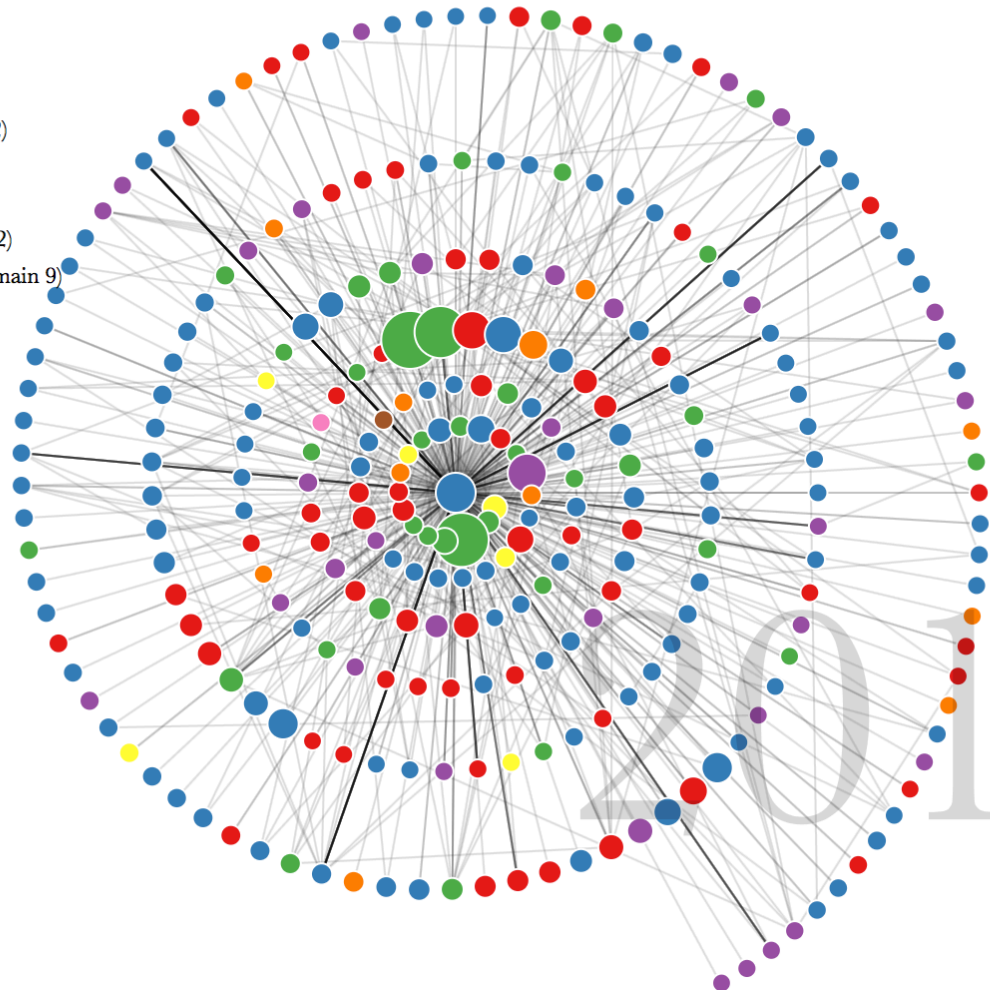
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- Papers in category "Chemistry" (domain 5)
- Papers in category "Medicine" (domain 6)
- Papers in category "Biology" (domain 4)
- Papers in category "Material Science" (domain 12)
- Papers in category "Engineering" (domain 8)
- Papers in category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

Mark W.
Grinstaff

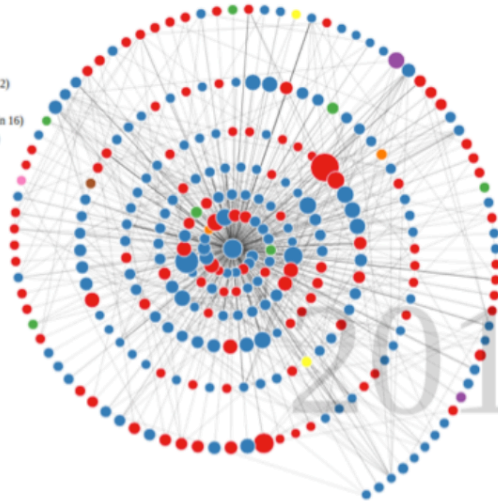


Pew Scholar
1999



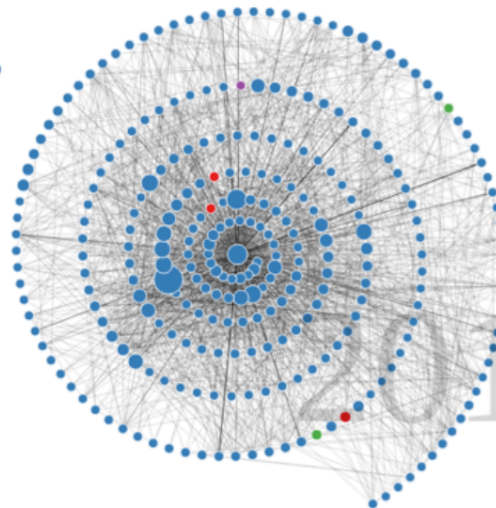
Comparing Authors

■ Papers in category "Medicine" (domain 6)
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■ 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.

■ 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)

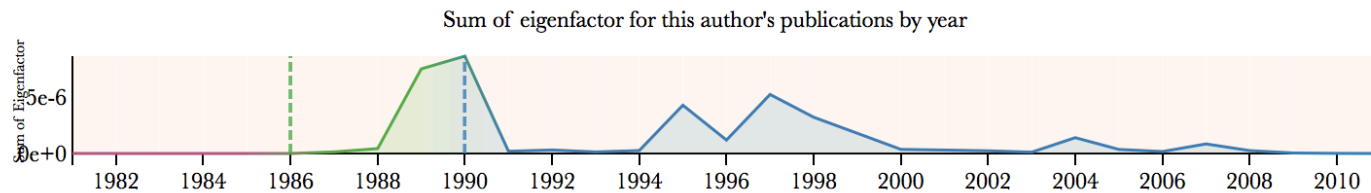
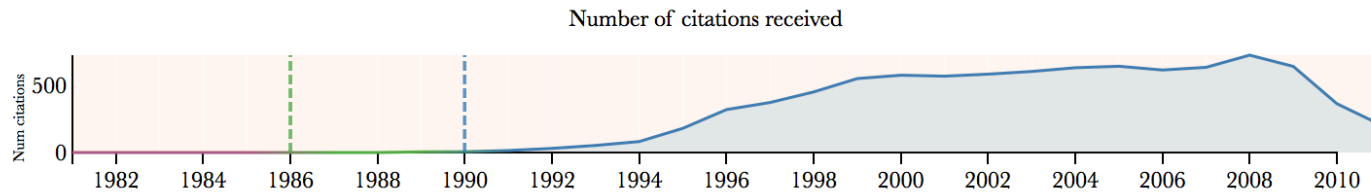
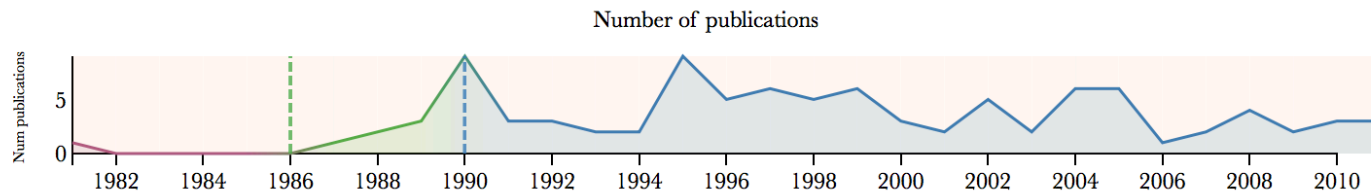
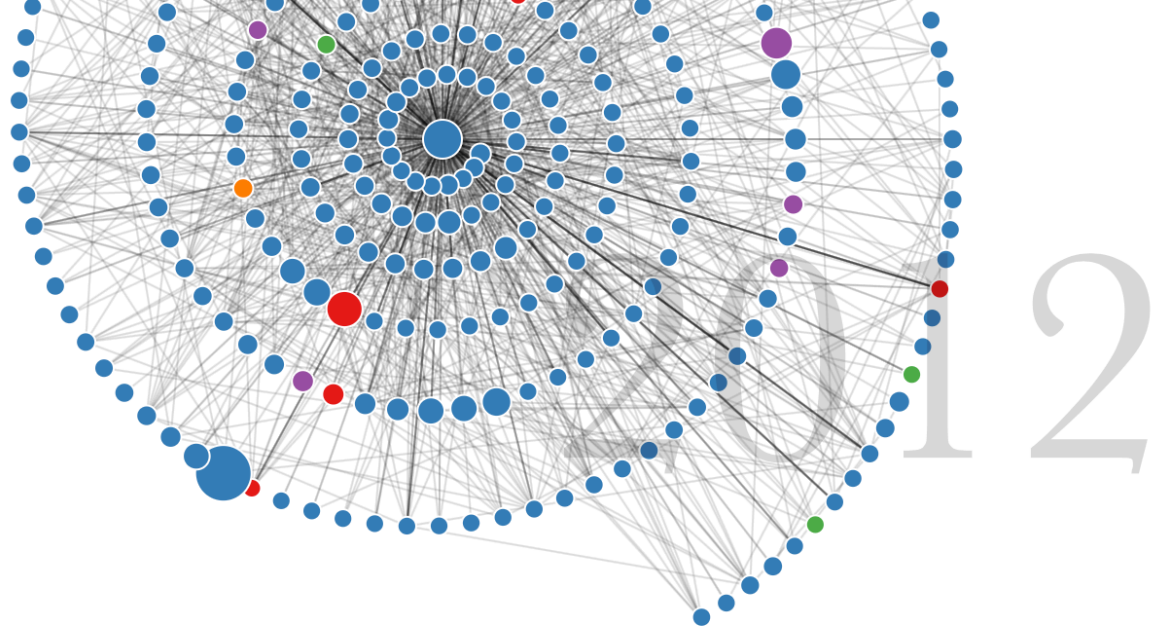


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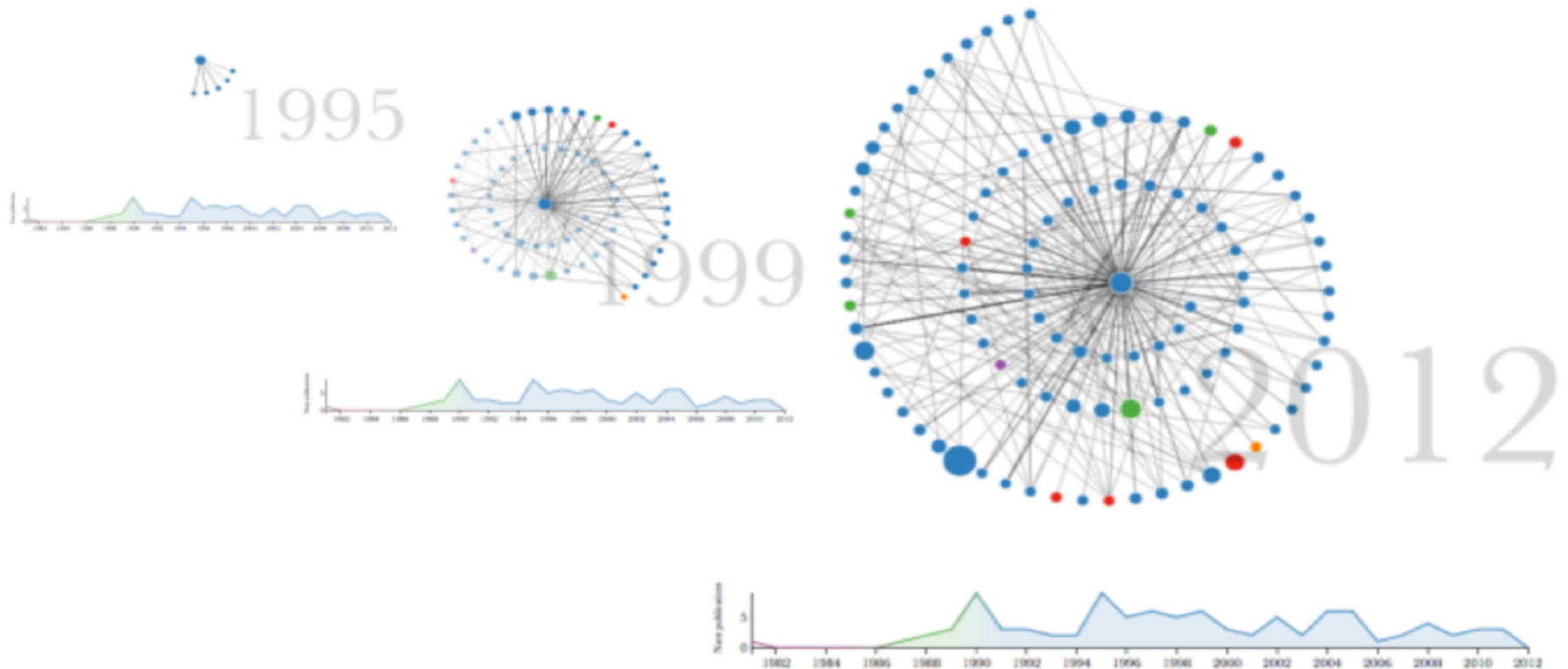
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Maladies of Science

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Incentives

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- Non-extensible
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[CT Bergstrom](#), [M Lo](#), [M Lipsitch](#) - [Proceedings of the ...](#), 2004 - [National Acad Sciences](#)

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Routine **cycling** of **antimicrobial** agents as an infection-control measure

[RA Weinstein](#), [SK Fridkin](#) - [Clinical infectious diseases](#), 2003 - [cid.oxfordjournals.org](#)




Abstract **Antimicrobial cycling** is the deliberate, scheduled removal and substitution of specific antimicrobials or classes of antimicrobials within an institutional environment (either




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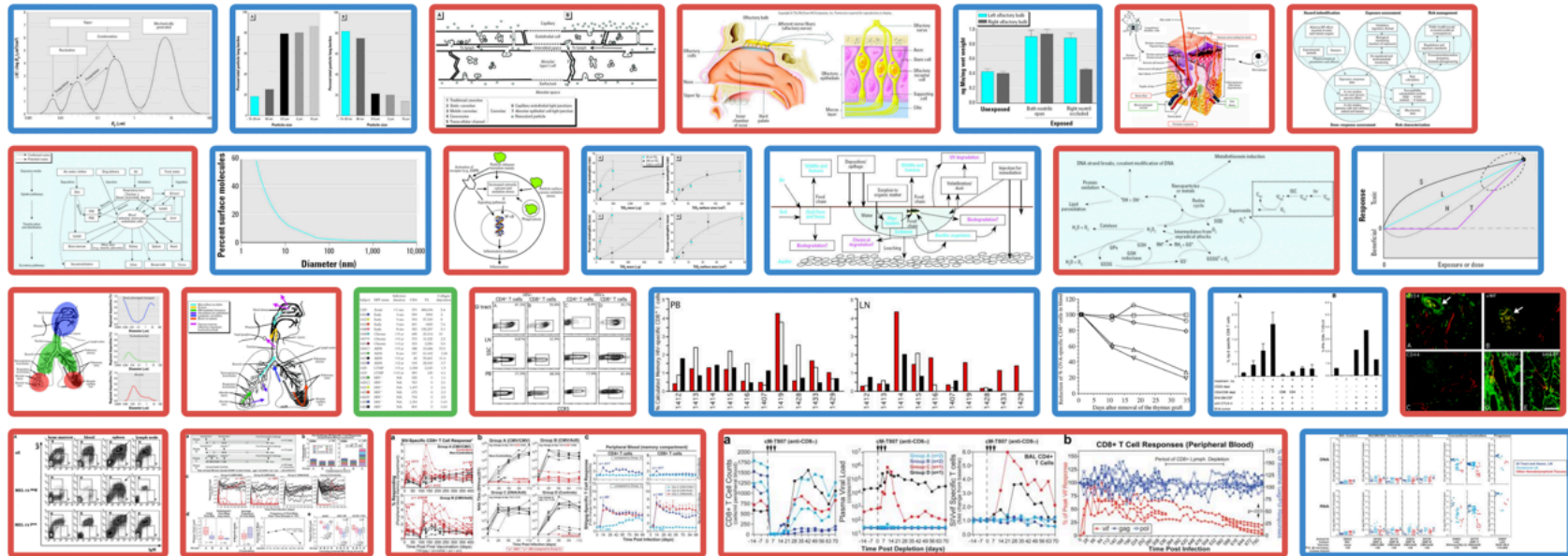


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Search

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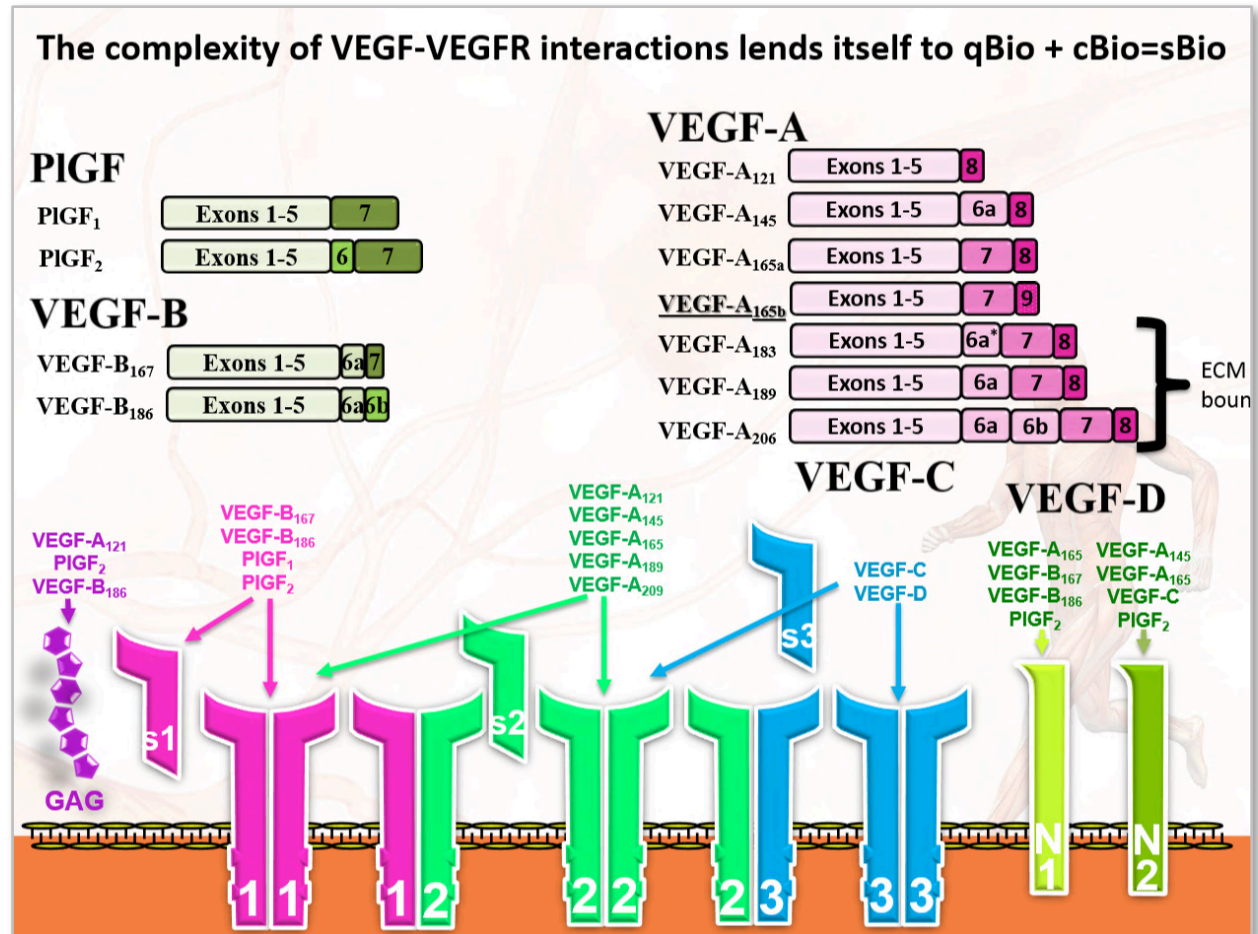


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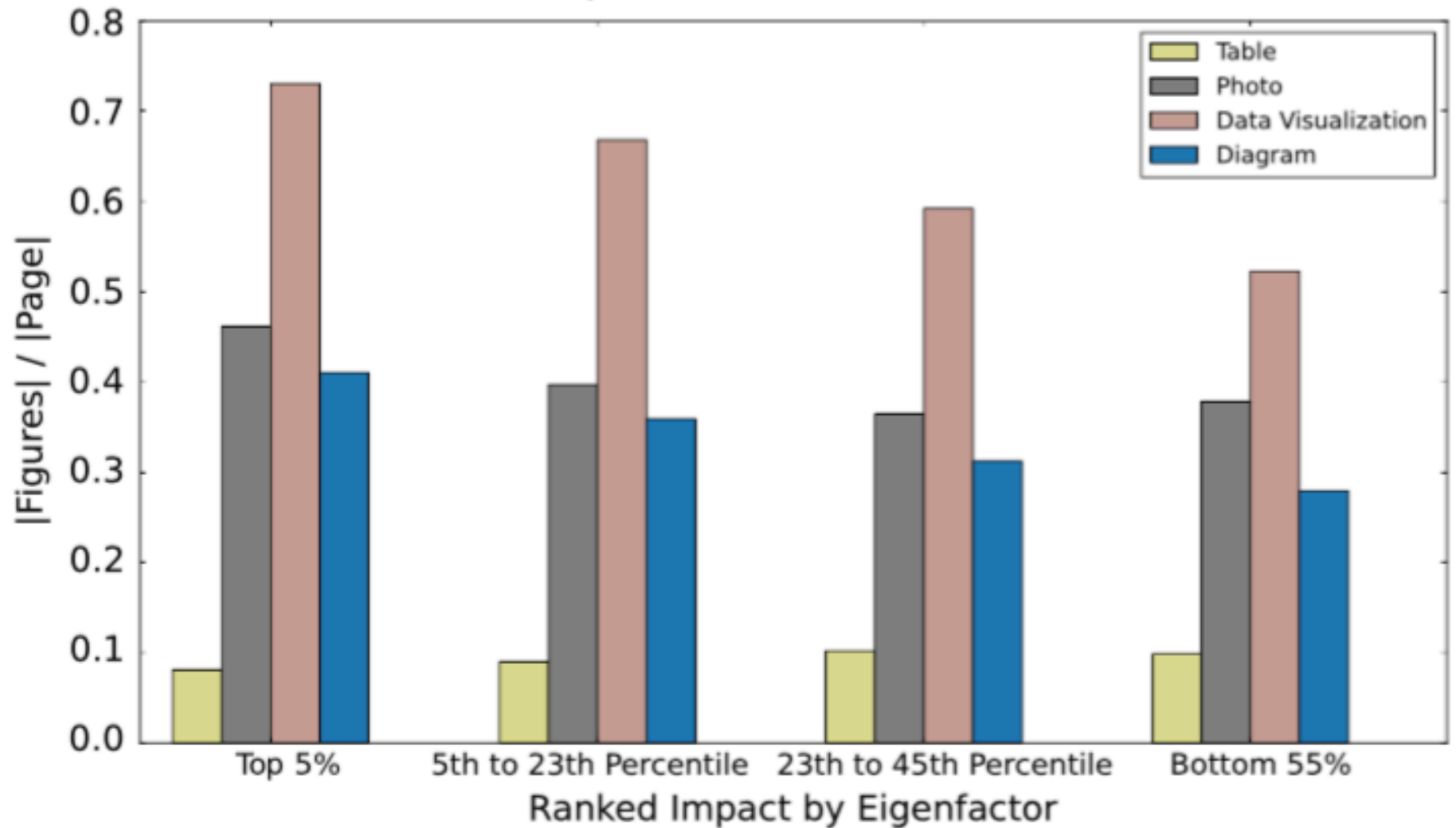
Mining for VEGF concentrations



P.I. Imoukhuede
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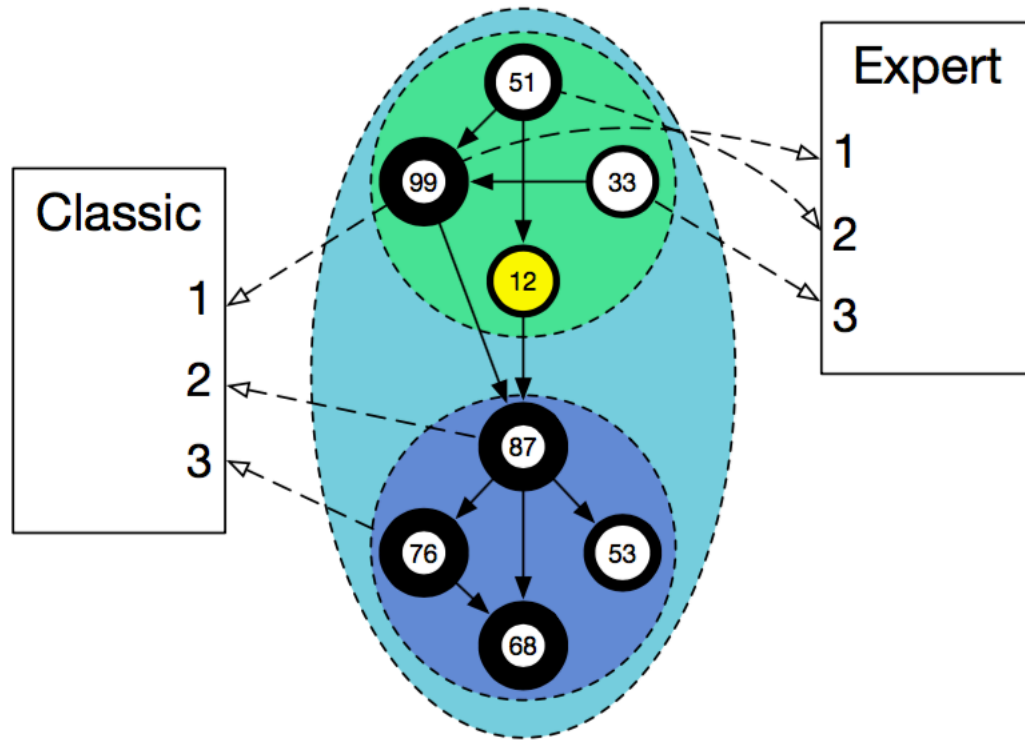


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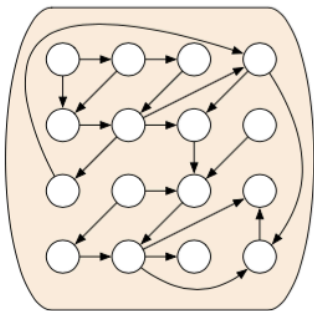


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<https://www.arxiv.org/pdf/1605.04951.pdf>

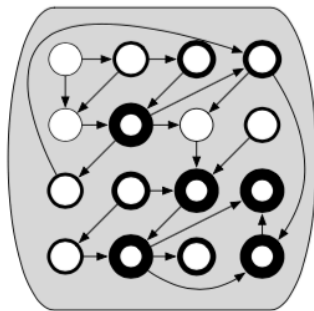
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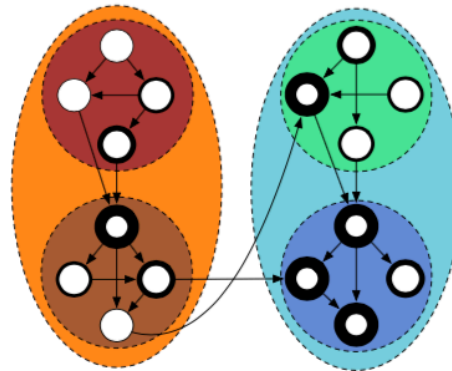
Assemble



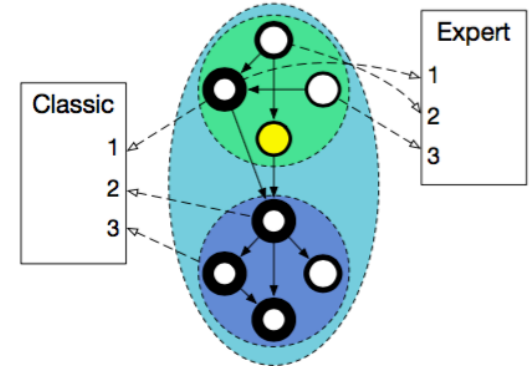
Rank



Cluster



Recommend



West, Wesley-Smith, Bergstrom (2016) A recommendation system based on hierarchical clustering of an article-level citation network. *IEEE, Transactions on Big Data* (in press)

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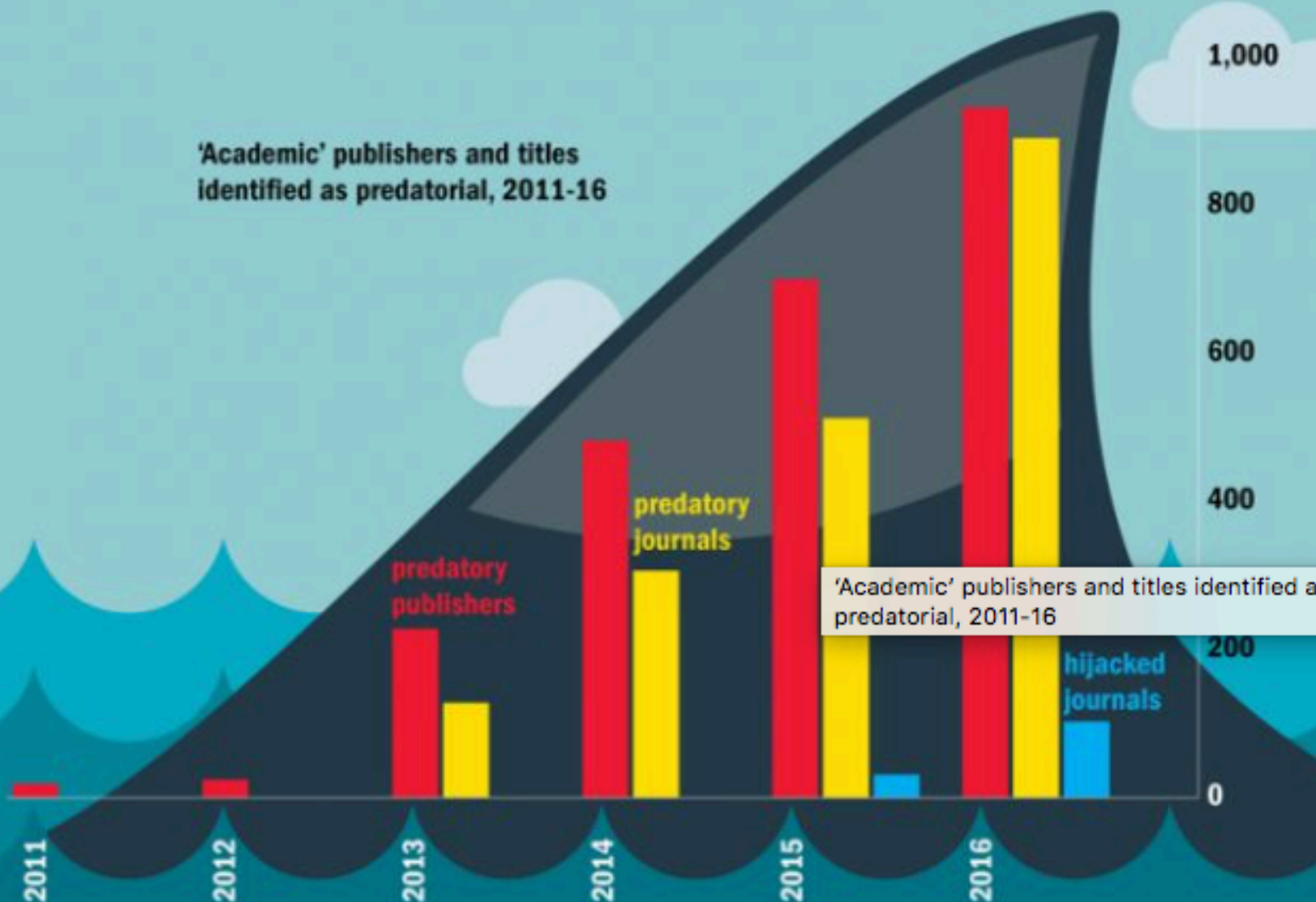
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Predatory Publishers



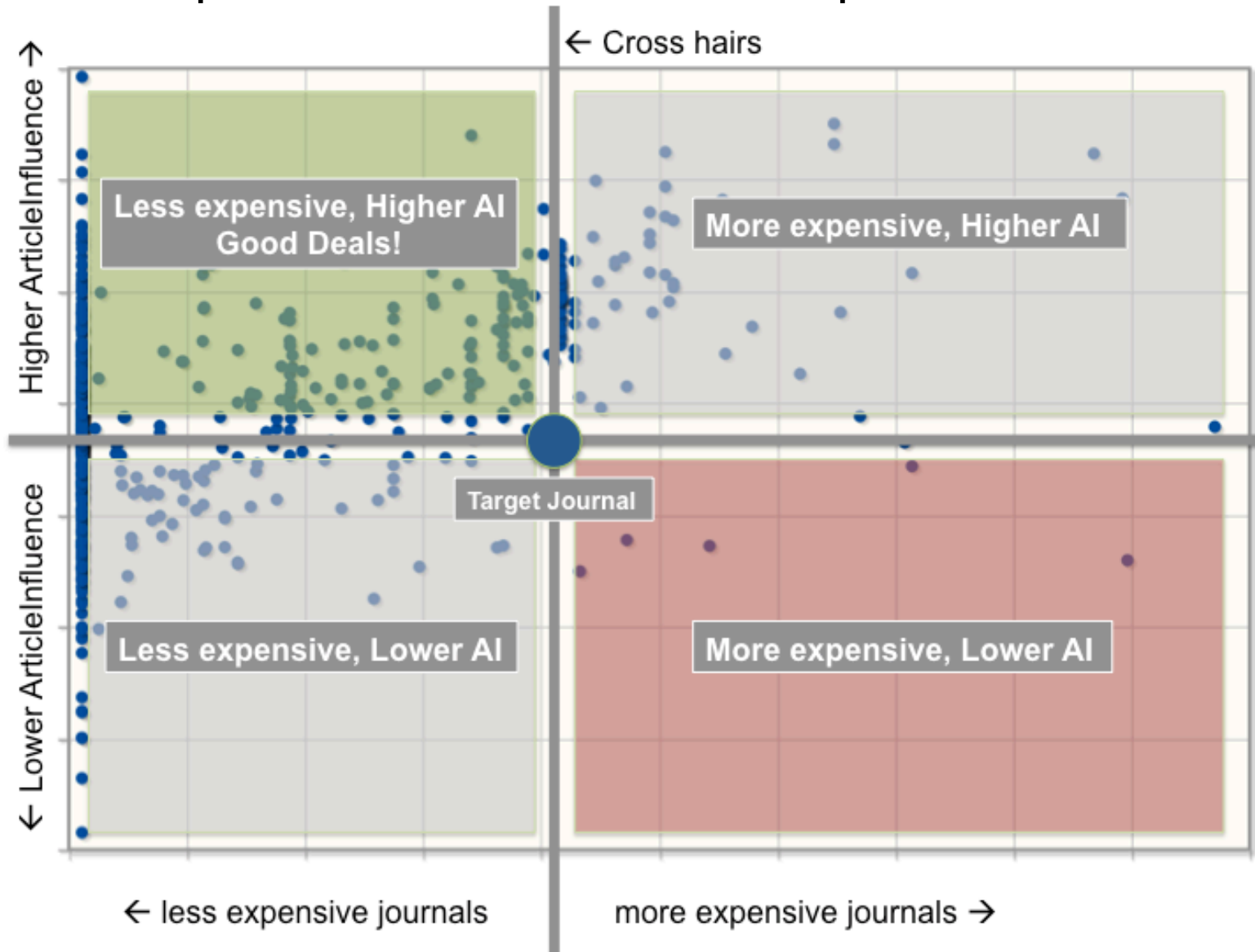
Image Source: <http://www.library.tudelft.nl/en/support/publishing-advice/open-access/>

**'Academic' publishers and titles
identified as predatory, 2011-16**



'Academic' publishers and titles identified as predatory, 2011-16

Open Market of Open Access



Maladies of Science

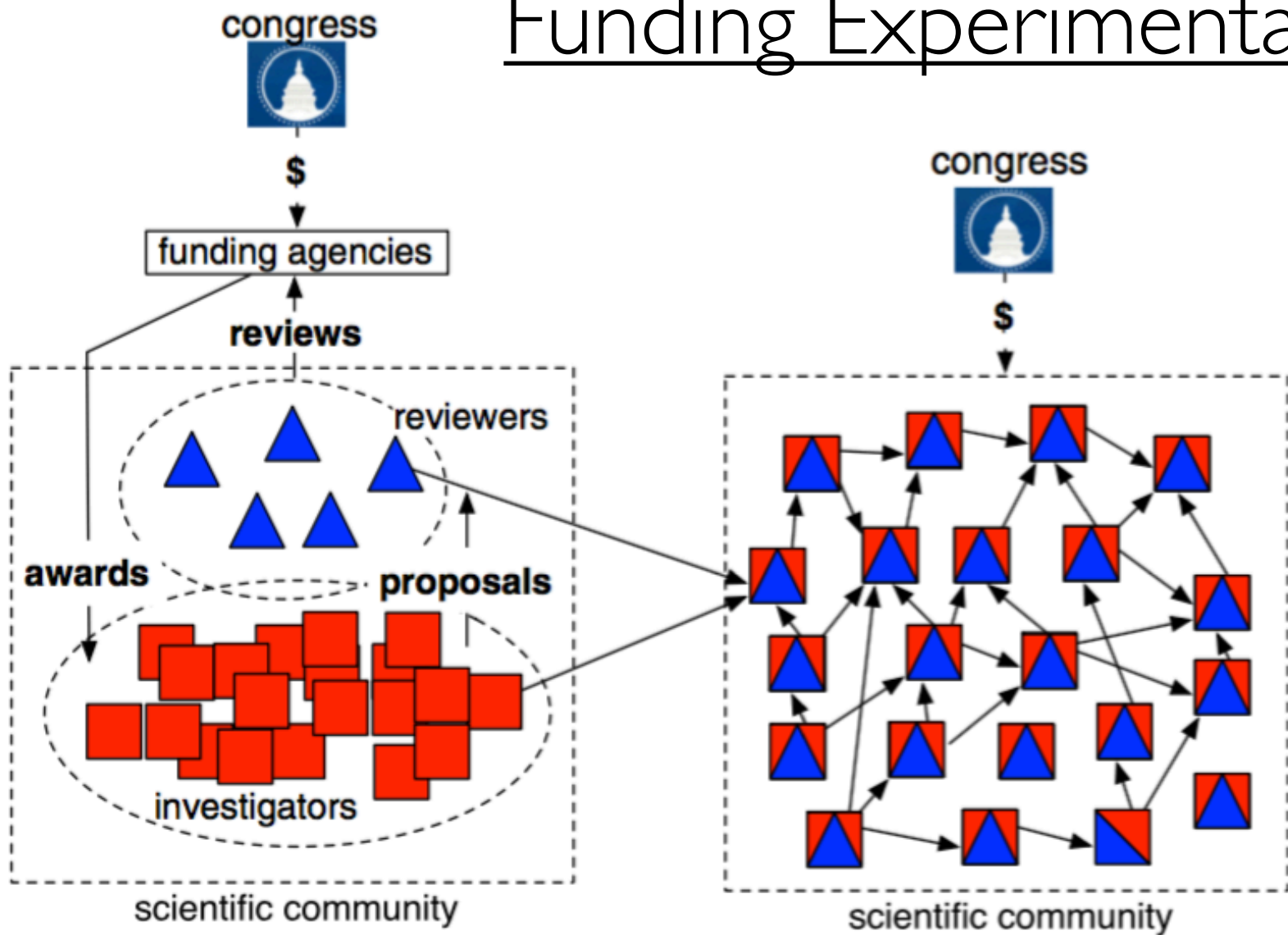
Process

Incentives

Infrastructure

Funding

Funding Experimentation



What is my impact on science?

\$7,933,670,366





22,756 awards

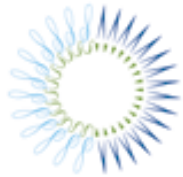
17,849 researchers

344,917 papers

8,174,533 citations

23.7 citations/paper

2006 - 2015



THE
PEW
CHARITABLE TRUSTS

Scholars Program
in the Biomedical Sciences

Science

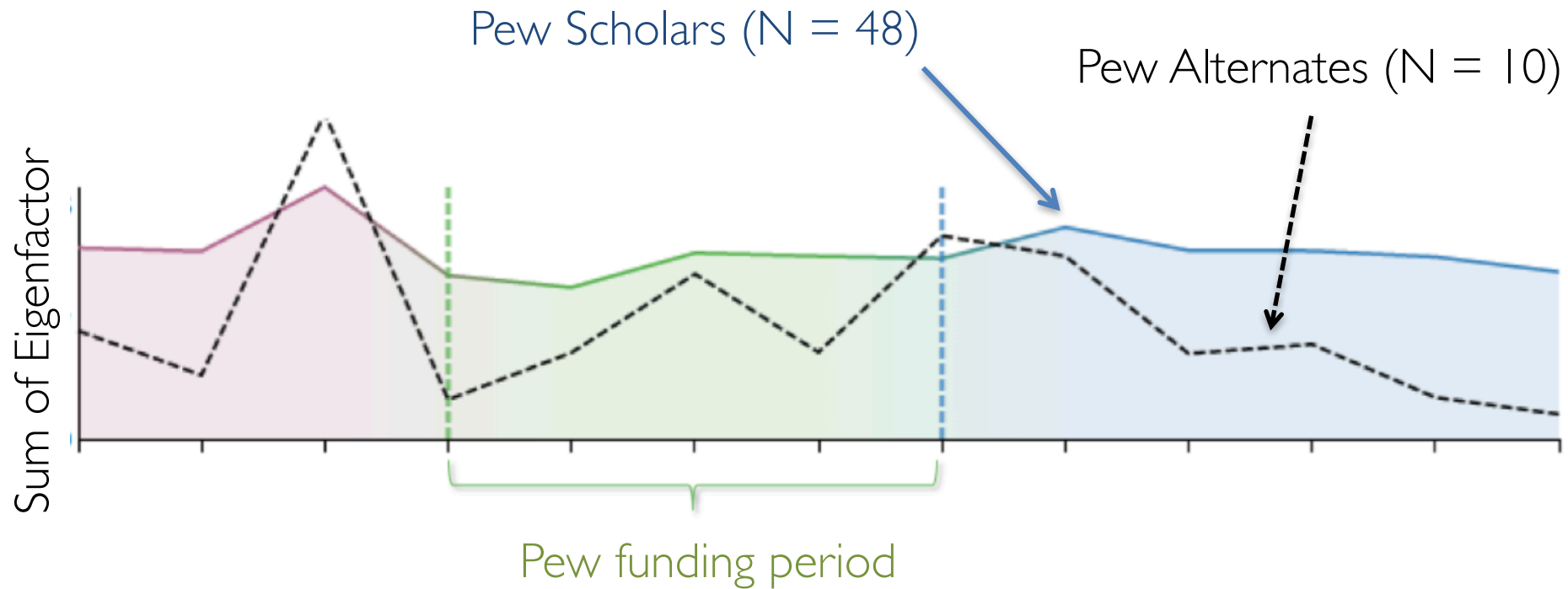
~ 37 citations/paper

median citations = 11

~ 5 citations/paper

median citations = 0

Comparing Alternates



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

How do we *map* the evolution of
scientific disciplines?

The map equation

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

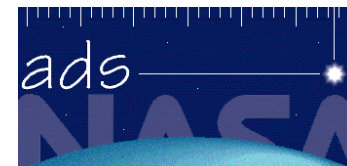
Data

Compressing \longleftrightarrow Finding patterns

Minimum description length (MDL) statistics.



The Scholarly Graph



THOMSON REUTERS

PatentVector™



PNAS





The Scholarly Graph



Tens of millions articles, patents, books

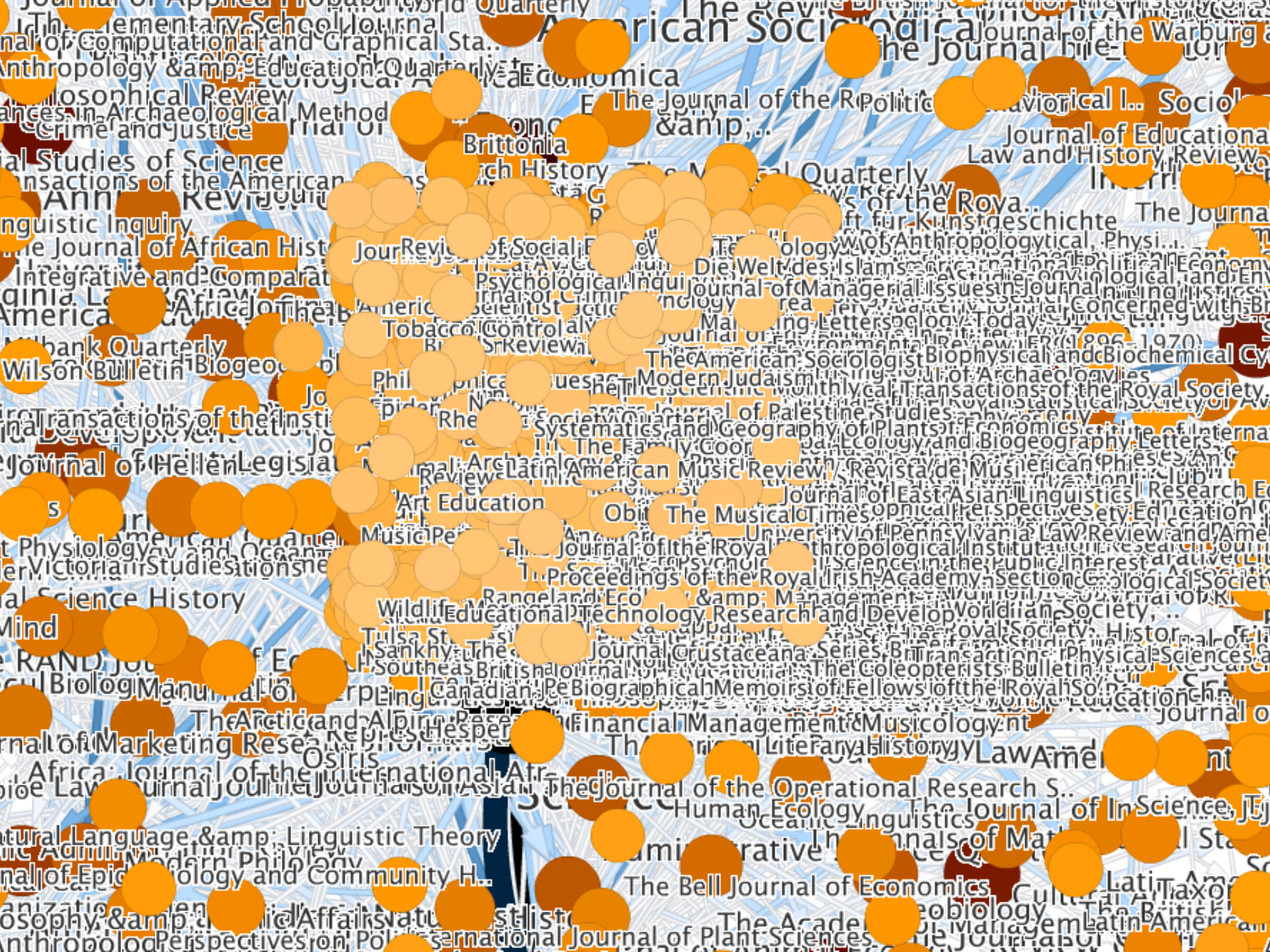


Billions of citation links

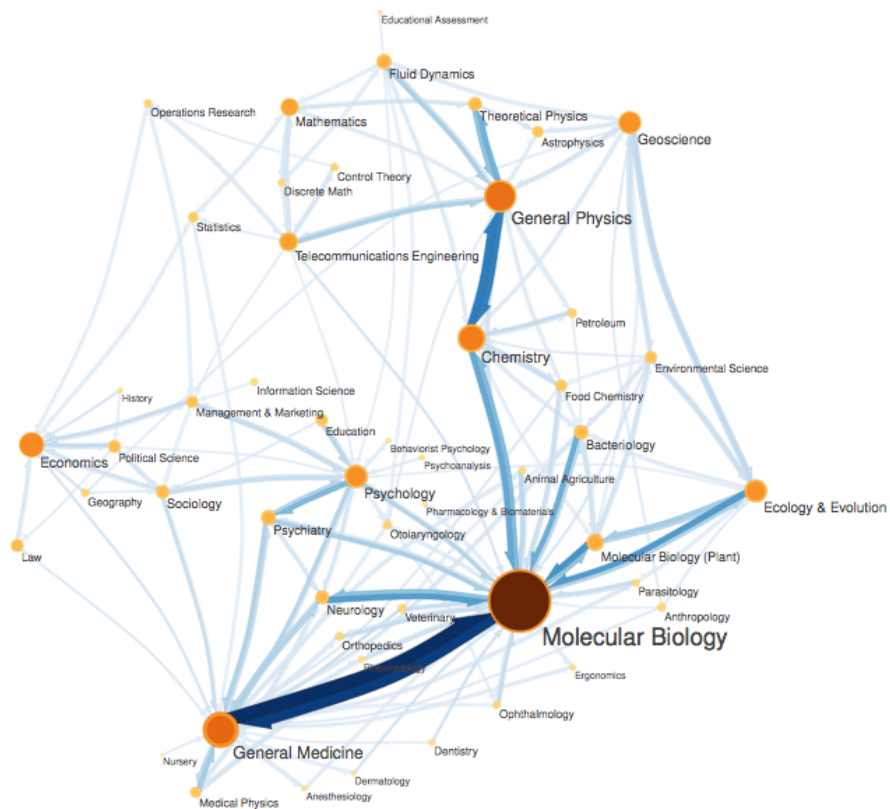


Years: 1600 - 2016

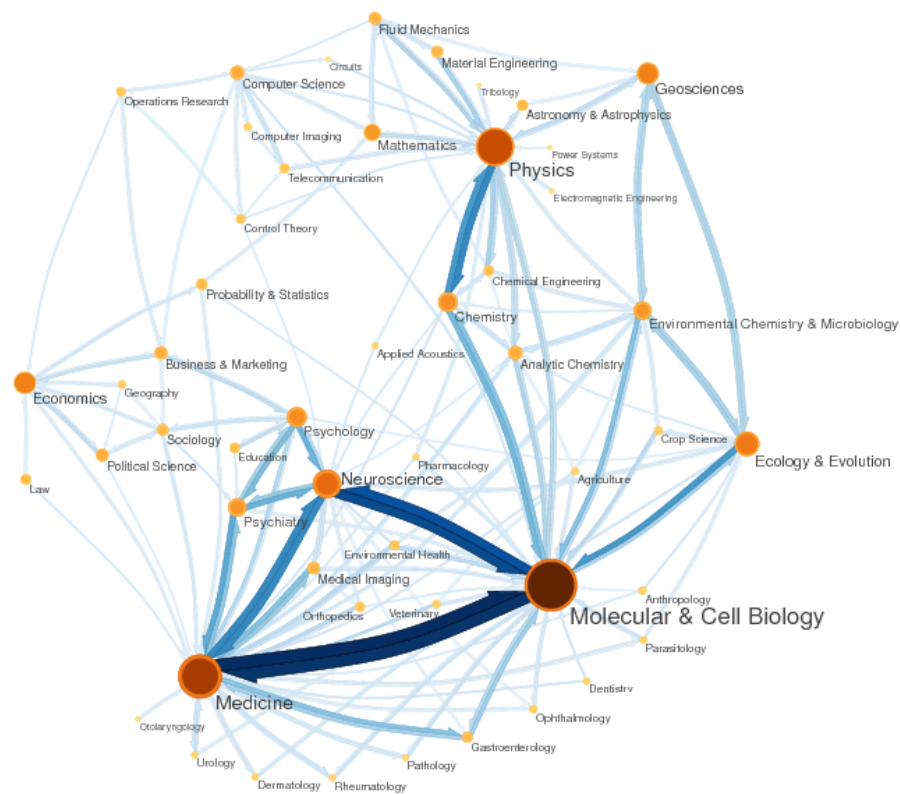




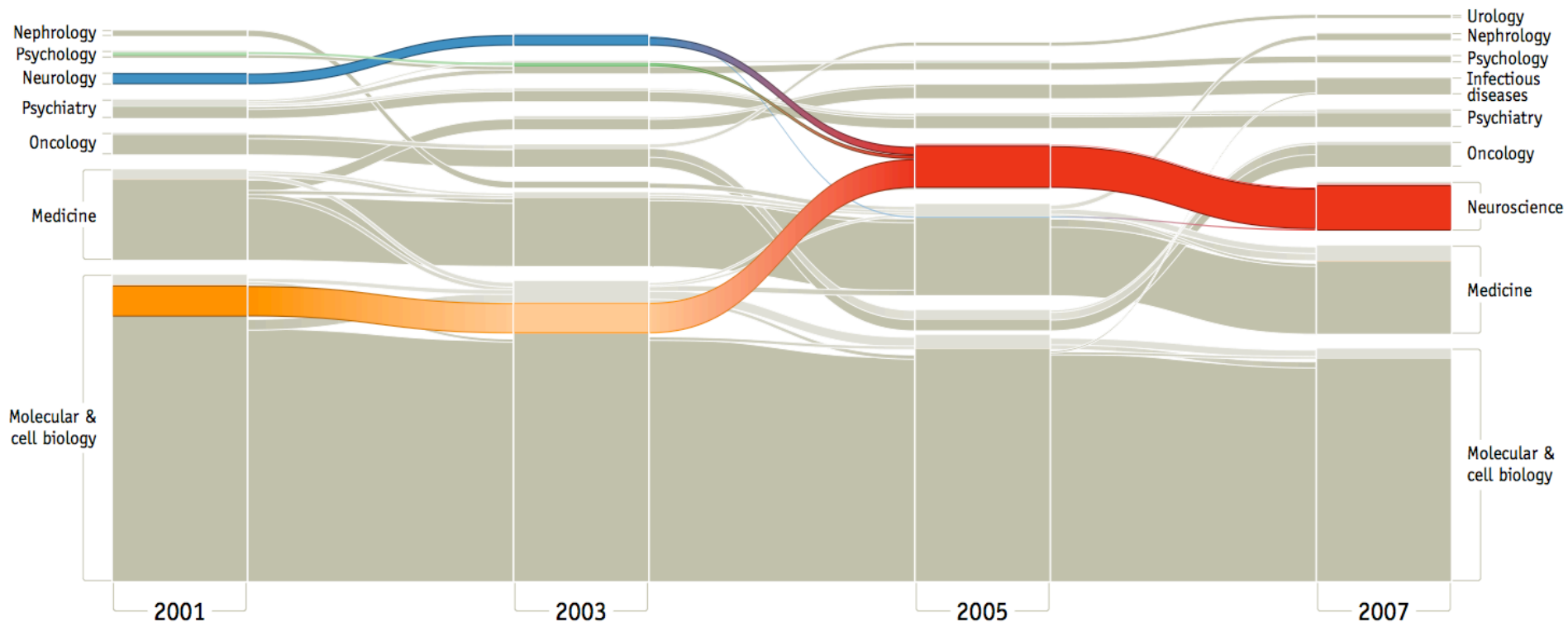
1995



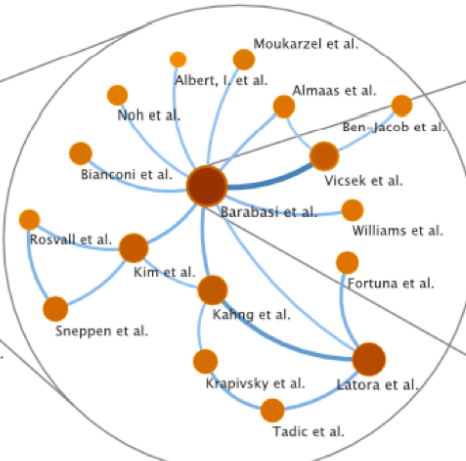
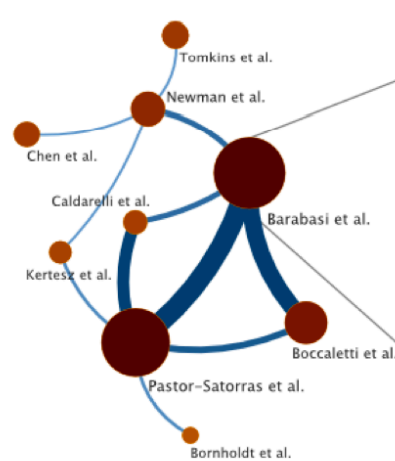
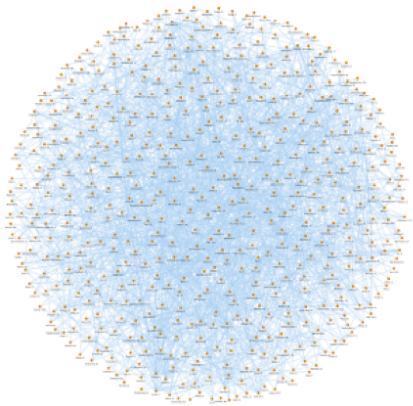
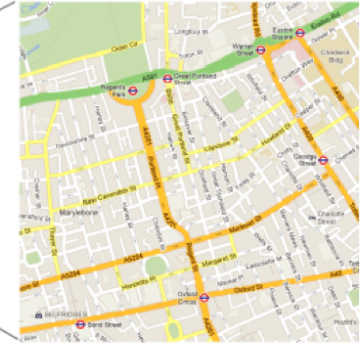
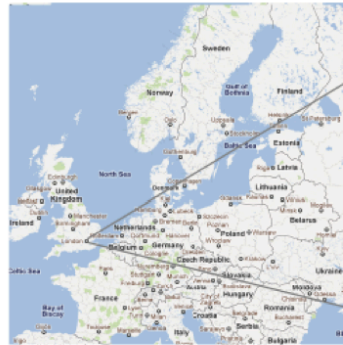
2004



The Emergence of Neuroscience



Zoomability



Current data technologies give you one level.

I want to give you the ability to zoom.

Maladies of Science

Process

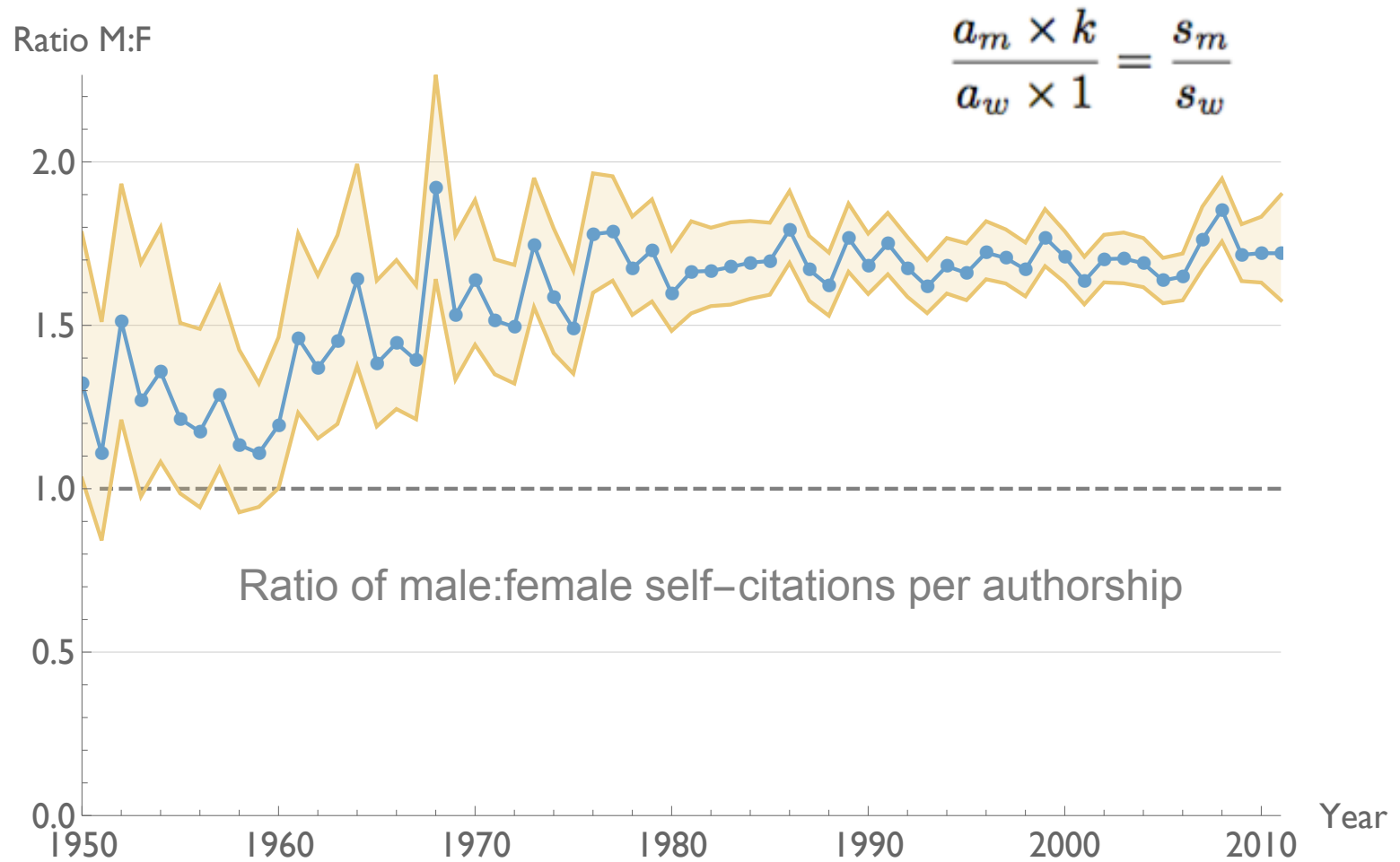
Incentives

Infrastructure

Funding

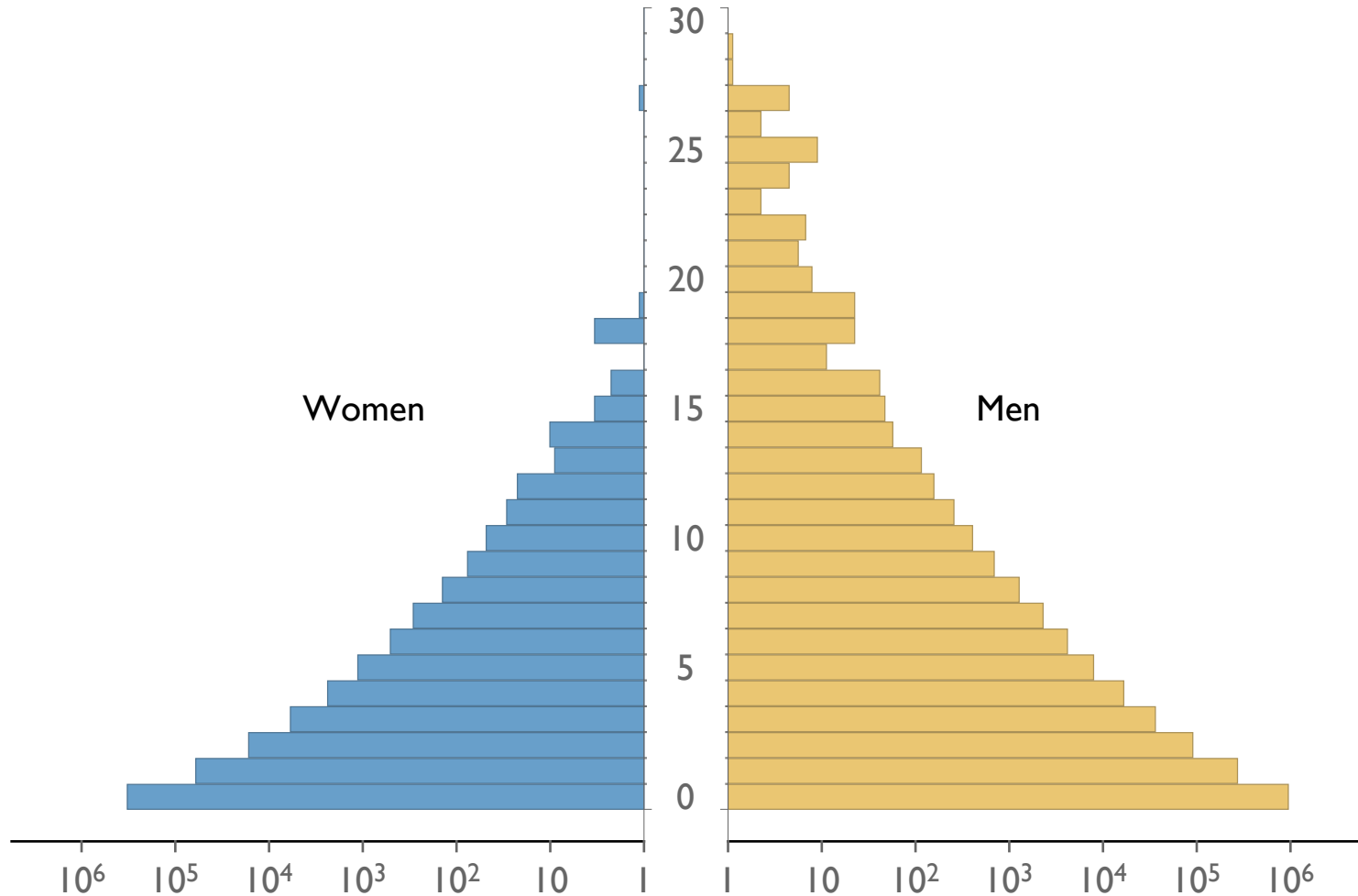
Bias

Self-citation over time



Excessive Self Citation

Number of authorships with n self-citations



Summarizing the *Maladies* of Science

Process

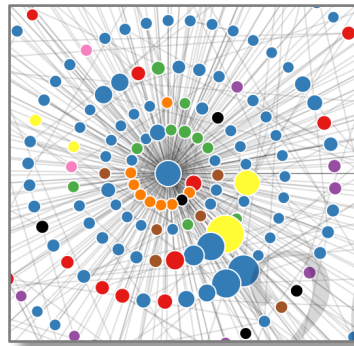
Incentives

Infrastructure

Funding

Bias

Science of Science



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