

# The MESUR project: studying science from usage data

## Implications for scholarly impact metrics

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# Outline

## 1 Introduction

- Problem statement
- Usage data

## 2 MESUR

- MESUR overview
- Creating MESUR's reference data set

## 3 Mapping Science

## 4 Metrics Survey

## 5 Services

## 6 Discussion

- Overview
- Future Research
- Relevant papers

# Why study science?

Tremendous importance

Enormous amount of resources and people involved:

**Allocation of resources:** funding agencies, policy makers, the public, corporations, the scientists themselves

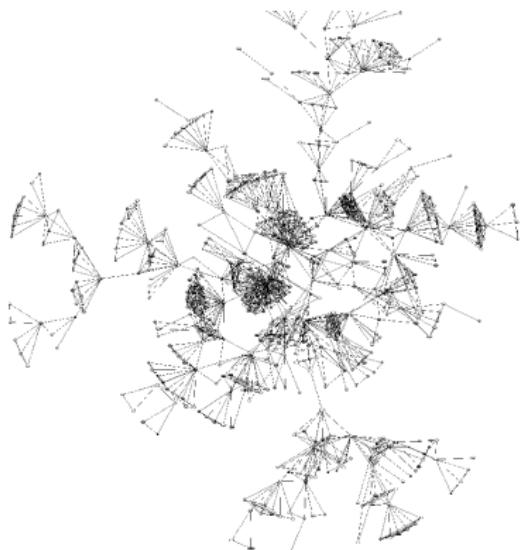
**Social systems research:** Ability to learn about general properties of similar social systems



Unrelated picture of my daughter

# Science as a social system

- Actors: scientists, stakeholders
- Relations (often focused on exchange of knowledge)
  - Informal interactions: meetings, workshops, conversations, messages, etc
  - Formal: affiliations, collaborations, projects, publications
- Artifacts: articles, journals, reports, data, software



Xiaoming Liu, Johan Bollen, Michael L. Nelson, and Herbert Van de Sompel. Co-authorship networks in the digital library research community. *Information Processing and Management*, 41(6):14621480, December 2005

# The measure of science in the print-era

Gold standard of measuring science: citation data

Extracted from “publication” data, so we know it’s good (right?)



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## ■ Consequence of print economy:

- Print-era: printed material (on “paper”! In “libraries”)
- Acknowledgement of influence: citations
- More citations, more influence
- Citation data: measurement of flow of influence and activity



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- Entire citation-based industry:

- End-user services
- Bibliometrics



# Two problems with present citation-based approach

Data (1) and metrics (2)

## Issues:

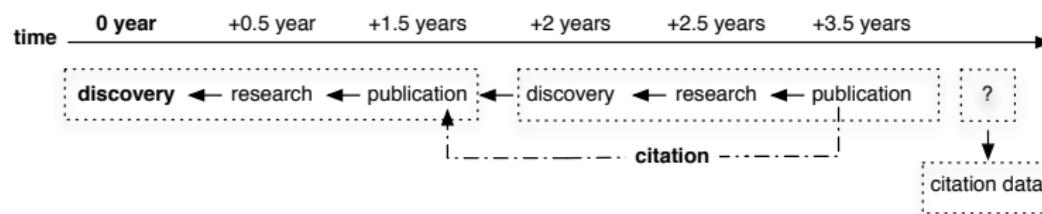
- 1 Inherent features of citation data due to its origins in published material
- 2 Existing metrics fail to acknowledge (1), and ignore network properties of science as a social system.

# Citation data

## Delays and partiality

### Citation problem 1: Data

Citation data is late and partial indicator of scientific activity



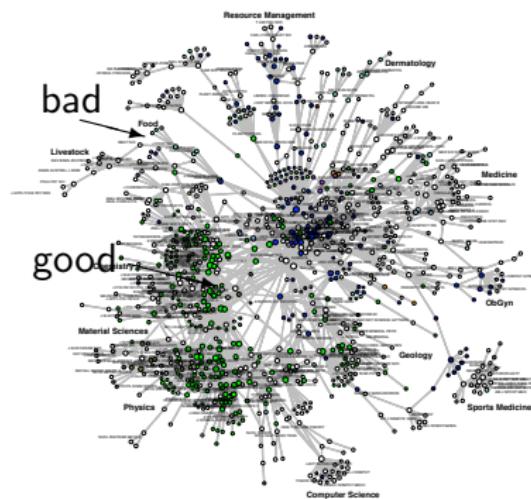
- Publication delays
- $\text{Publication}_t \rightarrow \text{publication}_{t-1} \rightarrow \text{citation DB}$
- Domain-dependent practices
- Community: Publishing authors only

# Citation data

## Metrics vs. networks

### Citation problem 2: networks

#### Ignoring network properties of science



Measuring impact, influence, prestige from citation data:

- More citations is better than less citations
- In citation network, central position is better than outskirts.

# Citation problem no. 2

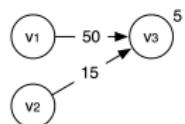
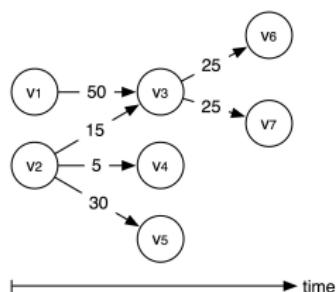
Metrics vs. networks

This approach is epitomized in Thomson-Reuters Impact Factor:

# Citation problem no. 2

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# Citation problem no. 2

Metrics vs. networks

This approach is epitomized in Thomson-Reuters Impact Factor:

Standard citation graph representation

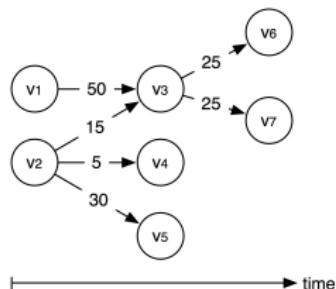
$$G = (V, E, W)$$

$$E \subseteq V^2$$

$$W : E \rightarrow \mathbb{N}^+$$

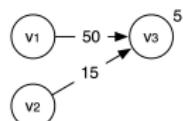
However:

$$\forall (n_i, n_j) \in E : \text{pubtime}(n_i) > \text{pubtime}(n_j)$$



Impact Factor: normalized in-degree

$$\text{IF}_j = \frac{\sum_i w_{ij}}{N_j}$$

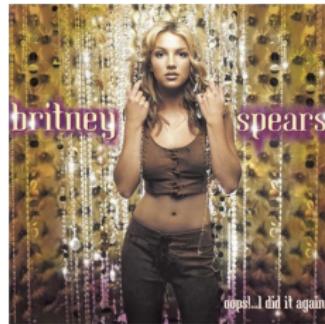


# Citation problem no. 2

Between Big Star and Britney Spears

Is more always better?

Why context matters

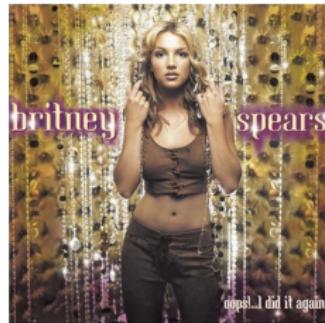


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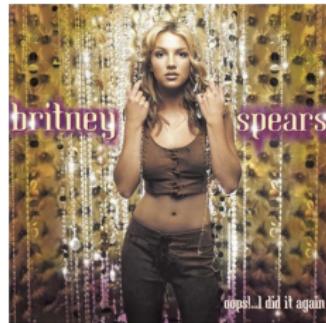
- Sold 50M records
- Influenced who?
- High popularity, low prestige.

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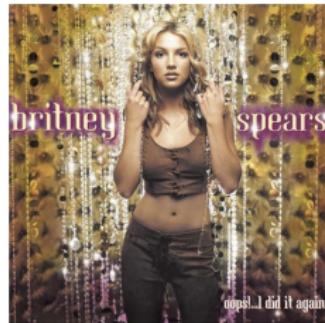
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- Sold 50k records
- Influenced REM, B52s, Teenage Fanclub, etc.
- Low popularity, high influence

# Citation problem no. 2

Between Big Star and Britney Spears

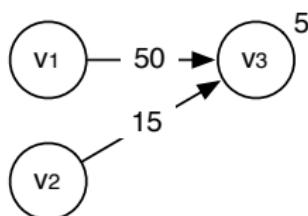
Silly? That's how we evaluate scientific impact right now!

The Impact Factor and lots of other citation-based metrics are presently being used to assess the quality of publications, journals, authors, institutions, and even entire countries by proxy.

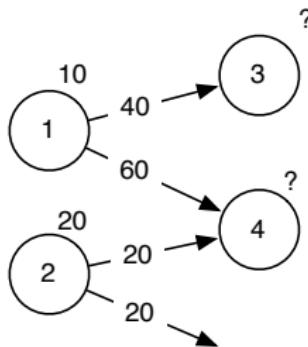
**Note:** Lots of developments on better metrics (cf. Eigenfactor: Bergstrom and Rosvall), but still reliance on citation data.

# New developments

## PageRank for citation graphs



Impact Factor:  $IF_j = \frac{\sum_i w_{ij}}{N_j}$ :  
normalized citation count  
**origin of citation disregarded**



A different route:

- $IF(v_i) \simeq \lambda \sum_j IF_j$
- $IF(v_i) \simeq \lambda \sum_j IF_j \times \frac{1}{O(v_j)}$
- $PR(v_i) \simeq \lambda \sum_j PR(v_j) \times \frac{1}{O(v_j)}$
- $PR(v_i) \simeq \frac{(1-\lambda)}{N} + \lambda \sum_j PR(v_j) \times \frac{1}{O(v_j)}$
- $PR_w(v_i) = \frac{(1-\lambda)}{N} + \lambda \sum_j PR_w(v_j) \times w(v_j, v_i)$

# New developments, contd

## PageRank for citation graphs

ISI IF			PR <sub>w</sub>	
rank	value	Journal	value ( $\times 10^3$ )	Journal
1	52.28	ANNU REV IMMUNOL	16.78	NATURE
2	37.65	ANNU REV BIOCHEM	16.39	J BIOL CHEM
3	36.83	PHYSIOL REV	16.38	SCIENCE
4	35.04	NAT REV MOL CELL BIO	14.49	P NATL ACAD SCI USA
5	34.83	NEW ENGL J MED	8.41	PHYS REV LETT
6	30.98	NATURE	5.76	CELL
7	30.55	NAT MED	5.70	NEW ENGL J MED
8	29.78	SCIENCE	4.67	J AM CHEM SOC
9	28.18	NAT IMMUNOL	4.46	J IMMUNOL
10	28.17	REV MOD PHYS	4.28	APPL PHYS LETT

Table: The highest ranking journals according to ISI IF and Weighted PageRank (JCR2003)

# Two problems with present citation-based approach

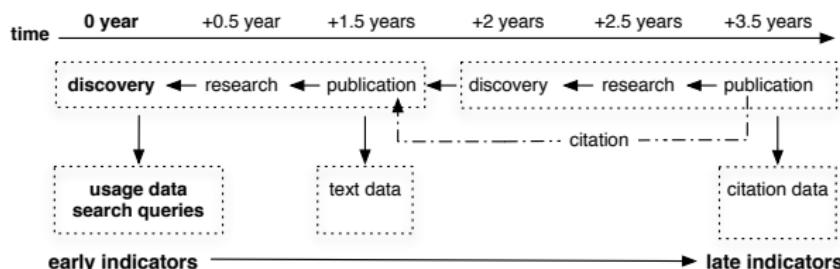
Data (1) and metrics (2)

## Issues:

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- 2 Existing metrics fail to acknowledge (1), and ignore network properties of science as a social system.

# Post-print, online era, aka known as “today”: usage data?

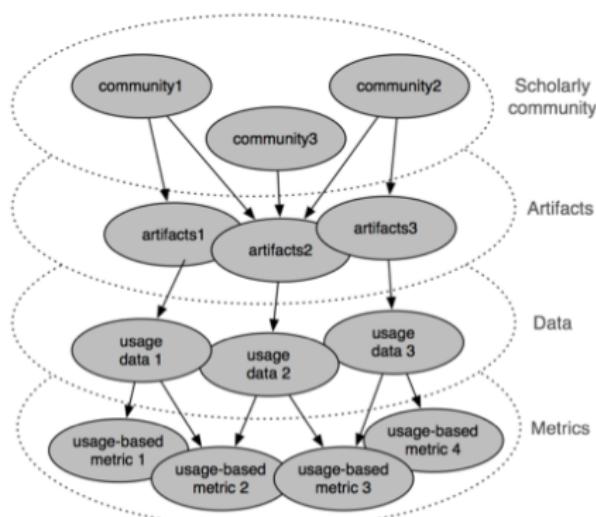
Most use of scholarly resources is now mediated by online services



- Recorded by nearly every scholarly service
- Captures early phases of scientific activity
- Includes a wide variety of resource types
- Activities of larger scientific communities, e.g. not limited only to authors
- Scale: cf. Elsevier announced 1B downloads in 2006 vs. 650M citation in WoS.

# Issues with usage data

Lots of interest in leveraging usage data for models of science, tracking scientific activity, metrics, etc. Cf. COUNTER, Ex Libris bX, and many more.



However many different issues:

- Silo: recorded for *particular service* for a *particular user community*
- Lack of standards: recorded in different manner, for different systems
- Lack of research: how leverage usage data for scientific tracking, modeling, metrics, etc.

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# MESUR project: survey the potential of usage data at very large-scale

Studying scientific activity



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## Studying scientific activity

- Can we model and study patterns of scientific activity from large-scale, representative usage data?



# MESUR project: survey the potential of usage data at very large-scale

## Studying scientific activity

- Can we model and study patterns of scientific activity from large-scale, representative usage data?
- What can we learn about impact and structure from patterns of scientific activity from usage data?



# MESUR history



The Andrew W. Mellon Foundation



# MESUR history



- 2006-2008: Andrew W. Mellon Foundation

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- 2009-: Indiana University, School of Informatics and Computing
- 2009-2013: National Science Foundation

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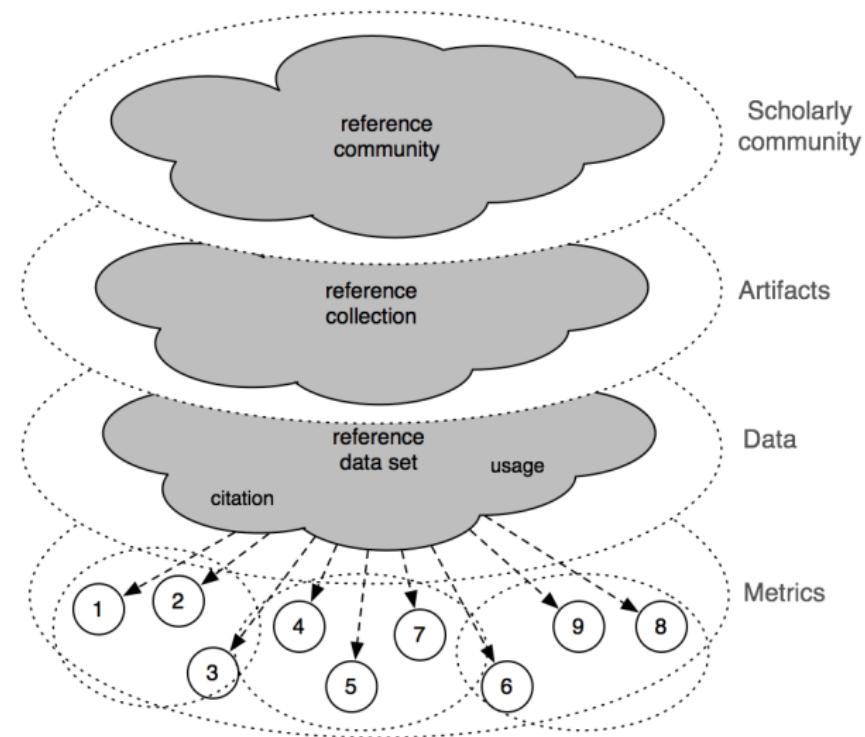
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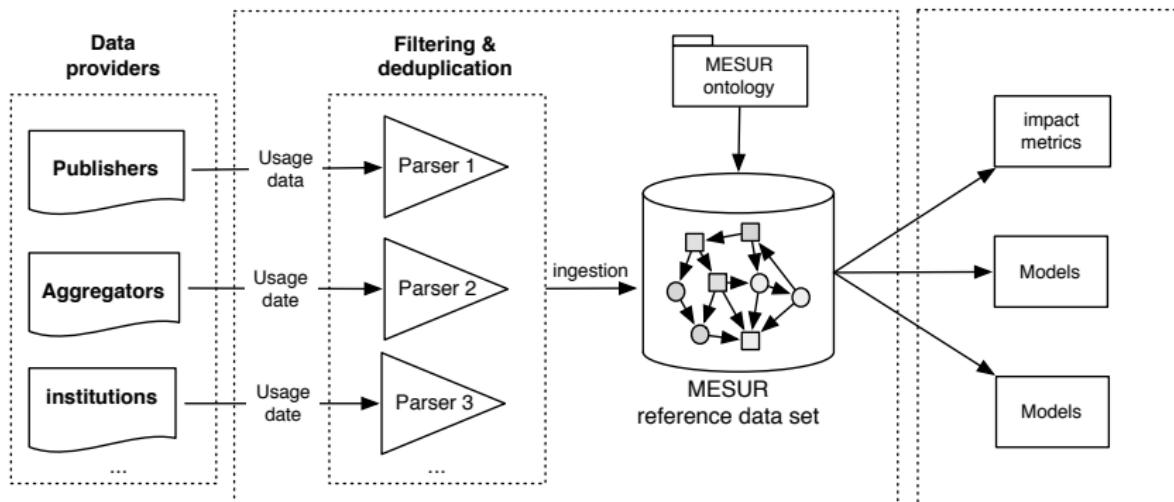
 INDIANA UNIVERSITY  
BLOOMINGTON

- 2006-2008: Andrew W. Mellon Foundation
- 2008-2009: Los Alamos National Laboratory
- 2009-: Indiana University, School of Informatics and Computing
- 2009-2013: National Science Foundation
- **Team: PI, co-PI, 2 scientists, 2 full-time developers, and 1 PhD student.**

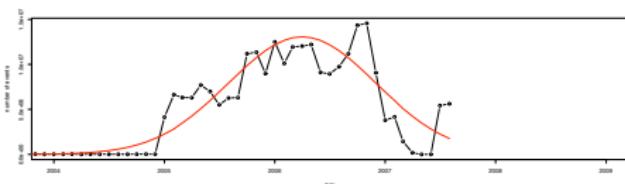
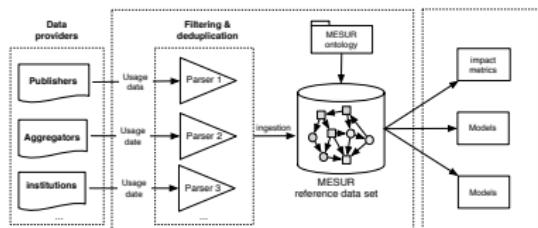
# MESUR objective



# MESUR dataflow



# Creating MESUR's data set



**Providers** 2006-2010: BMC, Blackwell, UC, CSU (23), EBSCO, ELSEVIER, EMERALD, INGENTA, JSTOR, LANL, MIMAS/ZETOC, THOMSON, UPENN (9), UTEXAS

**events** **1,000,000,000 usage events**

**citations** +500,000,000 citations,

**articles and serials** +50M articles, +-100,000 serials

# Data requirements

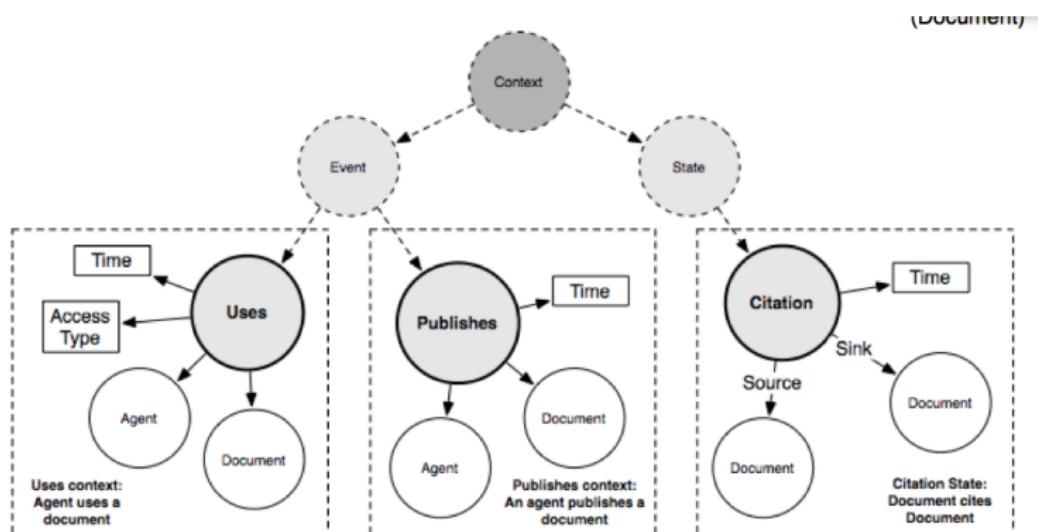
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10..1186/1471-2164-7-81.BMC Genomics.JUSTPUBLISHEDBSTRACT,01/05/2006 00:02:02,B0153,F033B664F540B88BE4A380650FC60F,,C304372C	
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## Minimum requirements

- Separate user requests
- Date-time stamp down to second
- Document identifier or sufficient metadata to de-duplicate
- Session ID (or anonymized user ID)
- Request type identifier

<http://tweety.lanl.gov/public/schemas/2007-01/mesur.owl>

# MESUR's OWL/RDF ontology



1

<sup>1</sup> Rodriguez, Bollen & Van de Sompel. A Practical Ontology for the Large-Scale Modeling of Scholarly Artifacts and their Usage. JCDL07 - Based on OntologyX work.

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# Reference data set

## Subsetting

Domain	Usage	UC Degrees	JCR
Natural Science	37%	39%	92.8%
Social Sciences	45%	46%	7.2%
Humanities	14%	15%	

Source: <http://www.ucop.edu/ucophome/uwnews/stat/statsum/fall2007/statsumm2007.pdf> (table 9)

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Scale: 346,312,045 usage events, 97,532 serials (many of which not journals)

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# Article clickstreams

usage data log:  $U = \{u_1, u_2, \dots, u_n\}$

$u = \{s, t, a\}$

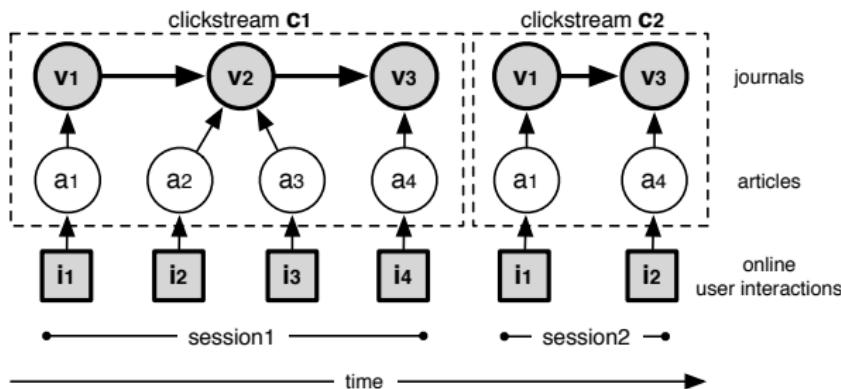
$s$  = session identifier ,  $t$  = a date-time and  $a$  = article

$f \in F$  = clickstreams extracted from  $U$

$f \subset U$ , where

$f = (\forall u \in U, \exists s : s(u) \wedge t(u_i) < t(u_{i+1}))$

$s(u)$  and  $t(u)$ : session identifier and date-time of interaction  $u$ .



# Journal Clickstreams

article clickstream:  $f_a = (a_1, a_2, \dots, a_k)$

journal clickstream:  $f_v = (v_1, v_2, \dots, v_k)$

We observe:  $N(v_i, v_j)$

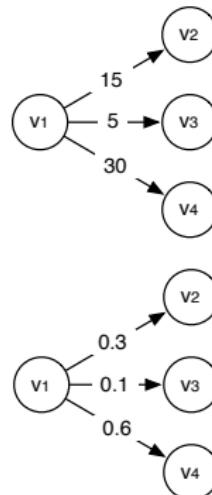
for all pairs  $(v_i, v_j)$  in which  $j = i + 1$

From which follows:

$$P(v_i, v_j) = \frac{N(v_i, v_j)}{\sum_j N(v_i, v_j)}$$

and

$M$  whose entries  $m_{i,j} = P(v_i, v_j)$ .



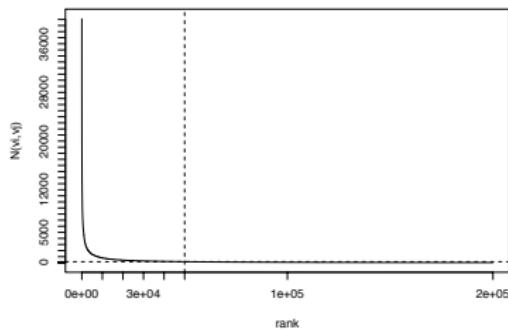
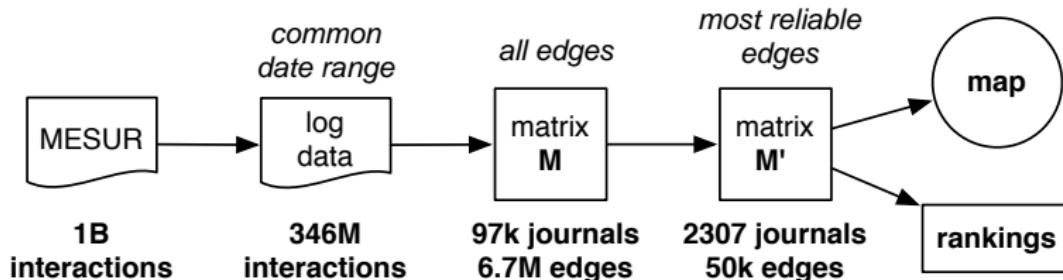
# Examples of prominent connections

$v_i$	$v_j$	$p(v_i, v_j)$	$N(v_i, v_j)$	$N(v_i)$
American Journal of International Law	International Organization International Affairs International and Comparative Law Quarterly Foreign Policy American Political Science Association	0.0207 0.0184 0.0171 0.0167 0.0140	9,292 8,254 7,654 7,500 6,291	448,034
Journal of Educational Sociology	American Journal of Sociology Journal of Higher Education Journal of Negro Education American Sociological Review Social Forces	0.0334 0.0303 0.0286 0.0276 0.0249	2,790 2,529 2,389 2,303 2,076	83,419
Surface Science	Physical Review B Applied Surface Science Physical Review Letters Journal of Chemical Physics Applied Physics Letters	0.0704 0.0341 0.0339 0.0333 0.0327	2,555 1,239 1,230 1,207 1,188	36,282
Journal of Organic Chemistry	Journal of the American Chemical Society Tetrahedron Letters Tetrahedron Organic Letters Angewandte Chemie	0.0873 0.0865 0.0602 0.0532 0.0305	4,141 4,105 2,857 2,526 1,448	47,439
Ecological Applications	Ecology Conservation Biology Bioscience Annual Review of Ecology and Systematics Clinical and Experimental Allergy	0.0965 0.0524 0.0215 0.0215 0.0191	13,659 7,408 3,043 3,043 2,699	141,481
Annals of Mathematics	American Journal of Mathematics American Mathematical Monthly PNAS Econometrica Mathematics Magazine	0.0705 0.0579 0.0156 0.0082 0.0077	5,392 4,432 1,195 624 587	76,526

# Network parameters

Network matrix

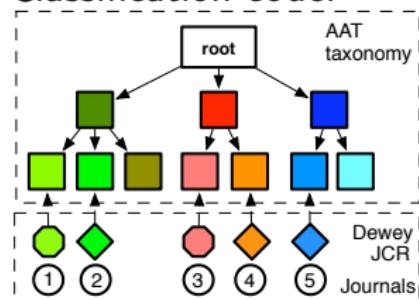
Parameter	$M$	$M'$
Journals	97,532	2,307
Edges	6,783,552	50,000
Matrix density	0.071%	0.939%
Strongly Connected Components (SCC)	16,474	236
Journals in SCC	80,934	1,944
Average journal clustering coefficient (SCC)	0.285	0.514
Diameter of largest SCC	37	14

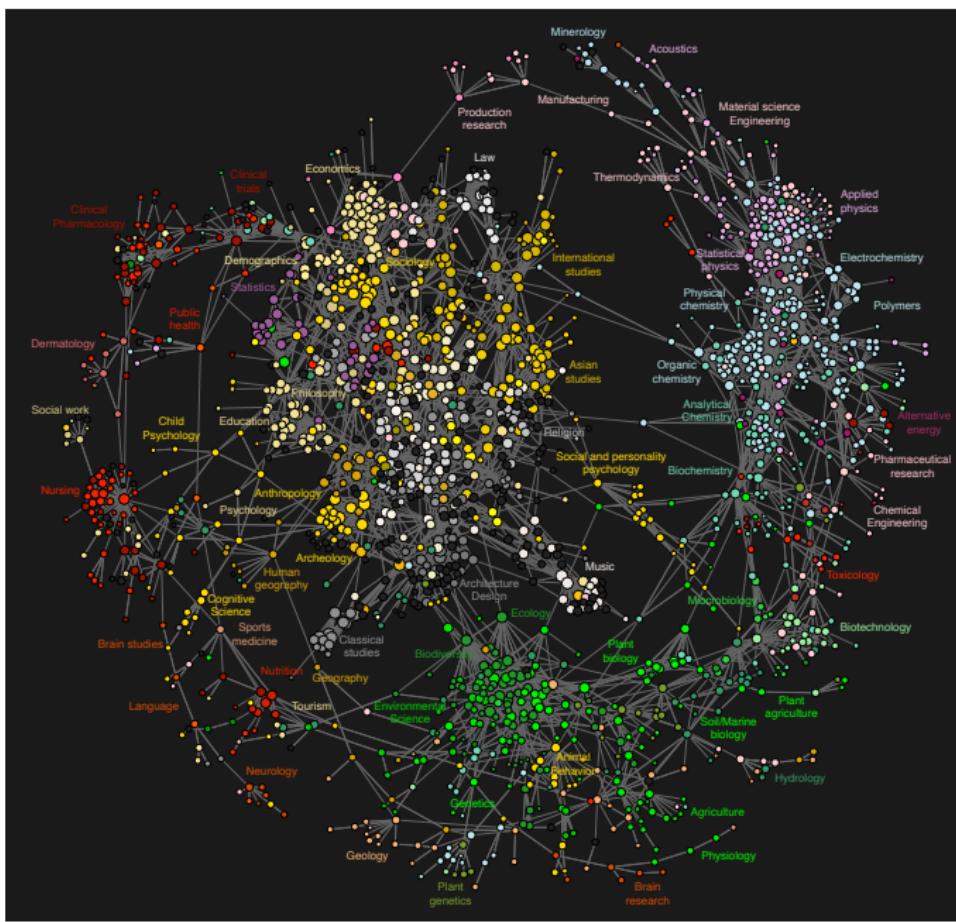


Fruchterman (1991) Graph Drawing by Force-directed

Placement. Software, 21(11):1129-1164

### Classification code:

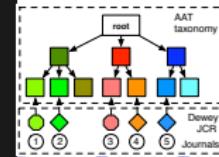
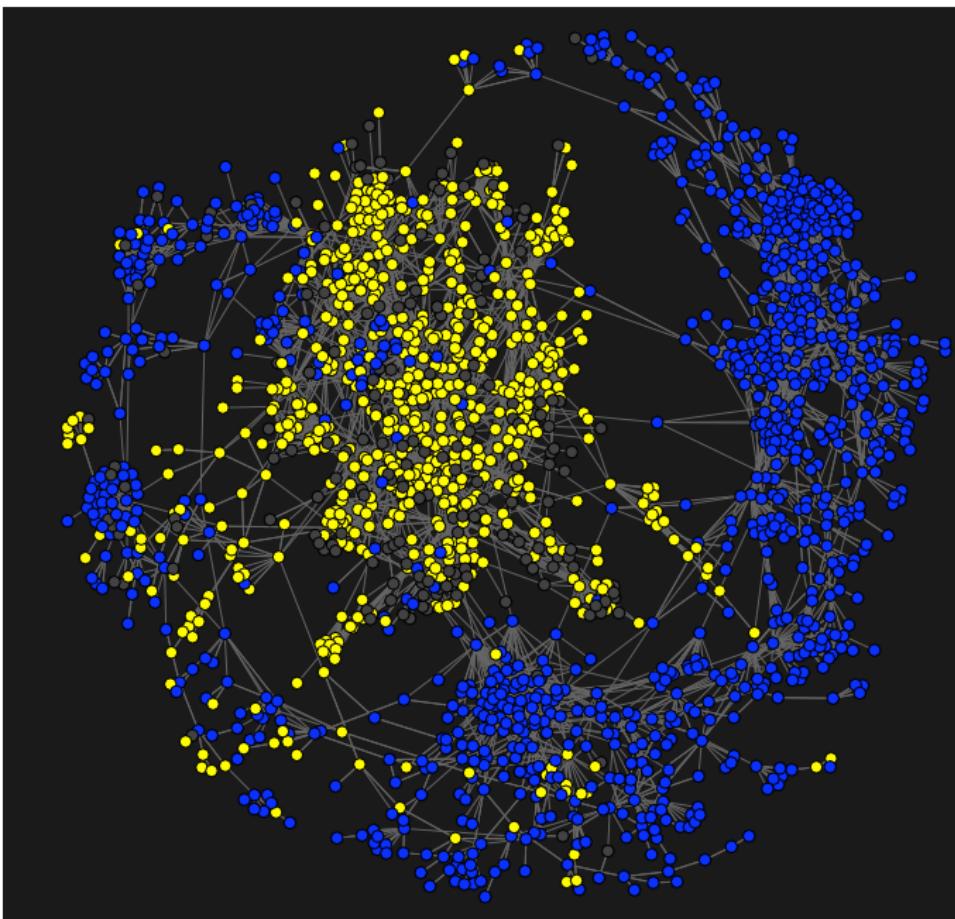




# The Arts and Architecture Thesaurus (Getty Research Center)

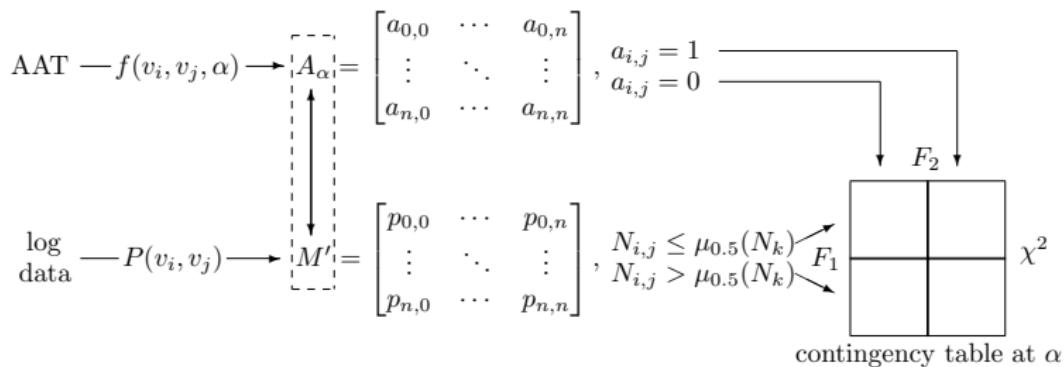
**Table: Distance from AAT root ( $\alpha$ ) and number of classifications  $N_c$  at that level. Each  $\alpha$  produces a finer-grained separation of scientific disciplines.**

Distance ( $\alpha$ )	$N_c$	Example classifications
1	4	Natural sciences, social sciences, humanities, ...
2	8	Biology, chemistry, physics, ...
3	31	Classics, communication, engineering, ...
4	195	Allergy, anesthesiology, applied linguistics, ...



# Cross-validating the clickstream map

$$f(v_i, v_j, \alpha) = \begin{cases} 1 & C_\alpha(v_i) = C_\alpha(v_j) \\ 0 & C_\alpha(v_i) \neq C_\alpha(v_j) \end{cases}$$



$$\alpha = 1 : p < 0.0001$$

$$\alpha = 2 : p < 0.0001$$

$$\alpha = 3 : p < 0.0001$$

$$\alpha = 4 : p < 0.0001$$

# Betweenness centrality

$$C_b(v_k) = \sum_{i \neq j \neq k} \frac{\sigma_{i,j}(v_k)}{\sigma_{i,j}} \quad (1)$$

**Table: Ranking of journals from  $M'$  according to betweenness centrality.**

Rank	Journal	Top-level AAT classification
1	Science	Natural Sciences
2	Proceedings of the National Academy of Sciences	Natural Sciences
3	Environmental Health Perspectives	Natural Science
4	Chemosphere	Natural Sciences
5	Journal of Advanced Nursing	Natural Sciences
6	Nature	Natural Sciences
7	Ecology	Natural Sciences
8	Milbank Quarterly	Natural Sciences
9	Applied and Environmental Microbiology	Natural Sciences
10	Child Development	Social Sciences
11	Behavioral Ecology and Sociobiology	Social Sciences
12	Journal of Colloid and Information Science	Natural Sciences
13	American Anthropologist	Social Sciences
14	Journal of Biogeography	Natural Sciences
15	Materials Science and Technology	Natural Sciences

# PageRank

$$PR(v_i) = \frac{1 - \lambda}{N} + \lambda \sum_j \frac{PR(v_j)}{O(v_j)} \quad (2)$$

**Table: Ranking of journals from  $M'$  according to PageRank ( $\lambda = 0.85$ ).**

Rank	Journal	Top-level AAT classification
1	Applied Physics Letters	Natural Sciences
2	Journal of Advanced Nursing	Natural Sciences
3	Journal of the American Chemical Society	Natural Sciences
4	Ecology	Natural Sciences
5	Nature	Natural Sciences
6	Physical Review B	Natural Sciences
7	Journal of Applied Physics	Natural Sciences
8	American Economic Review	Social Sciences
9	American Historical Review	Social Sciences
10	Physical Review Letters	Natural Sciences
11	Science	Natural Sciences
12	Langmuir	Natural Sciences
13	Journal of Chemical Physics	Natural Sciences
14	American Anthropologist	Social Sciences
15	Annals of the American Academy of Political and Social Science	Social Science

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## 5 Services

## 6 Discussion

- Overview
- Future Research
- Relevant papers

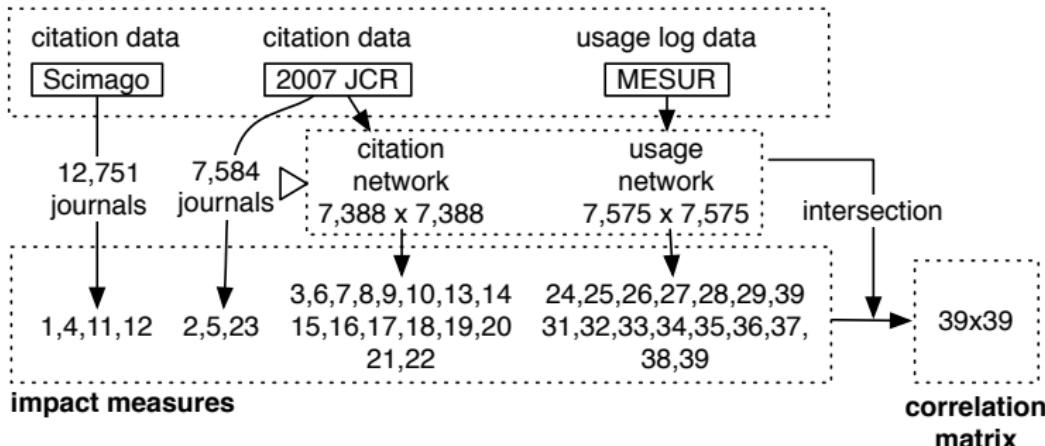
# Metrics

ID	Type	Measure	Source	Network parameters	PC1	PC2	$\bar{p}$
1	Citation	Scimago Journal Rank	Scimago/Scopus		-0.974	-8.296	0.556 *
2	Citation	Immediacy Index	JCR 2007		1.659	-7.046	0.508 *
3	Citation	Closeness Centrality	JCR 2007	Undirected, weighted	0.339	-6.284	0.565 *
4	Citation	Cites per doc	Scimago/Scopus		-1.311	-6.192	0.588 *
5	Citation	Journal Impact Factor	JCR 2007		-1.854	-5.937	0.592 *
6	Citation	Closeness centrality	JCR 2007	Undirected, unweighted	-1.388	-4.827	0.619
7	Citation	Out-degree centrality	JCR 2007	Directed, weighted	-3.191	-4.215	0.642
8	Citation	Out-degree centrality	JCR 2007	Directed, unweighted	-2.703	-4.015	0.640
9	Citation	Degree Centrality	JCR 2007	Undirected, weighted	-4.850	-2.834	0.690
10	Citation	Degree Centrality	JCR 2007	Undirected, unweighted	-4.398	-2.643	0.691
11	Citation	H-Index	Scimago/Scopus		-3.326	-2.003	0.681
12	Citation	Scimago Total cites	Scimago/Scopus		-4.926	-1.722	0.712
13	Citation	Journal Cite Probability	JCR 2007		-5.389	-1.647	0.710
14	Citation	In-degree centrality	JCR 2007	Directed, unweighted	-5.302	-1.429	0.717
15	Citation	In-degree centrality	JCR 2007	Directed, weighted	-5.380	-1.554	0.712
16	Citation	PageRank	JCR 2007	Directed, unweighted	-4.476	0.108	0.693
17	Citation	PageRank	JCR 2007	Undirected, unweighted	-4.929	0.731	0.726
18	Citation	PageRank	JCR 2007	Undirected, weighted	-4.160	0.864	0.696
19	Citation	PageRank	JCR 2007	Directed, weighted	-3.103	0.333	0.659
20	Citation	Y-factor	JCR 2007	Directed, weighted	-2.971	0.317	0.657
21	Citation	Betweenness centrality	JCR 2007	Undirected, weighted	-0.462	0.872	0.643
22	Citation	Betweenness centrality	JCR 2007	Undirected, unweighted	-0.474	1.609	0.642
23	Citation	Citation Half-Life	JCR 2007		/	/	0.037
24	Usage	Closeness centrality	MESUR 2007	Undirected, weighted	3.130	2.683	0.703
25	Usage	Closeness centrality	MESUR 2007	Undirected, unweighted	3.100	3.899	0.731
26	Usage	Degree centrality	MESUR 2007	Undirected, unweighted	3.271	3.873	0.729
27	Usage	PageRank	MESUR 2007	Undirected, unweighted	3.327	4.192	0.728
28	Usage	PageRank	MESUR 2007	Directed, unweighted	3.463	4.336	0.727
29	Usage	In-degree centrality	MESUR 2007	Directed, unweighted	3.463	4.015	0.728
30	Usage	Out-degree centrality	MESUR 2007	Directed, unweighted	3.484	3.994	0.727
31	Usage	PageRank	MESUR 2007	Directed, weighted	3.780	4.217	0.710
32	Usage	PageRank	MESUR 2007	Undirected, weighted	3.813	4.223	0.710
33	Usage	Betweenness centrality	MESUR 2007	Undirected, unweighted	3.988	4.271	0.699
34	Usage	Betweenness centrality	MESUR 2007	Undirected, weighted	3.957	3.698	0.693
35	Usage	Degree centrality	MESUR 2007	Undirected, weighted	5.293	3.528	0.683
36	Usage	Out-degree centrality	MESUR 2007	Directed, weighted	5.302	3.518	0.683
37	Usage	In-degree centrality	MESUR 2007	Directed, weighted	5.286	3.531	0.683
38	Usage	Journal Use Probability	MESUR 2007		8.914	1.833	0.593
39	Usage	Usage Impact Factor	MESUR 2007		/	/	0.279



$$R_{10 \times 10} = \begin{pmatrix} 1.00 & 0.71 & 0.77 & 0.52 & 0.79 & 0.55 & 0.69 & 0.63 & 0.60 & 0.18 \\ 0.71 & 0.99 & 0.52 & 0.69 & 0.79 & 0.85 & 0.49 & 0.44 & 0.49 & 0.22 \\ 0.77 & 0.52 & 1.00 & 0.62 & 0.63 & 0.39 & 0.70 & 0.73 & 0.68 & 0.20 \\ 0.52 & 0.69 & 0.62 & 1.00 & 0.68 & 0.78 & 0.49 & 0.56 & 0.65 & 0.06 \\ 0.79 & 0.79 & 0.63 & 0.68 & 1.00 & 0.82 & 0.66 & 0.62 & 0.66 & 0.15 \\ 0.55 & 0.85 & 0.39 & 0.78 & 0.82 & 1.00 & 0.40 & 0.40 & 0.50 & 0.13 \\ 0.69 & 0.49 & 0.70 & 0.49 & 0.66 & 0.40 & 1.00 & 0.89 & 0.85 & 0.53 \\ 0.63 & 0.44 & 0.73 & 0.56 & 0.62 & 0.40 & 0.89 & 1.00 & 0.97 & 0.45 \\ 0.60 & 0.49 & 0.68 & 0.65 & 0.66 & 0.50 & 0.85 & 0.97 & 1.00 & 0.42 \\ 0.18 & 0.22 & 0.20 & 0.06 & 0.15 & 0.13 & 0.53 & 0.45 & 0.42 & 1.00 \end{pmatrix}$$

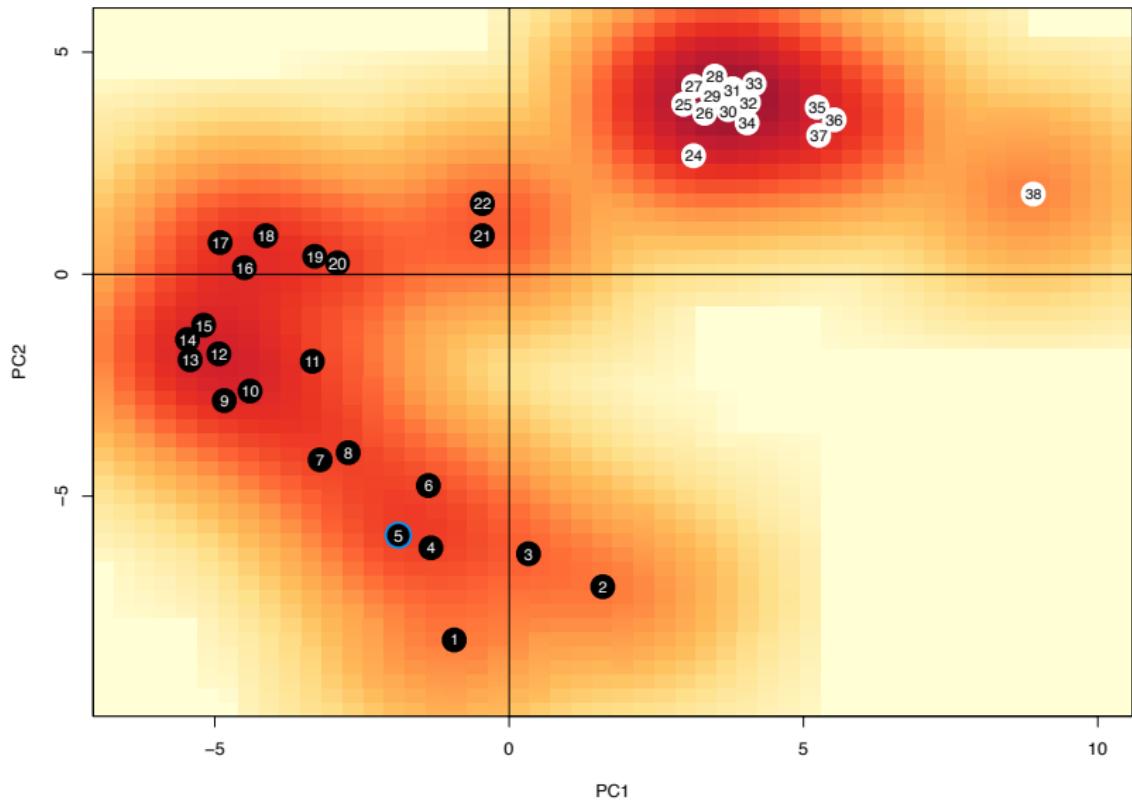
19: Citation PageRank  
 5: Journal Impact Factor  
 22: Citation Betweenness  
 6: Citation Closeness  
 11: Citation H-index  
 1: Citation Scimago Journal Rank  
 31: Usage PageRank  
 34: Usage Betweenness  
 24: Usage Closeness  
 39: Usage Impact Factor

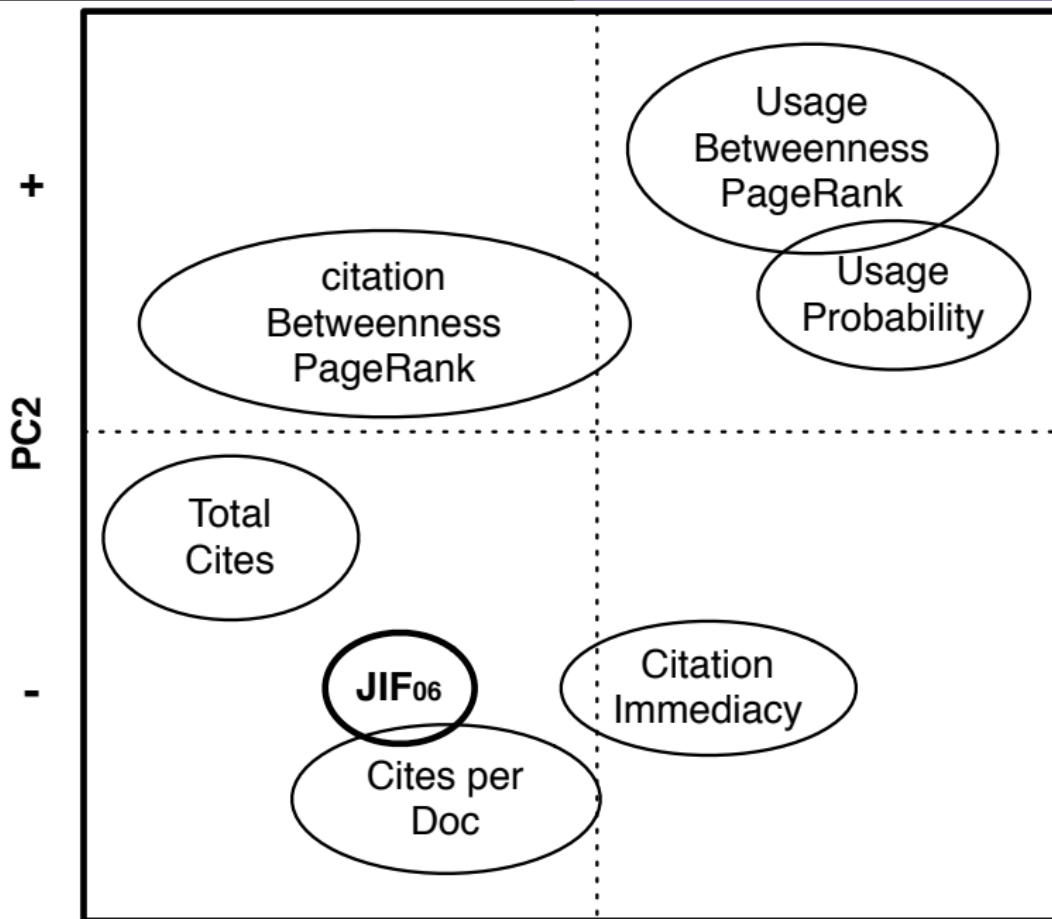
**data sources**

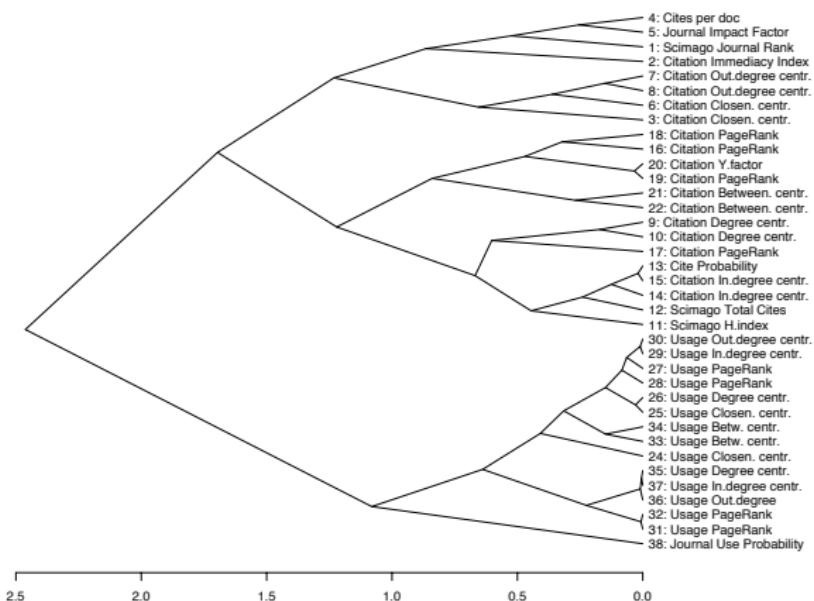
Schematic representation of data sources and processing. Impact measure identifiers.

# Principal Component Analysis

	PC1	PC2	PC3	PC4	PC5
Proportion of Variance	66.1%	17.3%	9.2%	4.8%	0.9%
Cumulative Proportion	66.1%	83.4%	92.6%	97.4%	98.3%







Cluster	Measures	Interpretation
1	38	Journal Use Probability
2	24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37	Usage measures
3	1, 2, 3, 4, 5	JIF, SJR, Cites per Document measures
4	6, 7, 8, 9, 10, 11, 12, 13, 14, 15	Total Citation rates and distributions
5	16, 17, 18, 19, 20, 21, 22	Citation Betweenness and PageRank

# MESUR's end-user analytics services

Aimed at showcasing our results

Demos and prototypes:

<http://mesur.informatics.indiana.edu/demos/>

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# MESUR: so far

Usage data: single largest reference data set of usage, citation and bibliographic data

- +1,000,000,000 usage events loaded
- multiple publishers, aggregators and institutions
- Infrastructure for research program

Usage graphs: track scientific flow of activity

- real-time studies of science
- inclusive of larger “scholarly community”

Metrics: valid, vetted indicators of scientific impact

- Different facets of scholarly impact
- Simple metrics, good results. Law of diminishing returns?
- Hybrid, consensus metrics?

# MESUR: the bad

Sustainability: burden of data collection

- ad hoc, customized agreements with data providers
- restrictive agreements with regards to data sharing
- high costs of maintaining infrastructure: funding

Research program: too much to do

- 3rd party access: better science
- many eyes on data

Metrics and services: Community support and acceptance

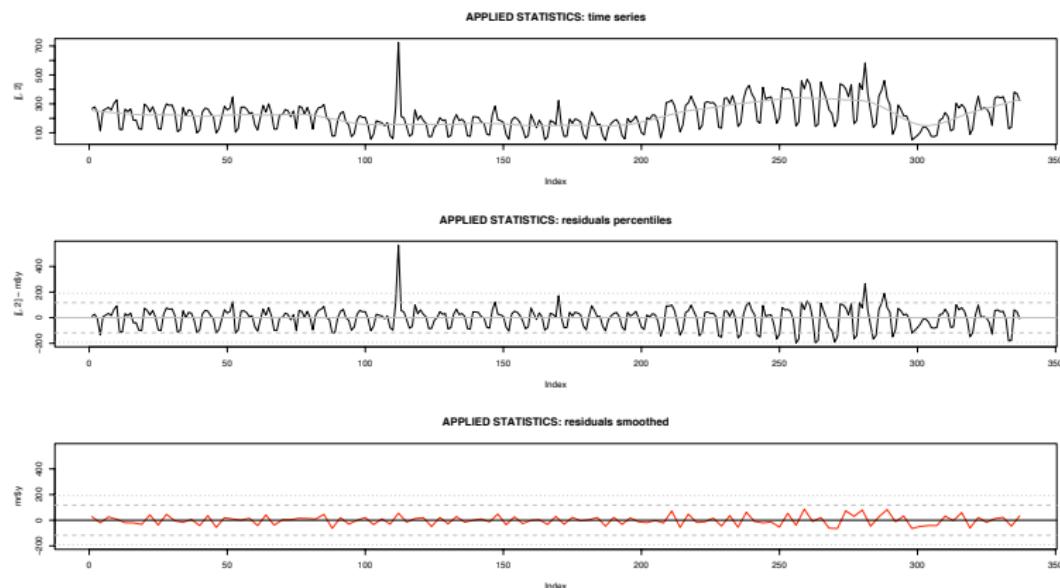
- Can't be limited to academic exercise
- Investigations need to be useful
- Accepted by community, become part of scholarly assessment system

# Future Research: Two directions

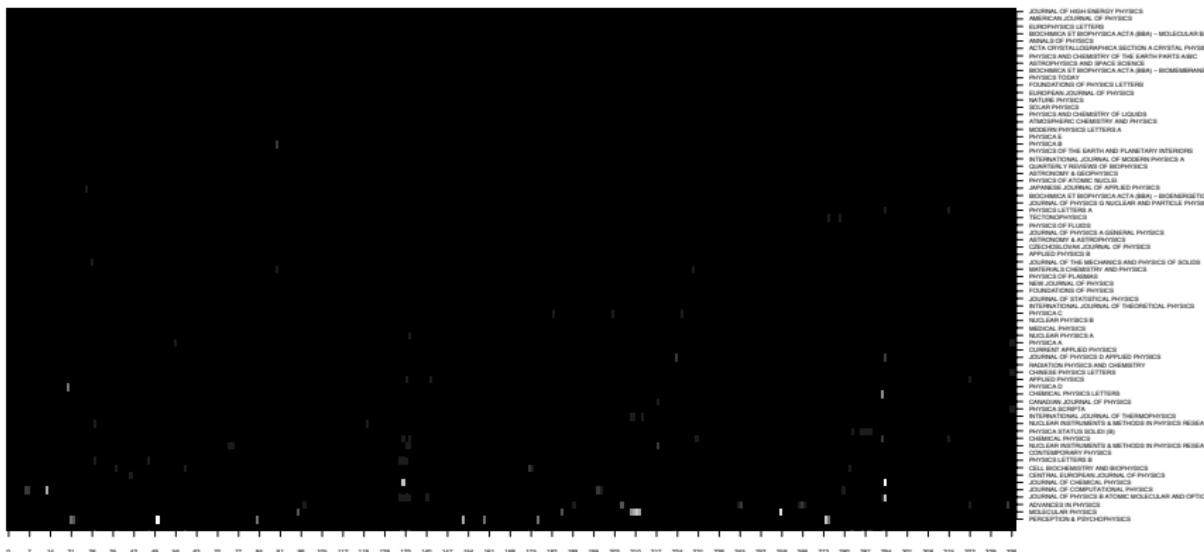
## Longitudinal research and dynamic models of science

- Bursty behavior of science
  - Timeseries of reads over time are bursty
  - Coordinated: social networks at work?
- Modeling
  - Stochastic and agent-based models of scientific activity
  - Citation following? Group-think? Find parameters of human information searching behavior.

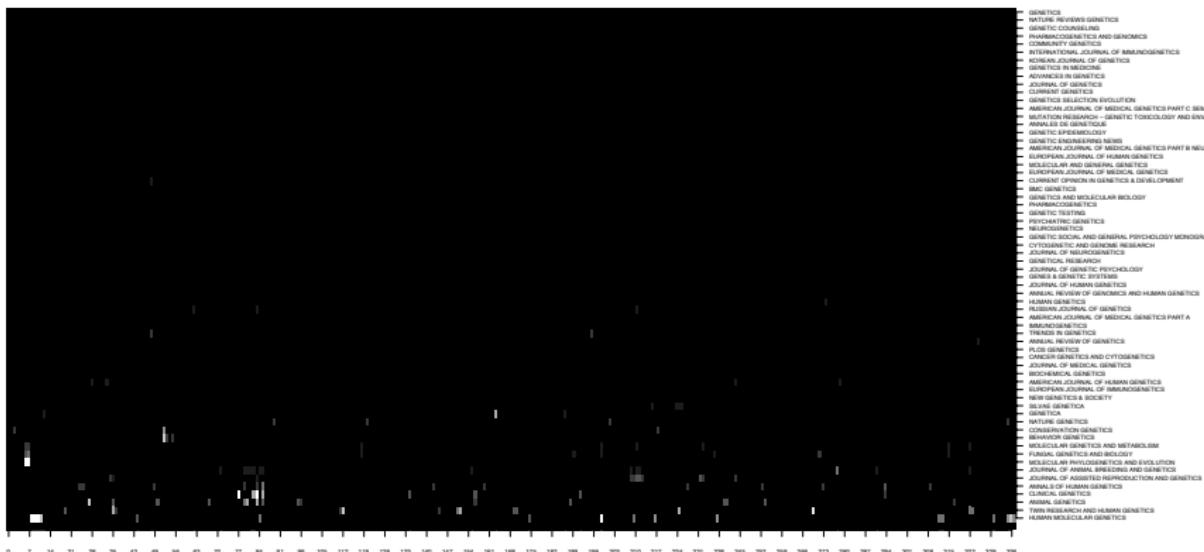
# Bursty behavior



# Contagion?



# Contagion?



# Relevant papers



Johan Bollen, Herbert Van de Sompel, Aric Hagberg and Ryan Chute

A Principal Component Analysis of 39 Scientific Impact Measures.

*PLoS ONE*, June 2009. URL:

<http://dx.plos.org/10.1371/journal.pone.0006022>.



Michael Kurtz and Johan Bollen.

Usage bibliometrics.

*Annual Review of Information Science and Technology*, 2010



Johan Bollen, Herbert Van de Sompel, Aric Hagberg, Luis Bettencourt, Ryan Chute, Marko A. Rodriguez, Lyudmila Balakireva.

Clickstream data yields high-resolution maps of science.

*PLoS One*, February 2009.