Mapping the Structure and Evolution of Science

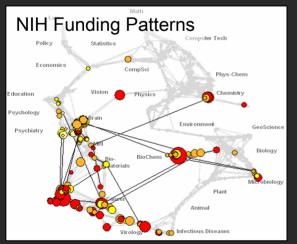
Dr. Katy Börner

Cyberinfrastructure for Network Science Center, Director Information Visualization Laboratory, Director School of Library and Information Science Indiana University, Bloomington, IN katy@indiana.edu

Symposium on Knowledge Discovery and Knowledge Management Tools at NIH Natcher Conference Center, Bethesda February 6, 2006.









Towards Effective Knowledge Discovery and Knowledge Management Tools for NIH

The Talk has Three Parts:

- 1. Mapping Aging Research
- 2. Improved Scholarly Data Collection
- 3. Computational Scientometrics

- = NIH Needs
- = NIH ToDo
- $= \overline{NIH} \overline{KM} \overline{Tool}$

Case Study: Mapping Aging Research

Part of this work was previously presented at the NIH Planning Workshop: "Assessments of Behavioral and Social Science Research Vitality" on Nov 9th, 2001.

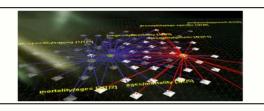
Publication:

Boyack, Kevin W. and Börner, Katy. (2003). Indicator-Assisted Evaluation and Funding of Research: Visualizing the Influence of Grants on the Number and Citation Counts of Research Papers, *Journal of the American Society of Information Science and Technology*, Special Topic Issue on Visualizing Scientific Paradigms, 54(5):447-461.

Funding Support:

Börner, Katy & Kevin W. Boyack. Mapping Aging Research. NIH Evaluation Express award on "Investigating Methodologies for Assessing Research Excellence in the Social and Behavioral Sciences" to Dr. Richard Suzman, BSR, NIA, Aug.-Dec. 2001





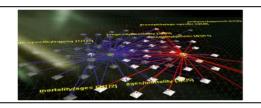
Mapping Aging Research: Client Needs



NIA's Most Actionable Questions

- 1. What are the major research areas, experts, institutions, regions, grants, publications, journals in aging research?
- 2. Which areas are most insular?
- 3. What are the main connections for each of our areas?
- 4. What is the relative speed of areas? Which areas are the most dynamic/static?
- 5. What new research areas are evolving?
- 6. Citation rate of grantees compared with other S&B&Life scientists?
- 7. How influential have NAS reports been?
- 8. How does NIA funding influence the number and quality of publications?
- 9. Impact of NIA funded research on other fields as measured by press coverage, Congressional presentations, etc.?
- 10. How can we measure progress?**

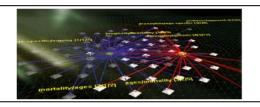
** Needs clarification.



Mapping Aging Research: Client Needs

NIAs Indices of an Individual Researcher's Productivity

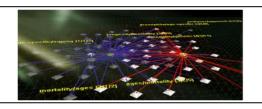
- Success/Discovery stories.
- Publications in top science journals such as *Science*, *Nature*, *PNAS*.
- Most significant publications in disciplinary journals or NIA volumes and publications.
- Citations of papers.
- Recent major honors received by funded investigators such as NAS membership, Nobel prizes, etc.
- Media coverage of grant research.
- Recent public policy activities by funded investigators. Impact on policy and practice.
- Export of ideas to other fields.



Mapping Aging Research: Client Needs

NIAs Indices of the Vitality of Research Field

- Average age of researchers in a field.
- Employment of post docs in the field.
- Employment of post docs in the private sector.
- Rapidity of publication of new results in leading journals in the field.
- Speed of promotion and tenure.
- Concentrations at Centers of Excellence; size of project teams over time; number of projects with multiple institutions.



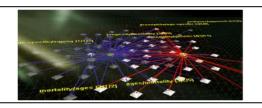
Mapping Aging Research: Datasets Used

Publication Data

- Medline publication data (Query for 'aging' in title, abstract or MeSH, 1996-2000) resulting in 28,700+ articles. (Query for selected 37 journals with >= 10 articles represented in NIA publication data) resulted in 101,800+ articles. Concerns: No citation links. Medline does not cover all relevant journals such as Journal of Marriage and the Family, Research on Aging, American Economic Review, Journal of Cross-cultural Gerontology, Journal of Family Issues, Aging Neuropsychology & Cognition.
- ➤ <u>ISI publication data</u> incl. citation links. 2,296 papers by 32 merit scholars, 26,880 papers that cited those 2,296, and the 14,828 ISI-indexed papers cited in (or by) the 2,296 for 1981-June 2001.

Concerns: For the demonstration work, we received no abstracts or keywords.

General Concerns: No data on books, media coverage, awards, etc.



Mapping Aging Research: Datasets Used

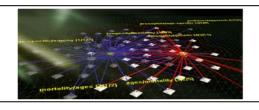
NIA/BSR Data

NIA grant data (PI names, institutions, years, award amounts, titles, abstracts, MeSH terms etc. for 1975-2001) resulting in 33,448 (sub)grants

Concerns: Different name formats, changing BAS codes. Most grants are

multi-year grants with several subcontracts, changing names & titles, etc.

- ► <u>BSR accomplishment reports</u> (CFA, DPE, HCO, PHB, PSP) resulting in 4,500+ publications
 - Concerns: grant publication linkage is not unique, grants by non-PI's are often missing, many duplicates.



Mapping Aging Research: Datasets Used

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Mental and Physical Health Trajectories in Adulthood

Bibliography

Spiro, A. III, Aldwin, C. M., Ward, K. D., & Mroczek, D. K. (1995). Personality and the incidence of hypertension among older men: Longitudinal findings from the Normative Aging Study. Health Psychology, 14, 563-569.

Aldwin, C. M., Ober, B. A., & Levenson, M. R. (1996). The psychology of aging. In D. B. Reuben, T. T. Yoshikawa, R. W. Besdine, & B. R. Popecki (Eds.). *Geriatrics Review Syllabus (GRS-III)* (pp. 24-29). New York: American Geriatrics Society.

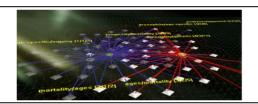
NIA Behavioral and Social Research Program DEMOGRAPHY AND POPULATION EPIDEMIOLOGY UNIT Partial Grantee Bibliography

Publications, 1992-present, in top journals in science (Science, Nature, PNAS) or medicine (NEJM, Lancet, JAMA, BMJ, Cell).

Bruce M. Alberts, Francisco J. Ayala, David Botstein, Ellen Frank, Edward W. Holmes, Ronald D. Lee, Eduardo, R. Macagno, Philippa Marrack, Suzanne Oparil, Stuart H. Orkin, Arthur H. Rubenstein, Carolyn W. Slayman, P., Frederick Sparling, Larry R. Squire, Peter H. von Hippel, and Keith R. Yamamoto. "Grant Review Process. Proposed Changes for NIH?s Center for Scientific Review" <u>Science</u> 1999 July 30; 285: 666-667. (in Policy Forum)

Bosma H, Marmot MG, Hemingway, H et al. 1997. Low Job Control and Risk of Coronary Heart Disease in Whitehall II (Prospective Cohort) Study. BMJ 314(7080): 558-65.

Figure 2. Example extracts (PSP and DPE, respectively) from BSR accomplishment reports.



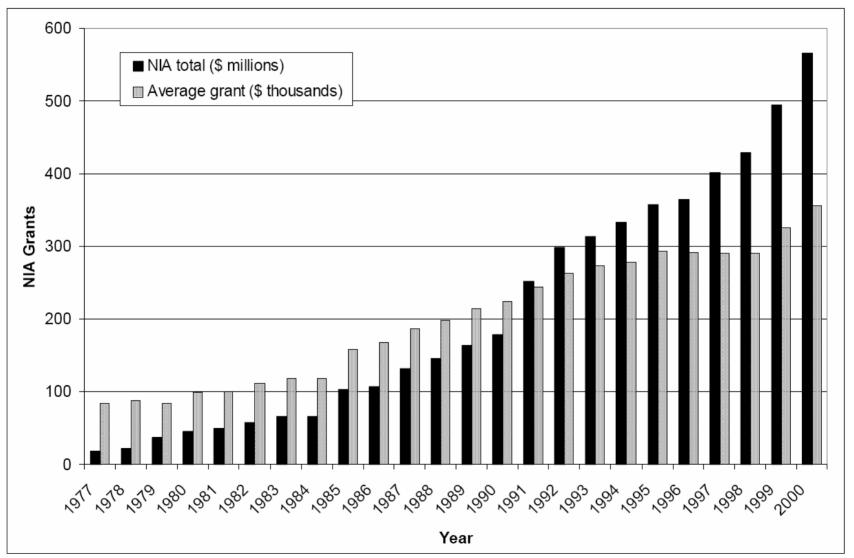
Mapping Aging Research: Data Analysis Top Journals

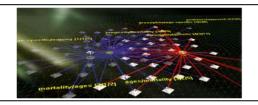
Table 2. Top journals represented in the BSR publication data (* 3-year impact factor)

Table 2. Top journals represented in the BSK public	meron contin	(b jeni	impact iat	
	Number	Medline	ISI	IF9599
Gerontologist	165	yes	yes	1.73
Journals of Gerontology Series B - Psychological Sciences	160	yes	yes	1.33
and Social Sciences				
Psychology and Aging	102	yes	yes	2.03
Journal of the American Geriatrics Society	81	yes	yes	2.66
Journals of Gerontology Series A - Biological Sciences and	75	yes	yes	1.02
Medical Sciences				
Journal of Marriage and the Family	48	no	yes	1.43
American Journal of Public Health	46	yes	yes	3.23
Research on Aging	40	no	yes	0.61
American Journal of Epidemiology	40	yes	yes	3.86
Journal of Aging and Health	37	yes	yes	0.85*
Demography	33	yes	yes	1.70
American Economic Review	30	no	yes	1.77
Health Psychology	29	yes	yes	2.67
JAMA - Journal of the American Medical Association	28	yes	yes	9.44
International Journal of Aging & Human Development	23	yes	yes	0.48
Experimental Aging Research	21	yes	yes	0.60
Social Science & Medicine	21	yes	yes	1.30
Medical Care	21	yes	yes	2.24
Journal of Family Issues	20	no	yes	0.86
Journal of Clinical Epidemiology	20	yes	yes	1.80
Aging Neuropsychology and Cognition	19	no	yes	0.86*
Journal of Human Resources	19	no	yes	1.17
Annals of Behavioral Medicine	19	yes	yes	1.47*
Journal of Health And Social Behavior	19	yes	yes	2.53
Neurology	18	yes	yes	4.80
Journal of Applied Gerontology	17	no	yes	0.38
Psychosomatic Medicine	17	yes	yes	2.94
Archives of Internal Medicine	17	yes	yes	5.14
Science	16	yes	yes	23.84
Journal of General Internal Medicine	16	yes	yes	1.80



Mapping Aging Research: Data Analysis # NIA Grants and Average Grant Amounts per Year



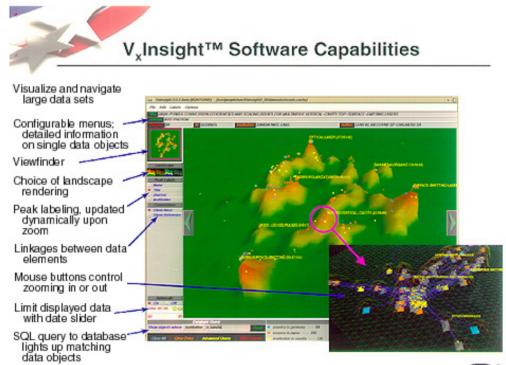


Mapping Aging Research: Visualization

Subsequent slides show visualizations of

- BSR grants and publications.
- Show impact of funding on number of publications over time.
- Grant input and publication output for U Michigan and Duke U.
- Author supplied linkage patterns for one grant and its publications.

VxInsight is used for interactive visualization.



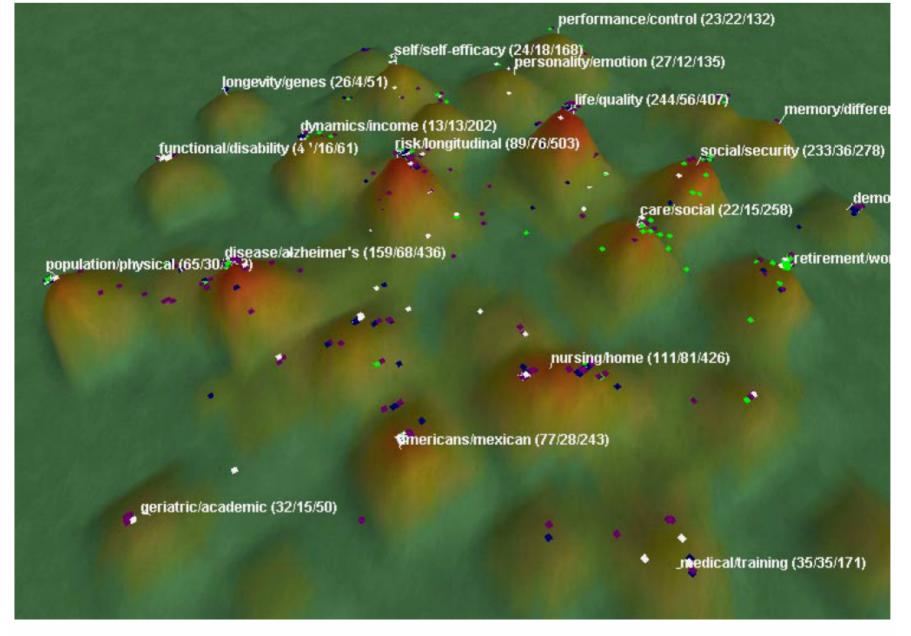
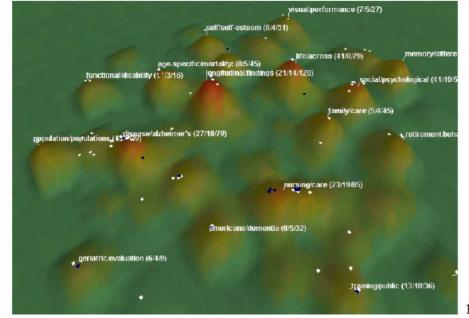


Figure 3. Map of BSR grants and publications using MeSH words as a basis for clustering. The height of each peak is proportional to the number of documents in the peak. Labels show the two most common words in the titles of the documents in each cluster (peak). Query markers for papers with certain words in either the title or MeSH words are shown as colored dots: disability – white, retirement – light gray, economics – dark gray, demography(ics) - black.



1995-1996

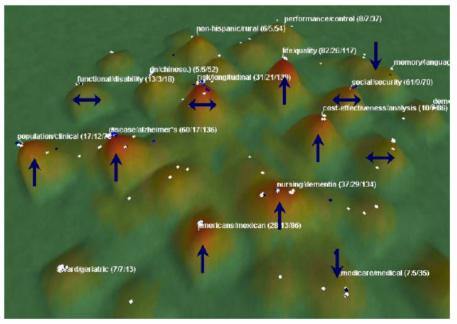


Figure 4. BSR map in two different time periods showing the impact of funding on the number of articles (size of peaks). Small grants are shown by light colored markers on the terrain. Large grants are shown by dark colored markers. Significant changes in the number of publications over time by peak are shown by dark arrows.

1999-2000

Table 4. Grant and citation data for 30 institutions funded by NIA.

	Grants		Awardee papers		Citing papers	
	1993-	1997	1995-	1999	1995-	1999
Institution	# Grant	\$M	# Papers	Cites /	# Papers	Cites /
	Records			paper		paper
University of Michigan	373	89.9	81	11.73	254	1.76
Duke University	314	69.1	71	13.04	198	1.79
Johns Hopkins University	252	63.3	2	21.00	189	1.40
University of California at San Diego	225	55.1	4	5.25	82	1.23
University of Washington	267	53.7	8	7.63	161	1.47
University of California at San Francisco	218	49.2	29	10.76	184	1.58
University of Southern California	230	48.8	19	10.05	113	1.88
University of Texas (system)	302	43.9	8	12.63	207	1.45
Case Western Reserve University	266	40.8	22	4.77	90	1.67
Washington University	251	36.7	3	8.00	83	1.67
University of California at Los Angeles	165	31.7	10	9.70	228	1.61
Massachusetts General Hospital	169	31.4			46	1.17
University of Pennsylvania	161	30.2	23	9.83	122	1.44
University of Pittsburgh	152	29.3	6	11.50	135	1.54
Columbia University	138	29.0	5	1.60	121	1.57
Rush Presbyterian – St. Luke's Medical Center	119	27.8			33	1.97
University of Kentucky	174	27.6			37	1.41
Boston University	146	27.2	23	5.17	123	1.59
New York University	165	27.0			43	1.47
Harvard University	154	25.5	19	13.16	321	1.60
Indiana University	105	25.1	30	12.43	80	1.61
Mayo Clinics & Mayo Foundation	101	24.8			30	1.27
University of Maryland	158	24.6	2	2.00	90	1.70
Mt. Sinai School of Medicine	150	23.6			71	1.34
University of Wisconsin	108	21.9	3	19.67	133	1.65
University of Colorado	148	21.7	3	26.33	67	1.76
Pennsylvania State University	99	21.2	12	4.67	115	1.54
Stanford University	134	21.0	39	14.36	180	1.63
University of Alabama	121	19.0	12	6.33	95	1.44
Brigham & Women's Hospital	97	17.8			53	1.64
TOTALS / AVERAGES	5462	35.3	434	10.85	3684	1.58

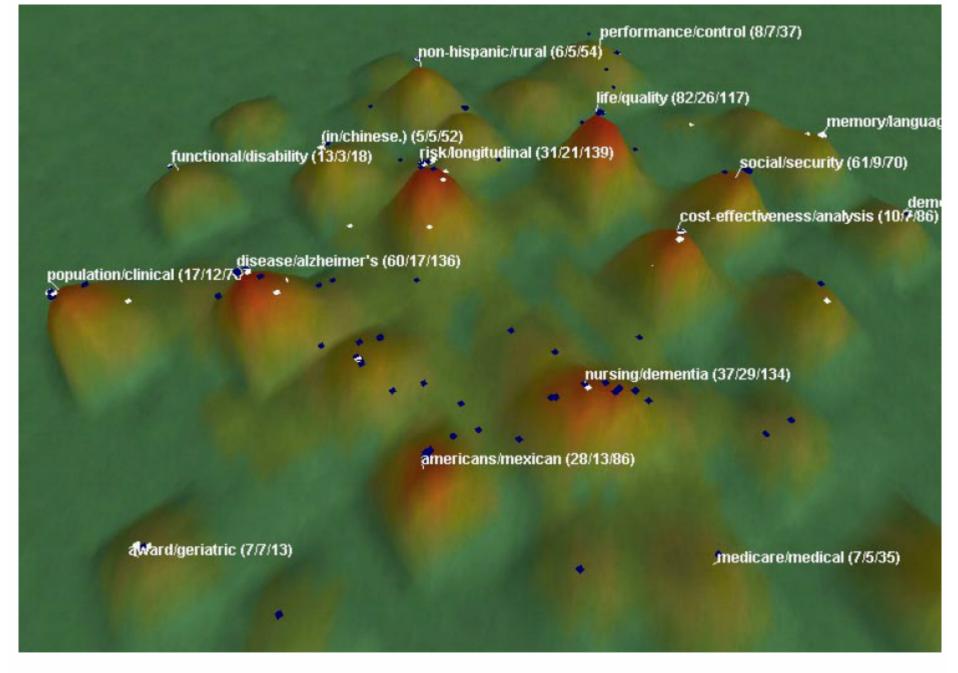
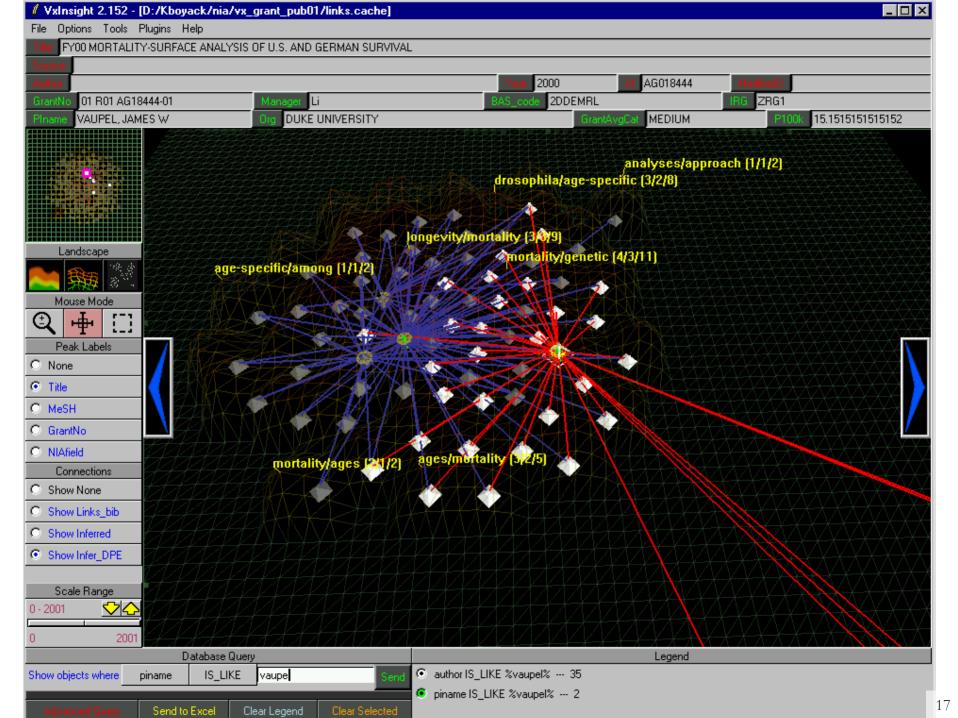
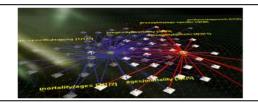


Figure 5. NIA grants awarded to and BSR-related publication output of the University of Michigan (dark markers) and Duke University (white markers) in the BSR domain during the 1999-2000 time period.



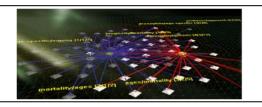


Recommendations

Create a clean and maintainable 'Scholarly Database' database for NIA grants & publications as a basis for the application of scientometric methods.

- NIA enters new program announcements, grants, etc.
- Researchers enter relevant data (papers, patents, media coverage, policy changes, honors, etc.), short "new result(s)/impact headline" in standard form entries.
- > Publishers input publications, e.g., Medline records, impact factors, etc.

Make data and results of scientometric analysis publicly available.



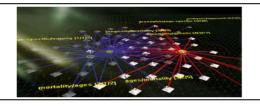
Recommendations cont.

This Scholarly Database & Associated KM Tools would

- Enable cross-disciplinary information access & transfer between different research areas.
- Reveal unproductive duplication, unrealized complementarity, gaps & opportunities, and overlapping topics.
- ➤ Help to facilitate (cross-disciplinary) collaborations and to establish research priorities.
- Provide opportunities for each scientist to think systematically about the state of their field.
- ldeally help achieve consensus on what should be funded.

Advantages for BSR/NIA

- Supports monitoring of (long-term) money flow and research developments, evaluation of funding strategies for different programs, decisions on project durations, funding patterns.
- > Staff resources can be used for scientific program development, to identify areas for future development, and the stimulation of new research areas.



Recommendations cont.

Advantages for Researchers

- Easy access to research results, relevant funding programs and their success rates, potential collaborators, competitors, related projects/publications (research push).
- More time for research and teaching.

Advantages for Industry

- Fast and easy access to major results, experts, etc.
- Can influence the direction of research by entering information on needed technologies (industry-pull).

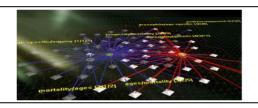
Advantages for Publishers

- Unique interface to their data.
- > Publicly funded development of databases and their interlinkage.

For Society

Dramatically improved access to scientific knowledge and expertise.

Improved Collection, Integration, Analysis and Management of Scholarly Knowledge & Expertise



The Challenge: Data Collection & Integration

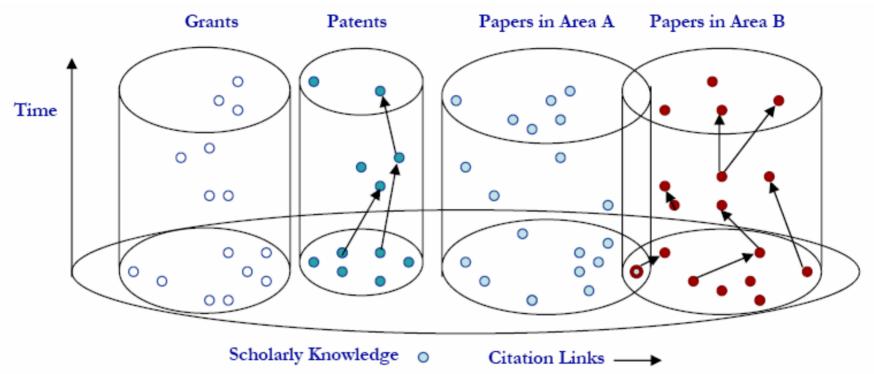
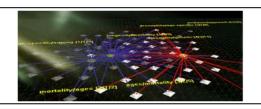


Figure 1: The interoperability and cross linkage problem. Many but not all of today's scholarly datasets, e.g., papers, patents, grants, are stored and made available so that 'vertical' citation linkages can be traversed. There are very few instances in which datasets of different origin and/or type are 'horizontally' interlinked.

Börner, K. (2006) Semantic Association Networks: Using Semantic Web Technology to Improve Scholarly Knowledge and Expertise Management. In Vladimir Geroimenko & Chaomei Chen (eds.) Visualizing the Semantic Web, Springer Verlag, 2nd Edition, chapter 11, pp. 183-198.



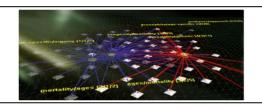
Text Mining & Data Mining Tools

Work well if the records are written in similar styles, using similar formatting and conventions, are of similar length, etc.

Work less well if applied to interdisciplinary or multi-lingual datasets because

- Words, e.g., 'prototype', have very different meanings in computer science, biology, psychology, architecture, etc.
- Paper titles are frequently used to demonstrate the creativity of authors, e.g., "All you ever wanted to know about x", "A unifying theory of x".
- Author supplied keywords are useful to identify an author but not to find similar papers. Note: Controlled vocabularies/thesauri work well.

Humans might simply be too different and too creative to produce proper raw material that can be analyzed using existing text mining and data mining algorithms.



Link Traversal & Link Mining

Link Search: Google

(Citation) Link Traversal in One Database: Thomson Scientific, Google Scholar, and CiteSeer already support citation link traversal. The Proceedings of the National Academy of Sciences of the United States of America (PNAS) online interface http://www.pnas.org>. provides citation maps that shows articles citing or being cited by a selected article.

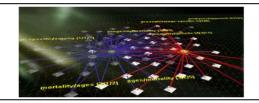
(Citation) Link Traversal Across Databases: The Library Without Walls project < http://library.lanl.gov/lww/ at the Los Alamos National Laboratory interlinks major publication databases and supporting citation based search across different holdings that have citation linkages.

(Co-Author) Link Traversal: Some digital libraries such as the citation indexes published by Thomson Scientific, DBLP Bibliography Server

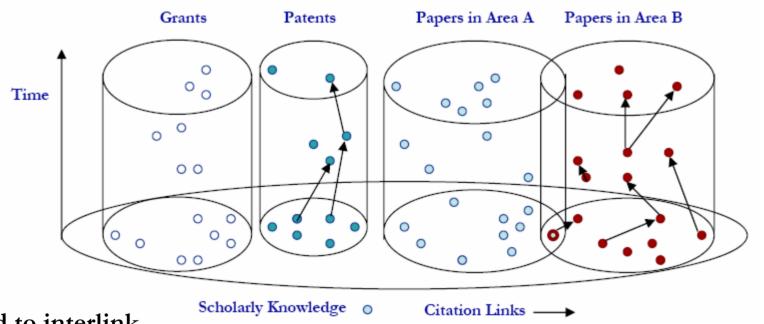
< http://www.sigmod.org/dblp/db/>, and ACM Digital Library

http://portal.acm.org provide information on co-authorships. Services comprise a listing of all papers by an author, a listing of all co-authors for one author, and co-author link traversal.

Link traversal and link mining might be a viable alternative.



The Challenge: Interlink \$ Input & Publication/Patent Citation Output



Need to interlink

- Grants and papers/patents.
- Grants/papers/patents and their PIs/authors/inventors, etc.

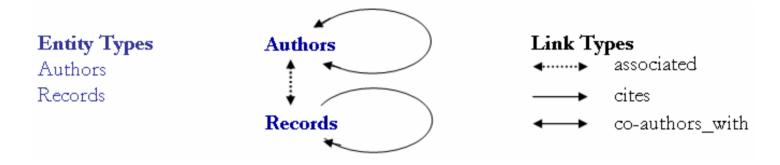
Use resulting networks to

- Count #papers, #citations, etc.
- Determine strength of co-PI/author/inventor relations, etc.



Improved Representation of Scholarly Knowledge

Entity and link types:

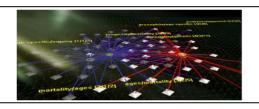


Attributes:

- Records often have a publication date, a publication type (e.g., journal paper, book, patents, grant, etc.), topics (e.g., keywords or classifications assigned by authors and/or publishers).
- Authors have an address with information on affiliation and geo-location.

Derived attributes:

- Because authors and records are associated, the geo-location(s) and affiliation(s) of an author can be attributed to the authors' papers.
- Similarly, the publication date, publication type and topic(s) can be associated with a paper's author(s).



Link Analysis

Statistics:

- Number of papers, grants, co-authorships, citation (over time) per author.
- > Bursts of activity (#citations, #\$, #patents, #collaborators, etc.).
- Changes of topics and geo-locations for authors and their institutions over time.

Visualizations:

- Geospatial and topical distribution of funding input & research output.
- Structure and evolution of research topics.
- Evolving research areas (e.g., based on young yet highly cited papers).
- Diffusion of information, people, \$s over geospatial and topic space.





Computational Scientometrics: Mapping the Structure and Evolution of Science





- Shiffrin, Richard M. and Börner, Katy (Eds.) (2004). Mapping Knowledge Domains. Proceedings of the National Academy of Sciences of the United States of America, 101 (Suppl_1).
- Börner, Katy, Chen, Chaomei, and Boyack, Kevin. (2003). Visualizing Knowledge Domains. In Blaise Cronin (Ed.), Annual Review of Information Science & Technology, Volume 37, Medford, NJ: Information Today, Inc./American Society for Information Science and Technology, chapter 5, pp. 179-255.

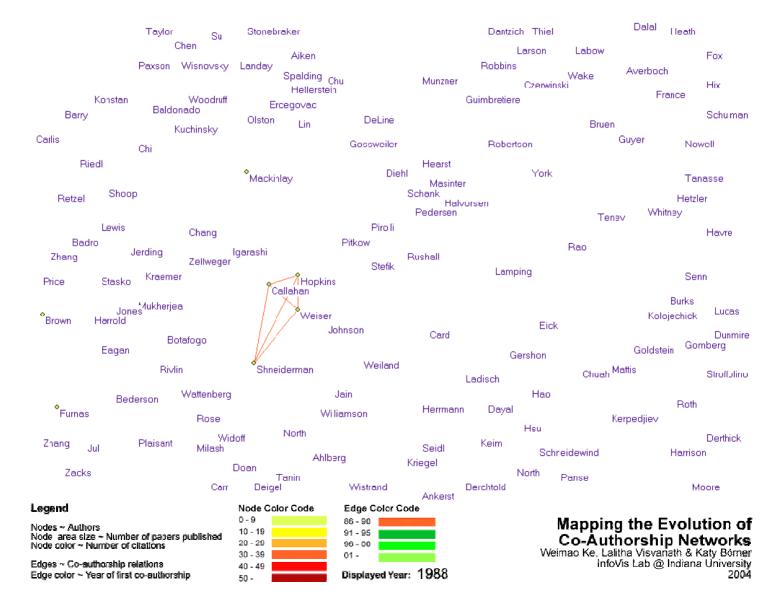
Part (I): Mapping Science Locally

Mapping the Evolution of Co-Authorship Networks





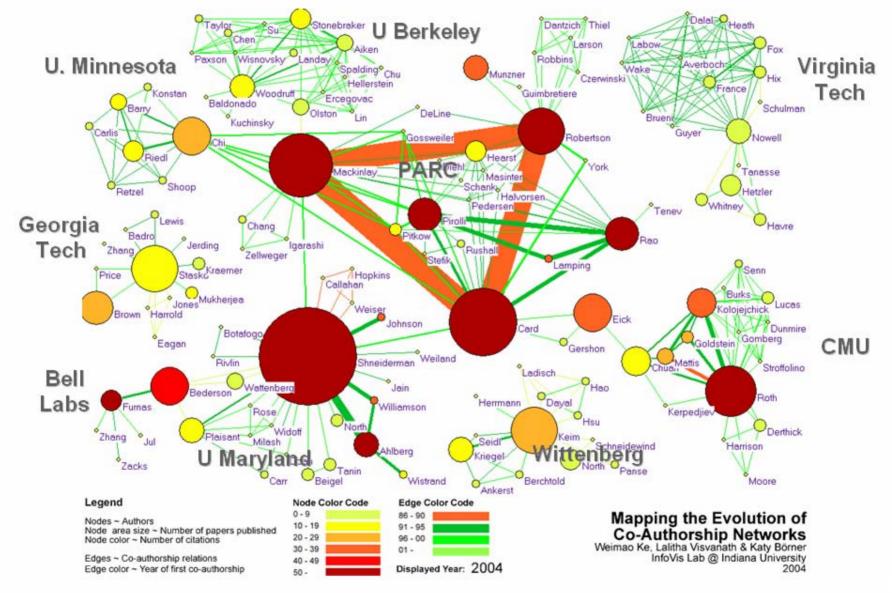
Ke, Visvanath & Börner, (2004) Won 1st price at the IEEE InfoVis Contest.



Mapping the Evolution of Co-Authorship Networks

Ke, Visvanath & Börner, (2004) Won 1st price at the IEEE InfoVis Contest.





Studying the Emerging Global Brain: Analyzing and Visualizing the Impact of Co-Authorship Teams

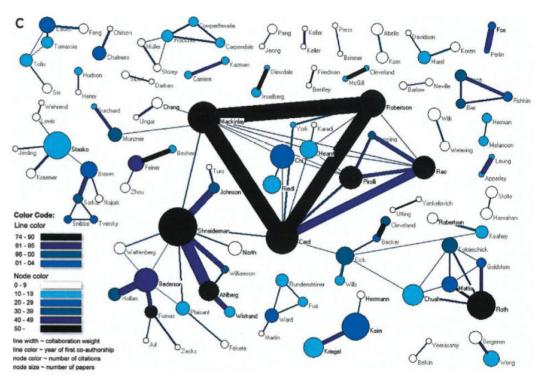
Börner, Dall'Asta, Ke & Vespignani (2005) Complexity, 10(4):58-67.

Research question:

➤ Is science driven by prolific single experts or by high-impact co-authorship teams?

Contributions:

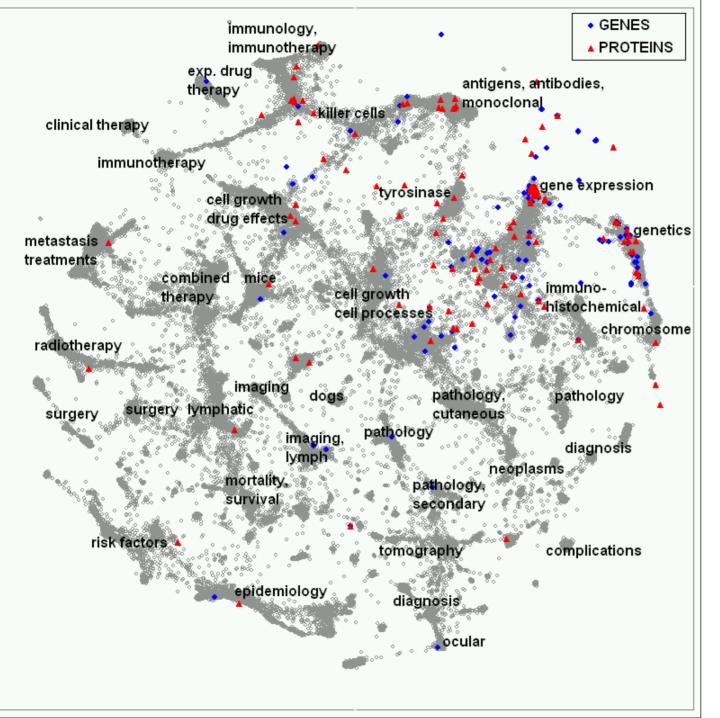
- New approach to allocate citational credit.
- Novel weighted graph representation.
- ➤ Visualization of the growth of weighted co-author network.
- Centrality measures to identify author impact.
- ➤ Global statistical analysis of paper production and citations in correlation with co-authorship team size over time.
- Local, author-centered entropy measure.











Mapping Medline
Papers, Genes, and
Proteins Related to
Melanoma
Research

Boyack, Mane & Börner. (2004) IV Conference, pp. 965-971.

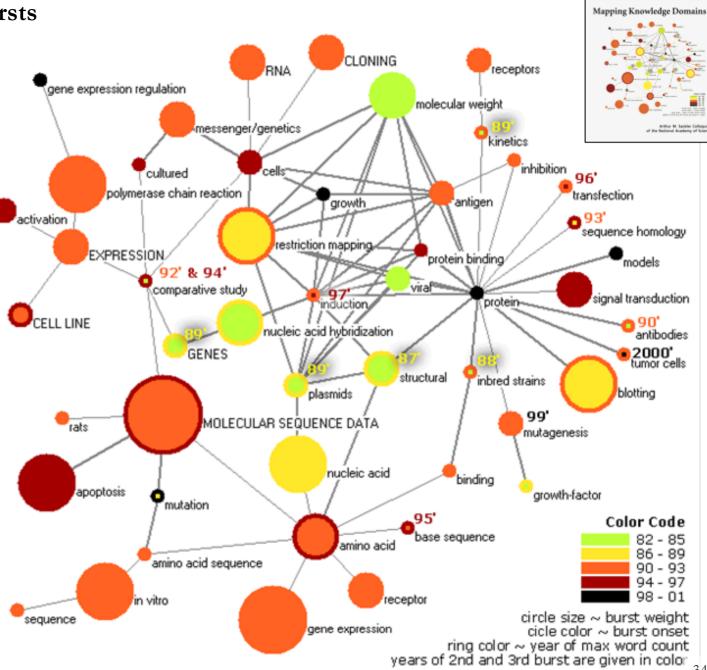


Mapping Topic Bursts

Co-word space of the top 50 highly frequent and bursty words used in the top 10% most highly cited PNAS publications in 1982-2001.

Mane & Börner. (2004) PNAS, 101(Suppl. 1): 5287-5290.





Spatio-Temporal Information Production and Consumption of Major U.S.

Research Institutions

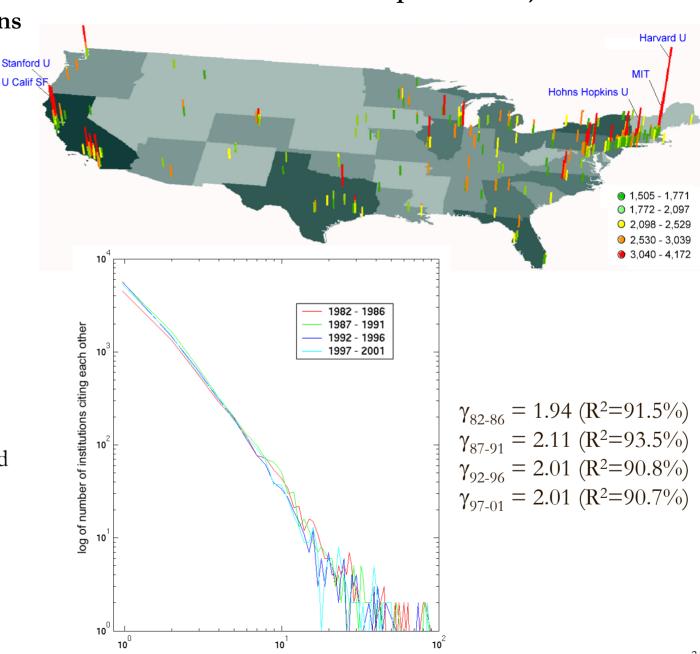
Börner & Penumarthy. (2005)

Does Internet lead to more global citation patterns, i.e., more citation links between papers produced at geographically distant research instructions?

Analysis of top 500 most highly cited U.S. institutions.

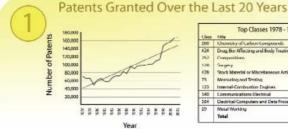
Each institution is assumed to produce and consume information.





log of geographic distance

Examining the Evolution and Distribution of Patent Classifications

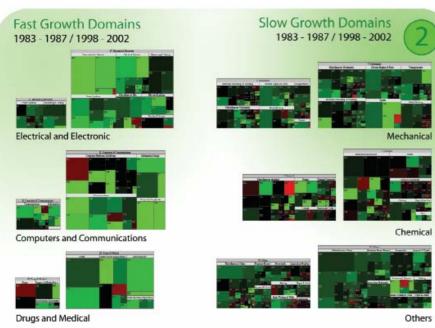




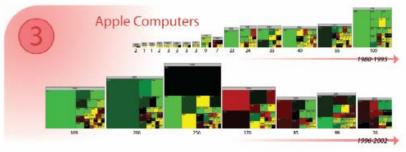
Top Classes 1998 - 2002					
Class Title					
514	Drug Bio-Milecting and Body Treating Composistions	18,778			
4RR	Senionochartor Device Manufarturing Process	17,77			
415	Chemistry, Molecular Biology and Wicrobiology	12,424			
474	Drug Bo-Affecting and Body Treating Compositions	13,630			
47R	Stock Material or Mescellaneous Amiries	13,31			
257	Active Solid-Grare Devices (e.g., Translators, Solid-soare Diodes	12,924			
395	Information Processing System Organization	0,955			
345	Computer Graphics Processing, Operator Interface Processing, and Selective Visual Display Systems	9,510			
359	Upbicat Systems and Elements	9,151			
365	Static Information Storage and Retrieval	8,397			
	Tetal	130,910			

In the United States, each patent gets assigned to one out of more than 450 classes covering broad application domains. An examination of the size and growth of patent classes provides insight about patenting trends.

Treemaps, a space filling technique developed in the HCI Lab at the University of Maryland, are used to communicate major results. Treemaps represent a tree structure as nested rectangles with each rectangle representing a node. A rectangular area is first allocated to hold the representation of the tree, and this area is then subdivided into a set of rectangles that represent the top level of the tree. This process continues recursively on the resulting rectangles to represent each lower level of the tree. The parent-child relationship is indicated by enclosing the child rectangle by its parent rectangle. Typically, the size of each rectangle corresponds to the size of the node. Additional information about a node, e.g., its age or value, can be represented by the color of the respective rectangle.



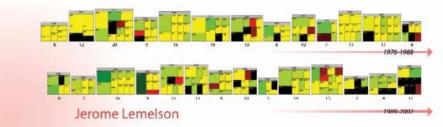
Shown is a comparison of the patent class space for 1983 to 1987 and 1998 to 2002. There is a predominance of growth in the 1998 to 2002 patent space, which correlates to the increase in patent grants during this period. By comparing the growth in categories, one can distinguish between domains that have been receiving a larger amount of patent grants.



Depicted above is how Apple Computers' portfolio has changed in yearly increments from 1980 to 2002.

Lemelson's patent holdings below show a more even distribution over multiple classes. No class dominates over a majority of the years for granted patents; instead they are distributed more broadly over the intellectual space.

Green = Increase in number of patent grants in particular class. Red = Decrease in number of patent grants in particular class. Yellow - No patents granted in that class in the past five years. Size = Number of patent grants in a particular class.



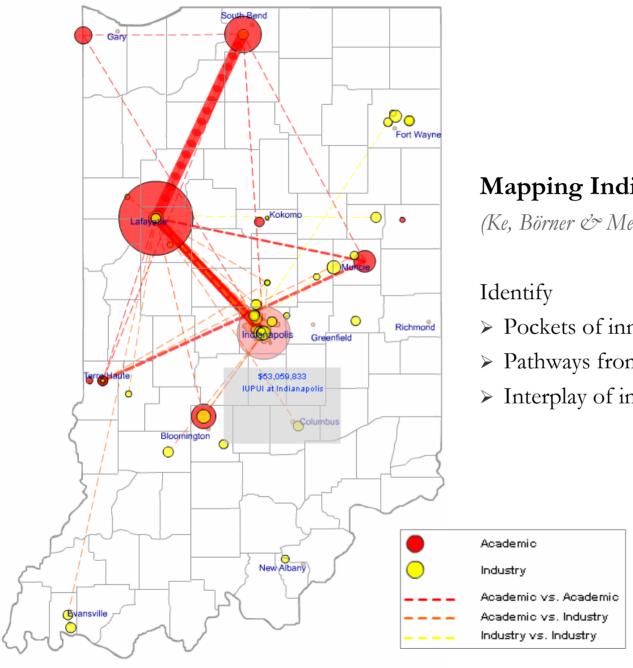


Kutz, Daniel O. Examining the Evolution and Distribution of Patent Classifications. Accepted for the Information Visualization Conference, London, UK, July 2004.

The material is based upon work supported bythe National Science Foundation under Grant No. IIS-0238261.



For more Information, contact Katy Borner at katy@indiana.edu.







Mapping Indiana's Intellectual Space

(Ke, Börner & Mei, 2005)

- > Pockets of innovation
- > Pathways from ideas to products
- > Interplay of industry and academia

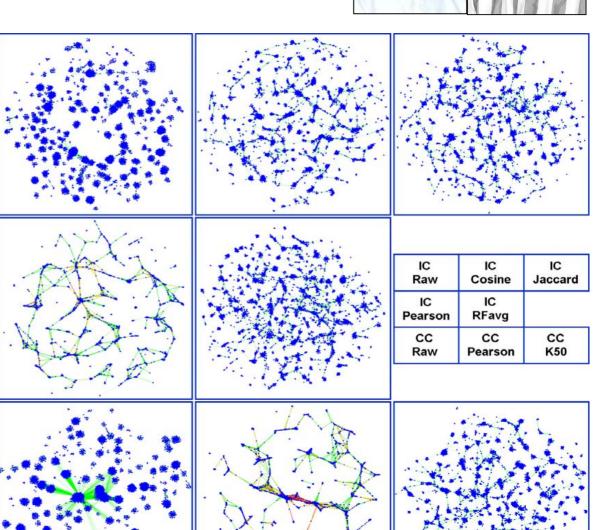
Part (II): Mapping Science Globally

Comparison of Similarity Metrics

- ➤ ISI file year 2000, SCI and SSCI: 7,121 journals.
- ➤ Different similarity metrics
 - Inter-citation (raw counts, cosine, modified cosine, Jaccard, RF, Pearson)
 - Co-citation (raw counts, cosine, modified cosine, Pearson)
- ➤ Maps were compared based on
 - regional accuracy,
 - the scalability of the similarity algorithm, and
 - the readability of the layouts.

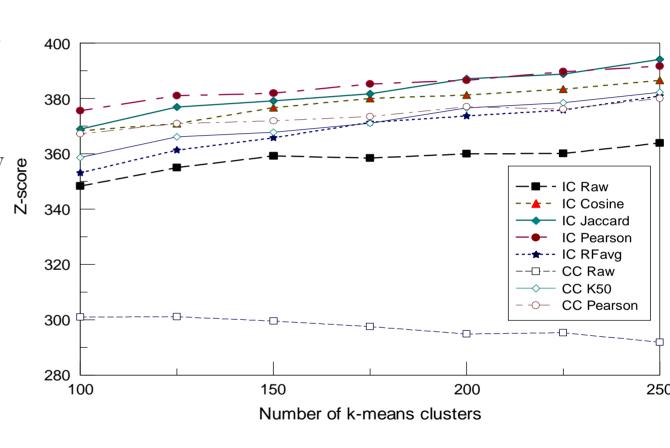
Boyack, Kevin W., Klavans, R. and Börner, Katy. (2005). Mapping the Backbone of Science. Scientometrics. 64(3), 351-374.





Selecting the similarity measure with the best regional accuracy

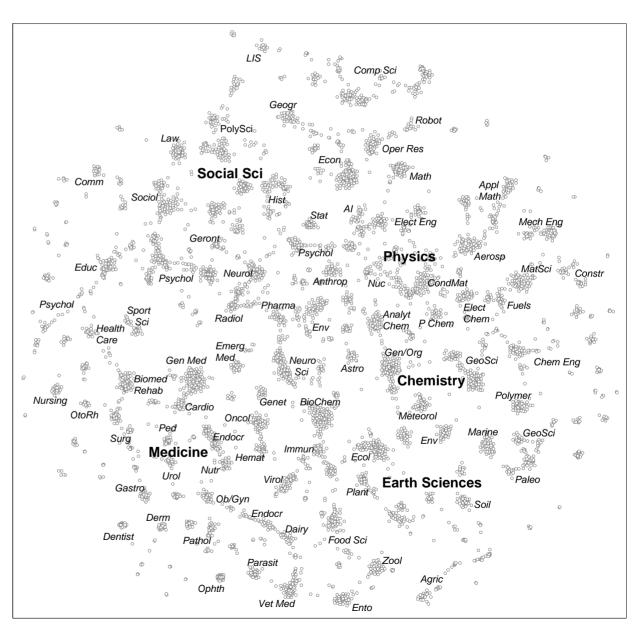
- For each similarity measure, the VxOrd layout was subjected to k-means clustering using different numbers of clusters.
- Resulting cluster/category memberships were compared to actual category memberships using entropy/mutual information method by Gibbons & Roth, 2002.
- Increasing Z-score indicates increasing distance from a random solution.
- Most similarity measures are within several percent of each other.

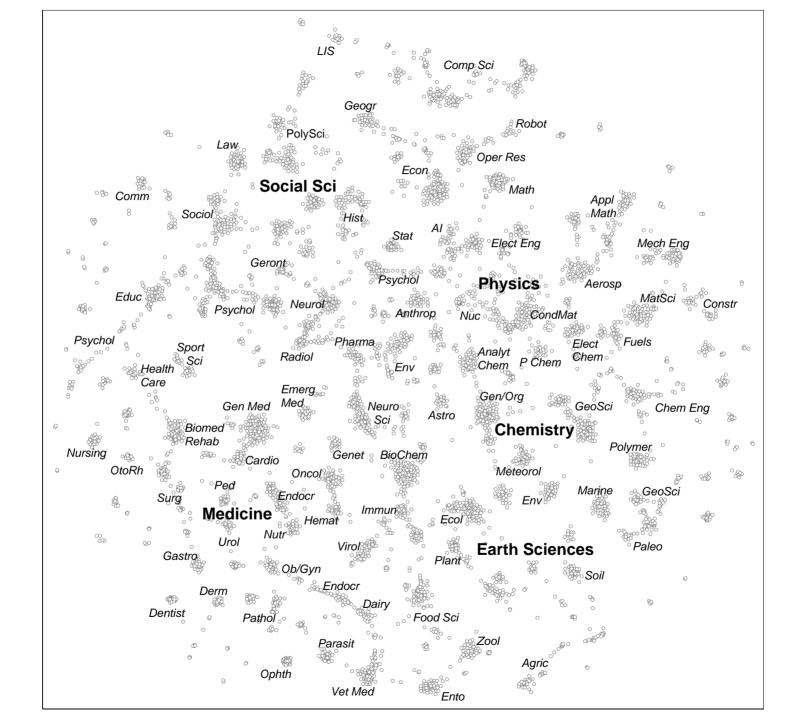


Boyack, Kevin W., Klavans, R. and Börner, Katy. (2005). Mapping the Backbone of Science. Scientometrics. 64(3), 351-374.

A 'Backbone' Map of Science & Social Science

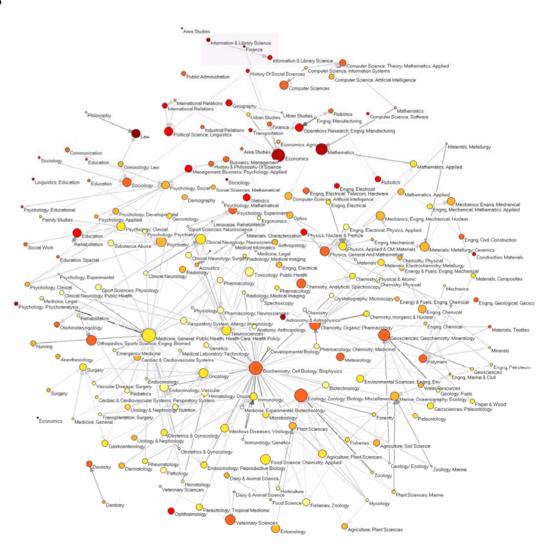
- The map is comprised of 7,121 journals from year 2000.
- Each dot is one journal
- An *IC-Jaccard* similarity measure was used.
- Journals group by discipline.
- Groups are labeled by hand.
- Large font size labels identify major areas of science.
- Small labels denote the disciplinary topics of nearby large clusters of journals.

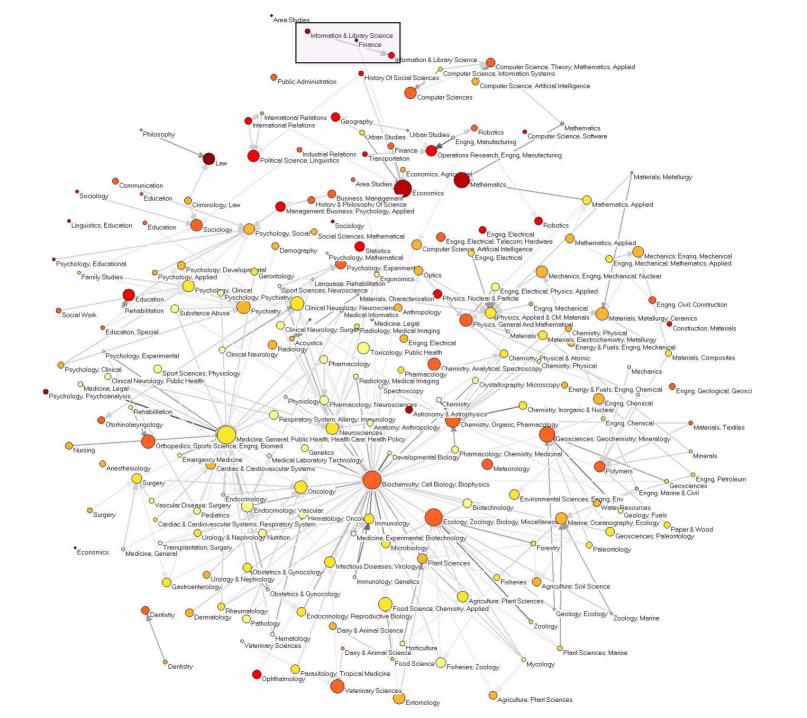




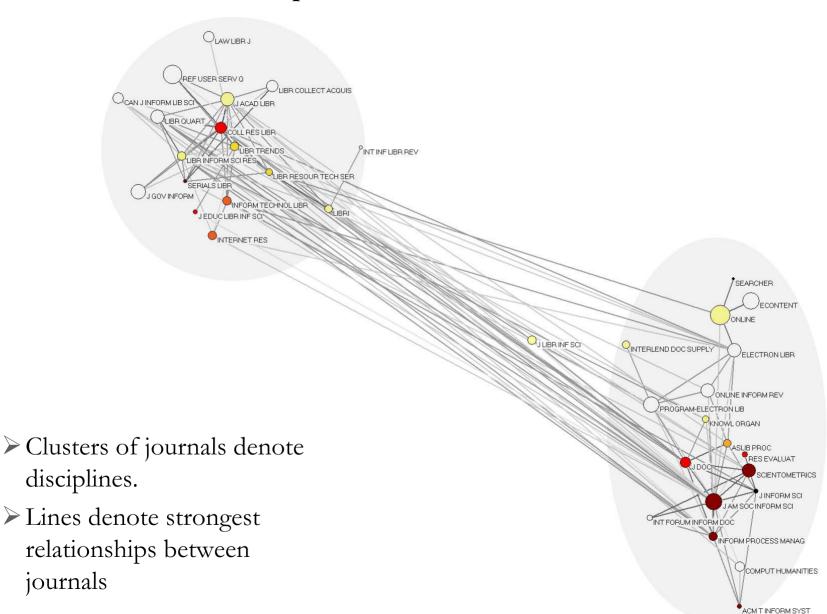
Structural Map: Studying Disciplinary Diffusion

- The 212 nodes represent clusters of journals for different disciplines.
- Nodes are labeled with their dominant ISI category name.
- Circle sizes (area) denote the number of journals in each cluster.
- Circle color depicts the independence of each cluster, with darker colors depicting greater independence.
- Lines denote strongest relationships between disciplines (citing cluster gives more than 7.5% of its total citations to the cited cluster).





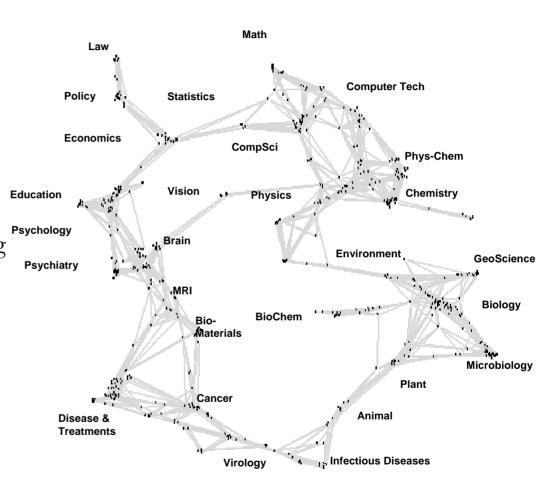
Zoom Into Structural Map



Latest 'Base Map' of Science

Kevin W. Boyack & Richard Klavans, unpublished work.

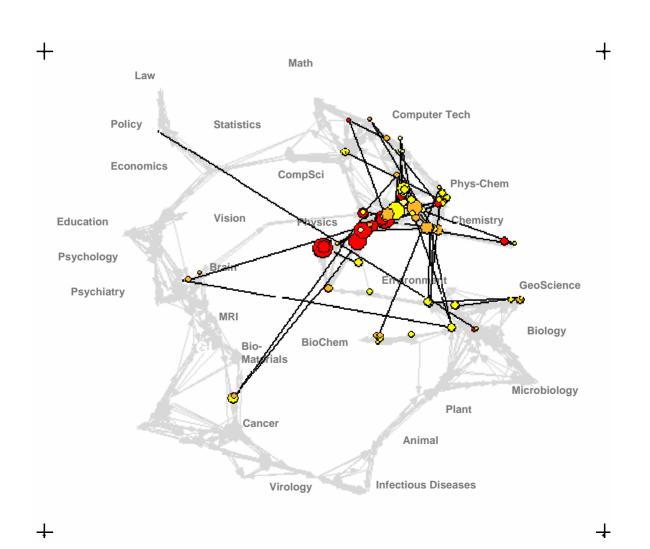
- ➤ Uses combined SCI/SSCI from 2002
 - 1.07M papers, 24.5M references, 7,300 journals
 - Bibliographic coupling of papers, aggregated to journals
- ➤ Initial ordination and clustering of journals gave 671 clusters
- Coupling counts were reaggregated at the journal cluster level to calculate the
 - (x,y) positions for each journal cluster
 - by association, (x,y)
 positions for each journal



Science map applications: Identifying core competency

Kevin W. Boyack & Richard Klavans, unpublished work.

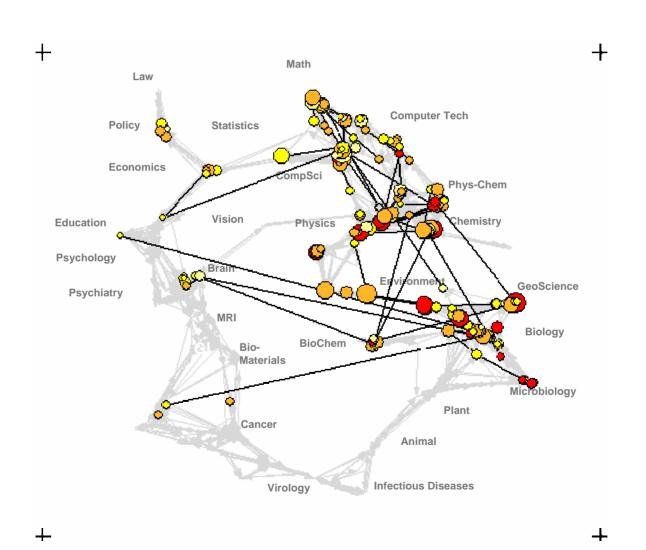
Funding patterns of the US Department of Energy (DOE)



Science map applications: Identifying core competency

Kevin W. Boyack & Richard Klavans, unpublished work.

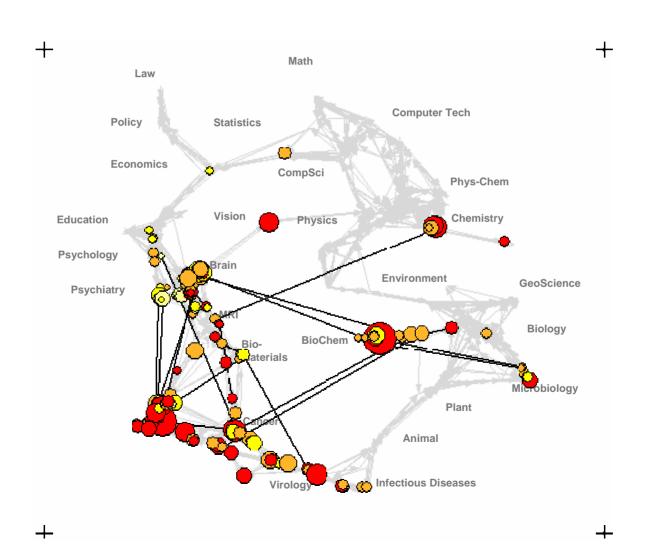
Funding Patterns of the National Science Foundation (NSF)



Science map applications: Identifying core competency

Kevin W. Boyack & Richard Klavans, unpublished work.

Funding Patterns of the National Institutes of Health (NIH)





Cartography of the Physical and the Abstract

An exhibition created for the conference "Mapping Humanity's Knowledge and Expertise in the Digital Domain" at the 2005 Meeting of the American Association of Geographers that is updated regularly with new maps and explainations.

Home

Browse Maps

Compare & Contrast Maps

Connect

Home



Exhibit Purpose and Goals

The Places &

Spaces exhibit has been created to demonstrate the power of maps.

An initial theme of this exhibit is to compare and contrast first maps of our entire planet with the first maps of all of science as we know it.



Come see with your own eyes the extent to which maps can be employed to help make sense of the flood of information we are confronted with and how domain maps can be used to locate complex and beautiful information.

This online part of the exhibit provides links to a selected series of maps and their makers along with detailed explanations of why these maps work. The physical counterpart supports the close inspection of high quality reproductions for display at conferences and education centers. It is meant to inspire cross-disciplinary discussion on how to best track and communicate human activity and scientific progress on a global scale.



Places & Spaces: Cartography of the Physical and the Abstract

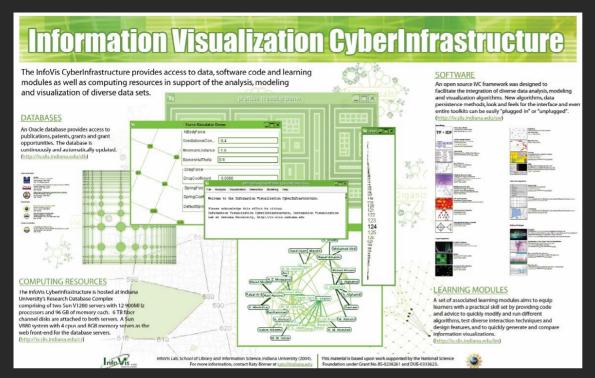
The first iteration of this science exhibit compares and contrasts first maps of our entire planet with the first maps of all of sciences.

<u> http://vw.indiana.edu/places&spaces/</u>







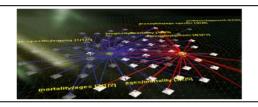


CAREER: Visualizing Knowledge Domains. NSF IIS-0238261 award (Katy Börner, \$440,000) Sept. 03-Aug. 08.





SEI: Network Workbench: A Large-Scale Network Analysis, Modeling and Visualization Toolkit for Biomedical, Social Science and Physics Research. NSF IIS-0513650 award (Katy Börner, Albert-Laszlo Barabasi, Santiago Schnell, Alessandro Vespignani & Stanley Wasserman, Craig Stewart (Senior Personnel), \$1,120,926) Sept. 05 - Aug. 08.



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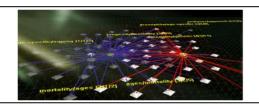






Pointers to Papers & Expertise

(not covered in the talk)



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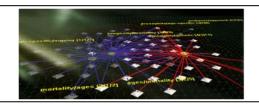
The International Workshop and Conference on Network Science will bring together leading researchers and practitioners in network science - analysts, modeling experts, and visualization specialists with graduate students from many different research areas for interdisciplinary communication and collaboration.

The primary objective of the Workshop/Conference is to facilitate interactions between social and behavioral scientists and the many other disciplines interested in and utilizing network science.

The event will be held over a two week period at Indiana University, Bloomington, Indiana, during May 2006. The first week, the Workshop, will feature tutorials (which present basic, educational material) focusing on a variety of network science research areas. It aims to present and support experimental, theoretical and applied network research by educating the research community on standard network data, tools, and powerful computational resources. The Conference comprises talks by social and behavioral scientists, information scientists, biologists, statistical physicists, mathematicians and statisticians.

Extra Slides

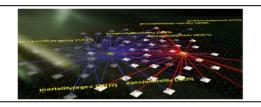
(not covered in the talk)



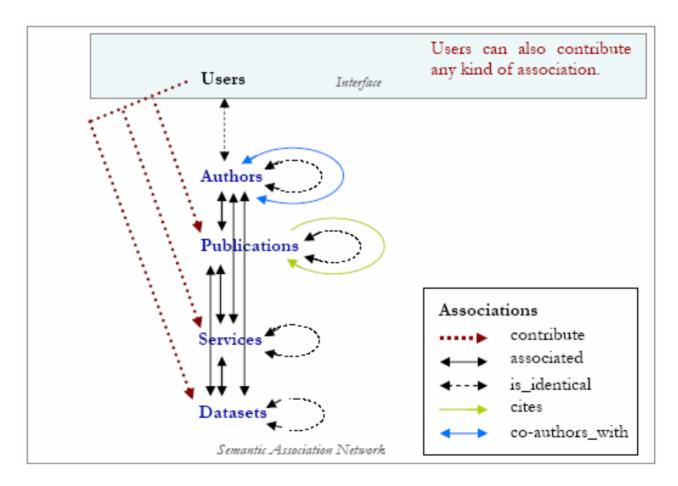
Process of Mapping Knowledge Domains

DATA EXTRACTION	UNIT OF ANALYSIS	MEASURES	LAYOUT (often one code does both similarity and ordination steps)		DISPLAY
		•	SIMILARITY	ORDINATION	
SEARCHES ISI	COMMON CHOICES	COUNTS/FREQUENCIES Attributes (e.g. terms)	SCALAR (unit by unit matrix) Direct citation	DIMENSIONALITY REDUCTION Eigenvector/ Eigenvalue solutions	INTERACTION Browse
INSPEC	Journal De autre de la contraction de la contrac	Author citations	Co-citation	Factor Analysis (FA) and	Pan Za awa
Eng Index Medline ResearchIndex	Document Author Term	Co-citations By γear	Combined linkage Co-word / co-term Co-dassification	Principal Components Analysis (PCA) Multi-dimensional scaling (MDS) LSA , Topics	Zoom Filter Query
Patents		THRESHOLDS		Pathfinder networks (PFNet)	Detail on demand
etc.		Bγ counts	VECTOR (unit by attribute matrix) Vector space model (words/terms)	Self-organizing maps (SOM) includes SOM, ET-maps, etc.	ANALYSIS
BROADENING By citation By terms			Latent Semantic Analysis (words/terms) incl. Singular Value Decomp (SVD)	CLUSTER ANALYSIS	
·			CORRELATION (if desired)	SCALAR	
			Pearson's R on any of above	Triangulation Force-directed placement (FDP)	

Börner, Chen & Boyack.. (2003) Visualizing Knowledge Domains. In Blaise Cronin (Ed.), Annual Review of Information Science & Technology, Volume 37, Medford, NJ: Information Today, Inc./American Society for Information Science and Technology, chapter 5, pp. 179-255.



Semantic Association Networks



Katy Börner. (2006) Semantic Association Networks: Using Semantic Web Technology to Improve Scholarly Knowledge and Expertise Management. In Vladimir Geroimenko & Chaomei Chen (eds.) Visualizing the Semantic Web, Springer Verlag, 2nd Edition, chapter 11, pp. 183-198.