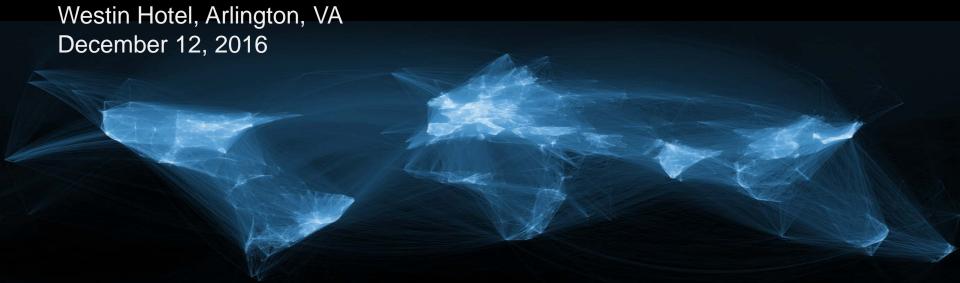
Visualizing Nanoscience and Technology

Katy Börner (@katycns)
Victor H. Yngve Distinguished Professor of Information Science
Director, Cyberinfrastructure for Network Science Center
Department of Intelligent Systems Enegineering
School of Informatics and Computing and IU Network Science Institute
Indiana University, Bloomington, USA

2016 NSF Nanoscale Science and Engineering Grantees Conference





The Structure of Science

The Social Sciences are the smallest and most diffuse of all the sciences. Psychology serves as the link between Medical Sciences (Psychiatry) and the Social Sciences. Statistics serves as the link with Computer Science and Mathematics.

is gradual.

Mathematics is our starting point, the purest of all sciences. It lies at the outer edge of the map. Computer Science, Electrical Engineering, and Optics are applied sciences that draw upon knowledge in Mathematics and Physics. These three disciplines provide a good example of a linear progression from one pure science (Mathematics) to another (Physics) through multiple disciplines. Atthough applied, these disciplines are highly concentrated with distinct bands of research communities that link them. Bands indicate interdisciplinary research.

Research is highly concentrated in Pl These disciplines have few, but very distinct, bands of research communities that link them. The thickness of these bands indicates an extensive amount of interdisciplinary research, which suggests that the boundaries between Physics and Chemistry are not as distinct as one Biochemistry The Life Sciences, including Biology and Biochemistry, are less concentrated than Chemistry or Physics. Bands of linking research can be seen between the larger areas in the Life Sciences; for instance between Biology and Microbiology, and The Medical Sciences include broad therapeutic studies and targeted areas of Treatment (e.g. central between Biology and Environmental Science nervous system, cardiology, gastroenterology, etc.) Biochemistry is very interesting in that it Unlike Physics and Chemistry, the medical disciplines are more spread out, suggesting a more multiis a large discipline that has visible links to disciplines in many areas of the map, disciplinary approach to research. The transition into including Biology, Chemistry, Neuroscience, and General Medicine. It is perhaps the Life Sciences (via Animal Science and Biochemistry)

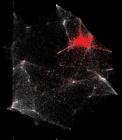
We are all familiar with traditional maps that show the relationships between countries, provinces, states, and cities. Similar relationships exist between the various disciplines and research topics in science. This allows us to map the structure of science.

One of the first maps of science was developed at the Institute for Scientific Information over 30 years ago. It identified 41 areas of science from the citation patterns in 17,000 scientific papers, that early map was intriguing, but it didn't cover enough of science to accurately define its structure.

Things are different today. We have enormous computing power and advanced visualization software that make mapping of the structure of science possible. This galaxy-like map of science (left) was generated at Sandia National Laboratories using an advanced graph layout routine (VxOrd) from the citation patterns in 800,000 scientific papers published in 2002. Each dot in the galaxy represents one of the 96,000 research communities active in science in 2002. A research community is a group of papers (9 on average) that are written on the same research topic in a given year. Over time, communities can be born, continue, split, merge, or die.

The map of science can be used as a tool for science strategy. This is the terrain in which organizations and institutions locate their scientific capabilities. Additional information about the scientific and economic impact of each research community allows policy makers to decide which areas to explore, exploit, abandon, or ignore.

We also envision the map as an educational tool. For children, the theoretical relationship between areas of science can be replaced with a concrete map showing how math, physics, chemistry, biology and social studies interact. For advanced students, areas of interest can be located and neighboring areas can be explored.



Nanotechnology

Most research communities in nanotechnology are concentrated in Physics, Chemistry, and Materials Science. However, many disciplines in the Life and Medical Sciences also have nanotechnology applications.



Proteomics

Research communities in proteomics are centered in Biochemistry. In addition, there is a heavy focus in the tools section of chemistry, such as Chromatography. The balance of the proteomics communities are widely dispersed among the Life and Medical Sciences.



Pharmacogenomics

Pharmacogenomics is a relatively new field with most of its activity in Medicine. It also has many communities in Biochemistry and two communities in the Social Sciences.

most interdisciplinary of the sciences.

The EMERGENCE of NANOTECHNOLOGY **MAPPING THE NANO REVOLUTION** The emergence of nanotechnology has been one of the major scientific-technological revolutions in the last decade and it led to a structural reorganization of major fields of science. Price (1965) showed that fields of science and their development can be mapped using aggregated citations among the journals in the fields and their relevant environments. The frames to the right show the evolving journal citation network for the years 1998-2003. Distances are proportional to cosine values between the citation patterns of the respective 1998 The journal Science interfaces journals. Textual descriptions of key events with relevant journals in both During the period 1996-2000, during the development of Nanotechnology sets: chemistry and applied the journal Nanotechnology are given below each frame. Most notably, physics. Nanotechnology is part of a group of journals leading papers in Science and Nature catalyzed the breakthrough around 2000. emerges as core journal. in applied physics. Increasingly, chemistry journals play a role in the citation impact environment of the journal Nanotechnology. The interdisciplinarity of a journal can be measured using betweenness centrality (BC)—journals that occur on many shortest paths between other journals in a network have higher BC value than those that do not. In the LEGEND maps, sizes of nodes are proportional to the betweenness centrality of the respective Science Nature journal in the citation network, pournal in the citation network. From being a specialist journal in applied physics, the journal Nanotechnology obtains a high BC value in the years of the transition, ca. 2001. This is preceded by the "intervention" of Science. After the transition, the new field of nanotechnology is established, new journals 0.8 0.22 0.33 Nanotechnology Nano Letters such as Nano Letters published by the influential American Chemical Society take the lead, and a new specialty structure with low BC value iournals results. **Betweenness Centrality** 2001 The journal Nanotechnology now provides the interface between chemistry and physics. The "intervention" by Science is no longer needed. An animated sequence of this evolution is at: http://www.leydesdorff.net/journals/nanotech. The journal Science is relevant in the References Leydesdorff, L. and T. Schank. 2008. Dynamic citation impact environment, but now functions as one of the specialist journals Animations of Journal Maps: Indicators of Structural in nanotechnology. Nanoscience further Change and Interdisciplinary Developments. Journal of the American Society for Information Science develops as an increasingly integrated network of journals. and Technology, 59(11), 1810-1818. Other journals in nanoscience and technology begin to emerge, and the bridging role of the journal Nanotechnology gradually Design by Michael J. Stamper and Katy Börner Price, Derek J. de Solla (1965). Networks of scientific subsides. Nano Letters and the Journal of Nanoscience and Cyberinfrastructure for Network Science Center | Indiana University

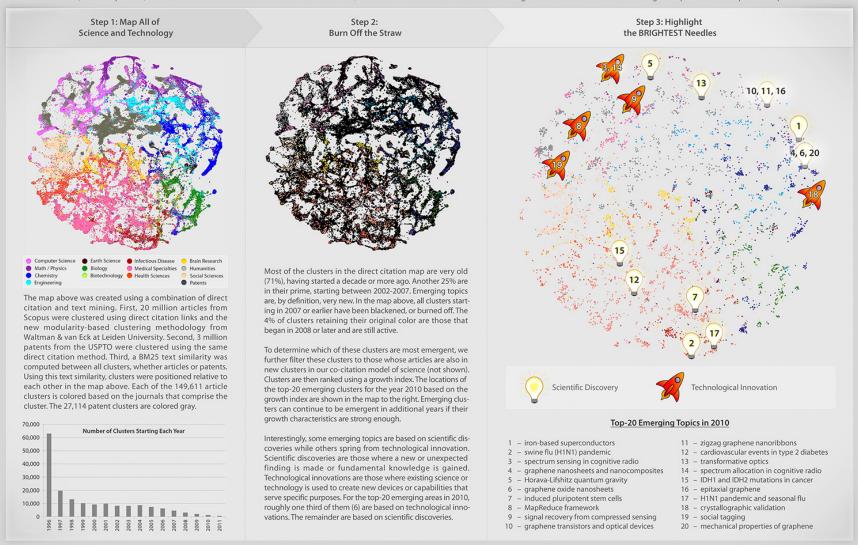
Nanotechnology join the new field of nanotechnology.

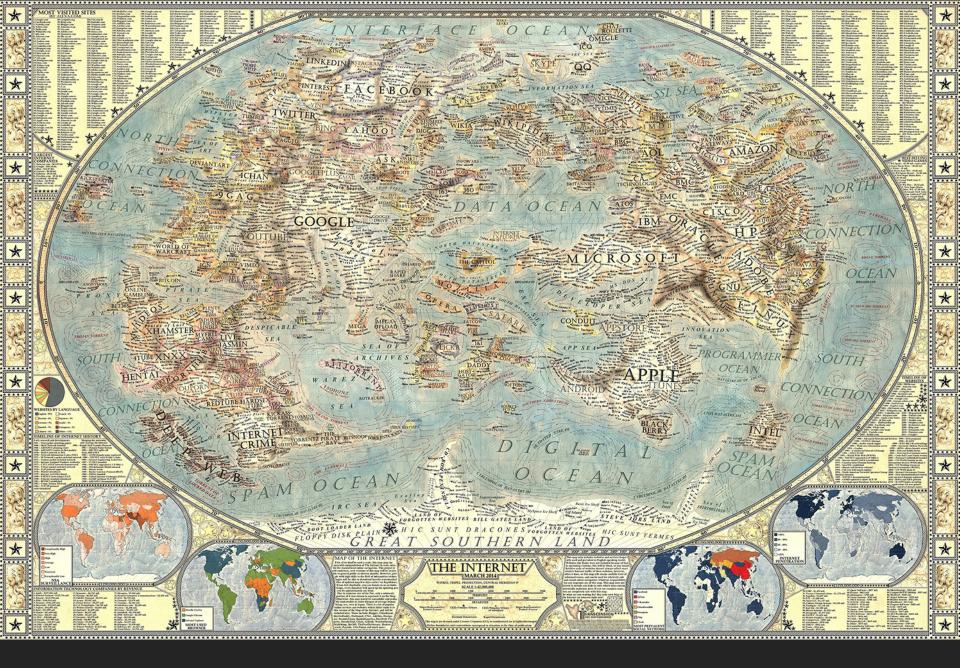
papers. Science, 149, no. 3683, 510-515.

IDENTIFYING EMERGING TOPICS IN SCIENCE AND TECHNOLOGY

(finding the needles in the haystack)

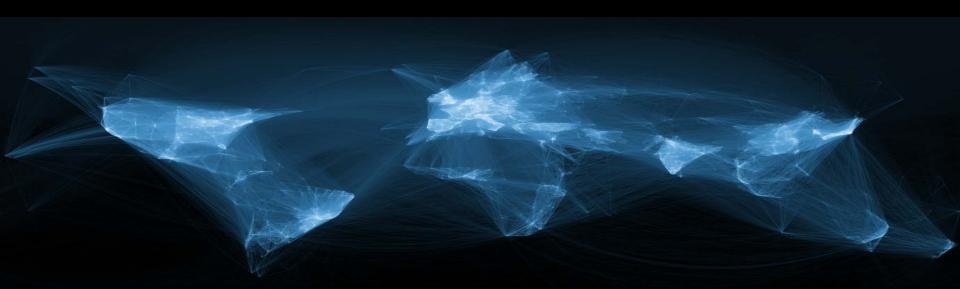
A novel approach to identifying emerging topics in science and technology has been developed. Two models of science and technology have been created using 16 years (1996-2011) of Scopus (20 million articles) and USPTO (3 million patents) data. These two models—one based on direct citation, and one based on co-citation—are used together to nominate the most emergent topics in S&T at a particular point in time.







Increasing Data Visualization Literacy



Data Visualization Literacy

Data visualization literacy (ability to read, make, and explain data visualizations) requires

- literacy (ability to read and write text, e.g., in titles, axis labels, legend),
- visual literacy (ability to find, interpret, evaluate, use, and create images and visual media), and
- data literacy (ability to read, create, and communicate data).



OVERVIEW

Information Visualization MOOC 2017

EVENTS

SCHEDULE

Tweets about ivmooc Lionel Villard Retweeted Katy Borner @katycns Patent cluster maps in support of automatic classification of 0.5M patent applications /year to patent examiner profiles #cdacmtg #ivmooc Patent Classification Data Visualized Information Visualization MOOC

INSTRUCTORS

READINGS

GRADING

FAO

CONTACT

The course can be taken <u>for free</u> or for Indiana University credits as part of the Online Data Science Program or the ILS M.S. program. Register for free at http://ivmooc.cns.iu.edu. Next class starts January 10, 2017.

Tasks

LEVELS

MICRO: Individual Level about 1–1,000 records page 6

MESO: Local Level about 1,001–100,000 records page 8

MACRO: Global Level more than 100,000 records page 10









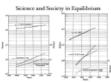
Statistical Analysis page 44



Knowledge Cartography page 135



Productivity of Russian life sciences research teams page 105



Number of scientists versus population and R&D costs versus GNP. page 103





Visualizing decisionmaking processes page 95



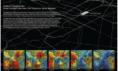
Key events in the development of the video tape recorder page 85



Increased travel and communication speeds page 83







Cell phone usage in Milan, Italy page 109



Victorian poetry in Europe page 137



Ecological footprint of countries page 99





Evolving patent holdings of Apple Computer, Inc. and Jerome Lemelson page 89



Evolving journal networks in nanotechnology page 139



Product space showing co-export patterns of countries page 93







World Finance Corporation network page 87



Electronic and new media art networks page 133



World-wide scholarly collaboration networks page 157

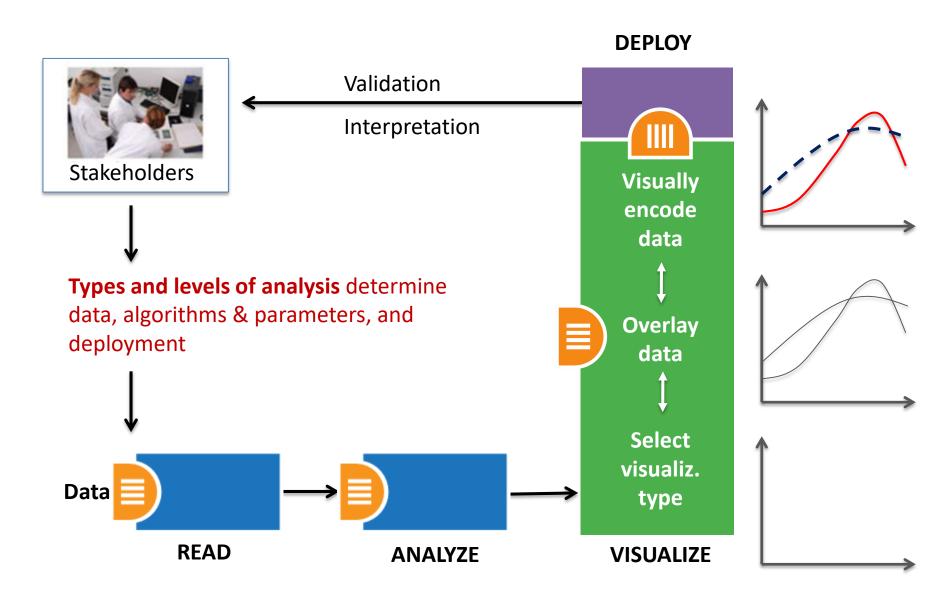


See page 5

11

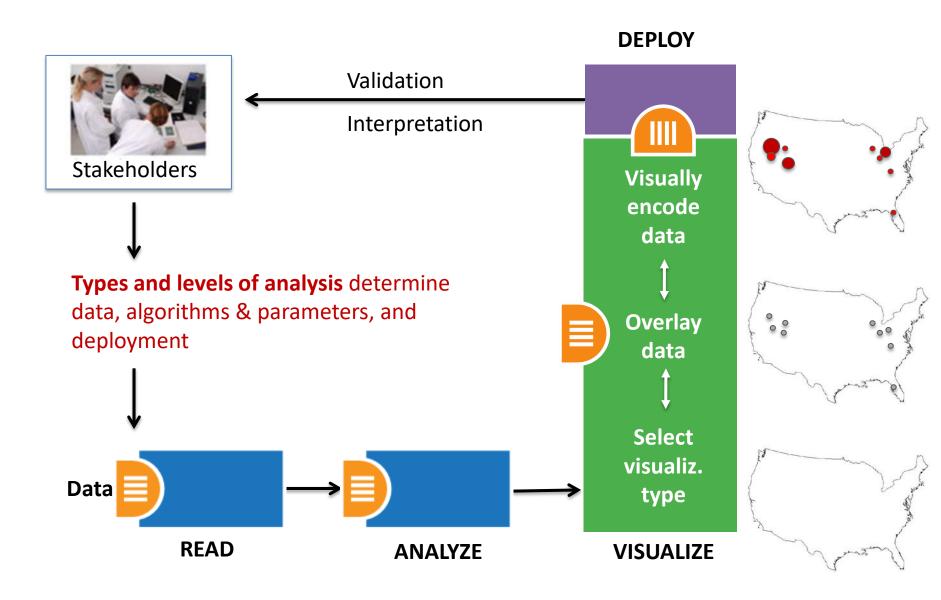


Needs-Driven Workflow Design





Needs-Driven Workflow Design



Course Schedule

Part 1: Theory and Hands-On

- **Session 1** Workflow Design and Visualization Framework
- Session 2 "When:" Temporal Data
- Session 3 "Where:" Geospatial Data
- Session 4 "What:" Topical Data

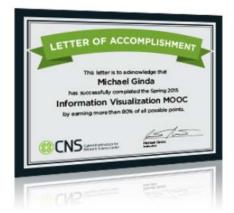
Mid-Term

- Session 5 "With Whom:" Trees
- Session 6 "With Whom:" Networks
- Session 7 Dynamic Visualizations and Deployment

Final Exam

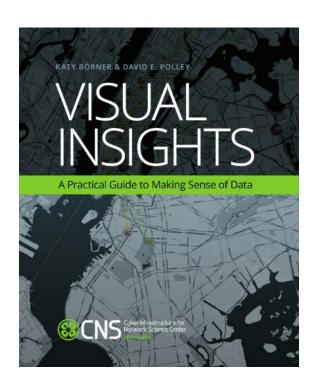
Part 2: Students work in teams on client projects.

Final grade is based on Homework (10%), Midterm (20%), Final Exam (30%), Client Project (30%), and Class Participation (10%).



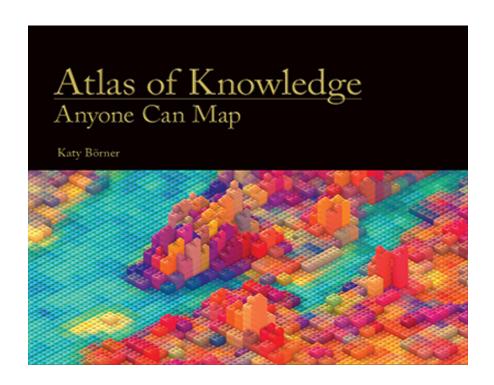


Books Used in the IVMOOC



Teaches timely knowledge:

Advanced algorithms, tools, and hands-on workflows.



Teaches timeless knowledge:

Visualization framework exemplified using generic visualization examples and pioneering visualizations.

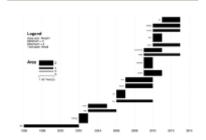
Load **One** File and Run **Many** Analyses and Visualizations

Times Cited	Publication Year	City of Publisher	Country	Journal Title (Full)	Title	Subject Category	Authors
12	2011	NEW YORK	USA	COMMUNICATI ONS OF THE ACM	Plug-and-Play Macroscopes	Computer Science	Borner, K
18	2010	MALDEN	USA	CTS-CLINICAL AND TRANSLATIONA L SCIENCE	Advancing the Science of Team Science	Research & Experimental Medicine	Falk-Krzesinski, HJ Borner, K Contractor, N Fiore, SM Hall, KL Keyton, J Spring, B Stokols, D Trochim, W Uzzi, B
13	2010	WASHINGTON	USA	TRANSLATIONA	A Multi-Level Systems Perspective for the Science of Team Science	Cell Biology Research & Experimental Medicine	Borner, K Contractor, N Falk- Krzesinski, HJ Fiore, SM Hall, KL Keyton, J Spring, B Stokols, D Trochim, W Uzzi, B

Statistical Analysis—p. 44

Location	Count	# Citations
Netherlands	13	292
United States	9	318
Germany	11	36
United Kingdom	1	2

Temporal Burst Analysis—p. 48



Geospatial Analysis—p. 52

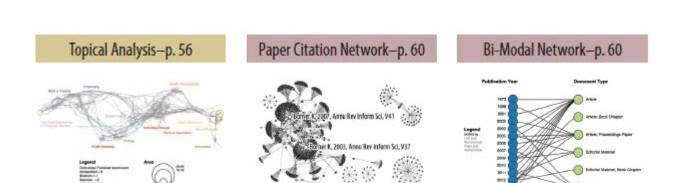


Geospatial Analysis—p. 52



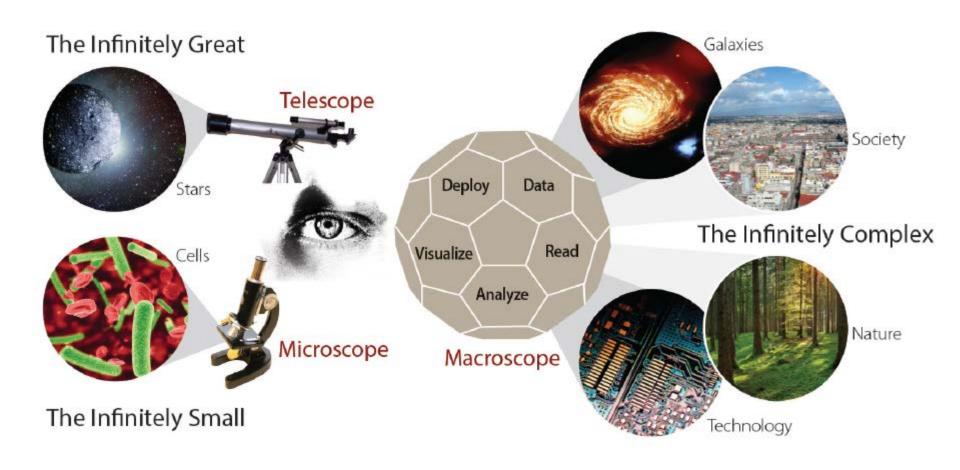
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Co-author and many other bi-modal networks.

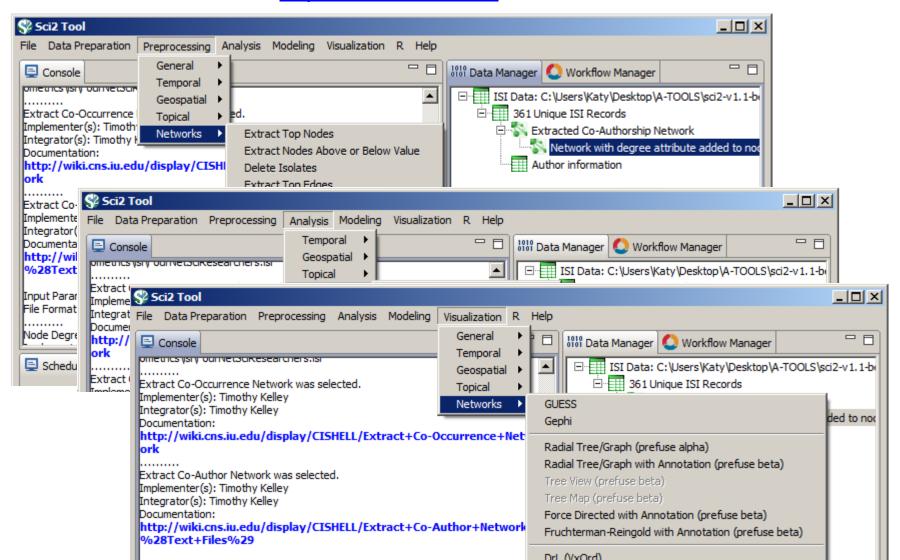
Microscopes, Telescopes, Macroscopes Plug-and-Play Macroscopes





Sci2 Tool Interface Components

Download tool for free at http://sci2.cns.iu.edu





Visualization Framework

Insight Need Types page 26	Data Scale Types page 28	Visualization Types page 30	Graphic Symbol Types page 32	Graphic Variable Types page 34	Interaction Types page 26
 categorize/cluster order/rank/sort distributions (also outliers, gaps) comparisons trends (process and time) geospatial compositions (also of text) correlations/relationships 	nominal ordinal interval ratio	table chart graph map network layout	geometric symbols point line area surface volume linguistic symbols text numerals punctuation marks pictorial symbols images icons statistical glyphs	spatial position retinal form color optics motion	overview zoom search and locate filter details-on-demand history extract link and brush projection distortion



See page 24

Visualization Framework

Basic Task Types									
Bertin, 1967	Wehrend & Lewis, 1996	Few, 2004	Yau, 2011	Rendgen & Wiedemann, 2012	Frankel, 2012	Tool: Many Eyes	Tool: Chart Chooser	Börner, 2014	
selection	categorize			category				categorize/ cluster	
order	rank	ranking					table	order/rank/ sort	
	distribution	distribution					distribution	distributions (also outliers, gaps)	
	compare	nominal comparison & deviation	differences		compare and contrast	compare data values	comparison	comparisons	
		time series	patterns over time	time	process and time	track rises and falls over time	trend	trends (process and time)	
		geospatial	spatial relations	location		generate maps		geospatial	
quantity		part-to- whole	proportions		form and structure	see parts of whole, analyze text	composition	compositions (also of text)	
association	correlate	correlation	relationships	hierarchy		relations between data points	relationship	correlations/ relationships	



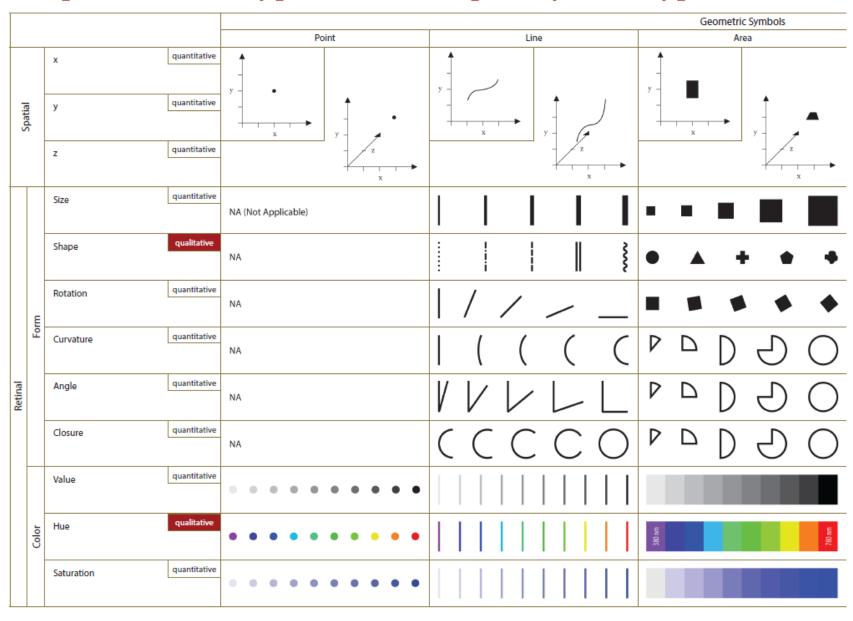
Visualization Framework

Insight Need Types page 26	Data Scale Types page 28	· · ·	Graphic Symbol Types page 32	Graphic Variable Types page 34	Interaction Types page 26
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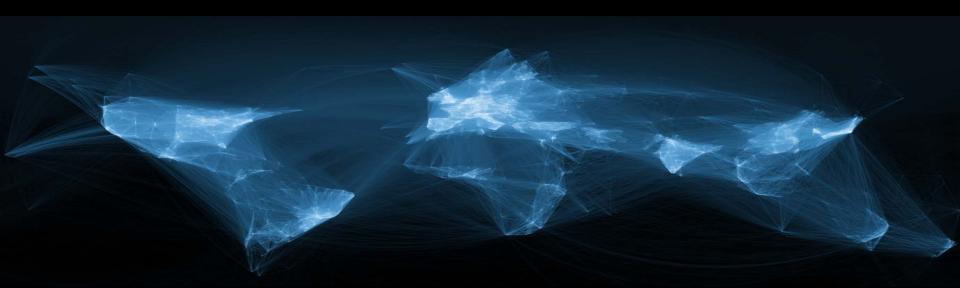
See page 24

Graphic Variable Types Versus Graphic Symbol Types



Graphic Variable Types Versus Graphic Symbol Types Grometric Symbols Linguistic Symbols Fictorial Symbols question quantitative size. NA Nat Applicable Text That curature ome table cells are left blank to encour questions clours substation specing combally **Fallery** quantitative crimitation credient Tampanno stading quantitative steroscopic copfit wixty detim Blinking paint feet. feet:

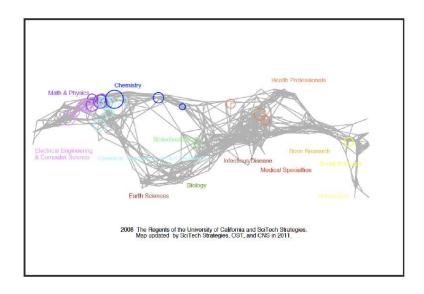
Visualizing Engineering Research Centers

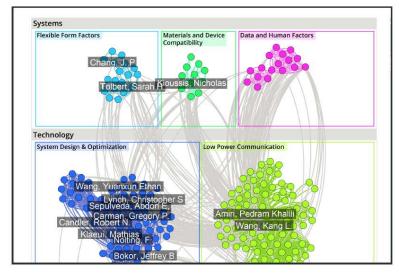


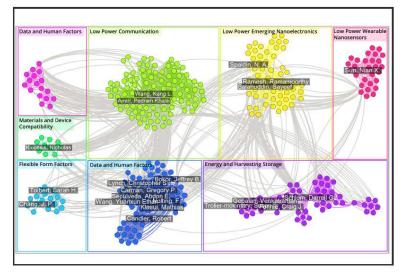
Engineering Research Observatory

CNS Team at ISE, IU and NanoHub team at Purdue U









TANMS Engineering Research Center Evolving Co-Authorship Network

CNS @ Indiana University 2016

Legend

Nodes ~ Authors Node area size ~ Number of papers published Node color ~ Year of first paper

Edges ~ Co-authorship relations Edge thickness ~ Number of times co-authored

Node Color Code



Node Area Size



Edge Thickness



Wang, Kang L.

Amiri, Pediram Khalili
Upadhyaya, Pramey

Alzate, Uuan G

Displayed Year: 2012

TANMS Engineering Research Center Evolving Co-Authorship Network

CNS @ Indiana University 2016

Legend

Nodes ~ Authors Node area size ~ Number of papers published Node color ~ Year of first paper

Edges ~ Co-authorship relations Edge thickness ~ Number of times co-authored

Node Color Code

2012

2013

2015

2016

Node Area Size



Edge Thickness

2

1

Chang, J. P.

Tolbert, Sarah H.

Lynch, Christopher S

Pisani, David M Cui, dizhai Liang, Cheng-yen

Klaeui, Mathias Nolting, F

Schelhaso Laura T. Keller, Scott M.

Hockel, Joshua L.

Nordeen, Paul K

Carman, Gregory P.

WetzlaroKyle P.

Wang Kang L.

Bur, Alexandre

Amiri, Pedram Khalili

Upadhyava, Pramey

Alzate uan G

Dusado Ritika

Garcia, Ephrahim

Udalow, O. G.

Beloborodov, I. S.

Ramesh, Ramamoorthy

Displayed Year: 2013

TANMS Engineering Research Center Evolving Co-Authorship Network

CNS @ Indiana University 2016

Legend

Nodes ~ Authors Node area size ~ Number of papers published Node color ~ Year of first paper

Edges ~ Co-authorship relations Edge thickness ~ Number of times co-authored

Node Color Code



2015

2016

Node Area Size



Edge Thickness



Yaoo,Zhi Wang, Yuamxun Ethan

Chang, J. P.

Carka, Dorinamaria

Lynch, Christopher S

Pisani, David M Liang, Cheng-yen

Sepulvedan Abdon E.

Klaeui, Mathias

Nolting, F

Sun, Wei-yang Schelhas Laura T.

Cui, Jizhai

Keller, Scott M.

Tolbert, Sarah H.

Hockel, Joshua L.

Nordeen, Paul K

Carman, Gregory P.

Wetzlan Kyle P.

OngoP, V

Valdovinos, John

Bur, Alexandre

Kioussis Nicholas

Amiri, Pedram Khalili

Upadhyaya) Pramey

Li, Xjang

Wang, Kang L

Yu, Goqiang Alzate, Juan G

Dusado Ritika

Salahuddin, Sayeef

Schlom, Darrell G.

Ramesh, Ramamoorthy

Heron J. T.

Gopalan, Venkatraman

Marti, X

Displayed Year: 2014

Fennie Craig J

Garcia, Ephrahim

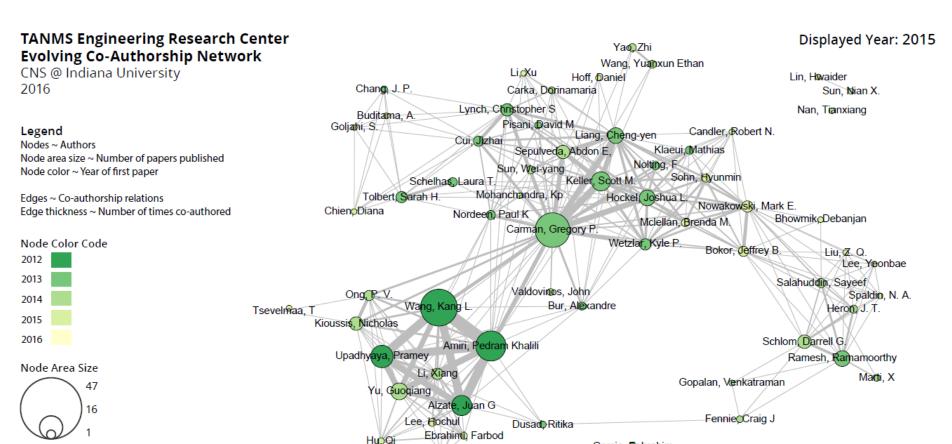
Udalov, O. G.

Beloborodov, I. S.

Youssefp George

Chavez, Andres Cornel

Jaco, Stephanie Marie



Grezes Cecile

29

14

Edge Thickness

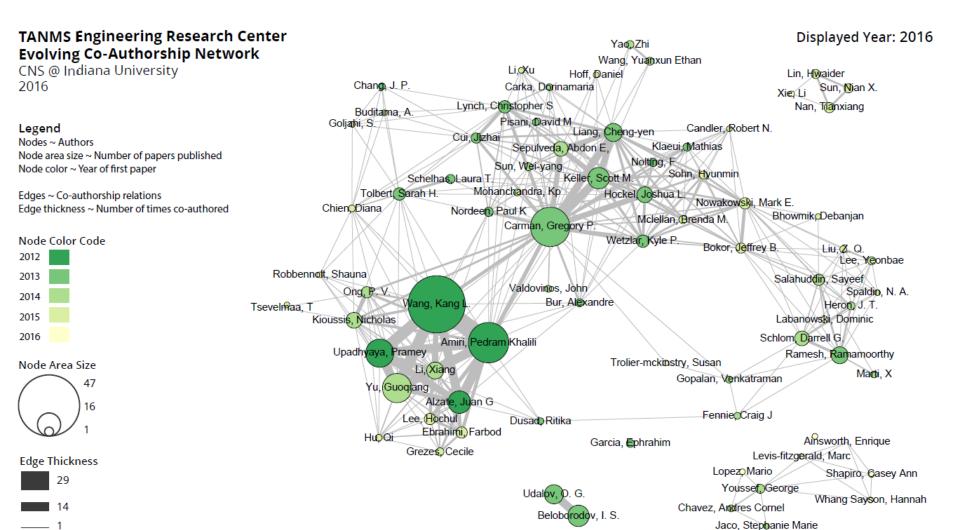
1 Beloborado

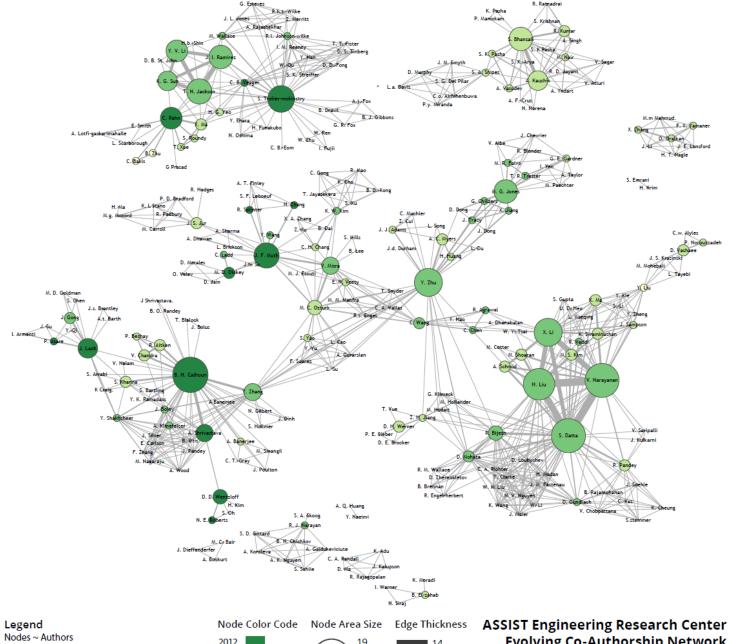
Udalov, O. G.

Beloborodov, I. S.

Youssef George
Chavez, Andres Cornel
Jaco, Stepbanie Marie

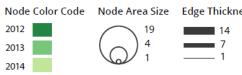
Garcia, Ephrahim





Nodes ~ Authors Node area size ~ Number of papers published Node color ~ Year of first paper

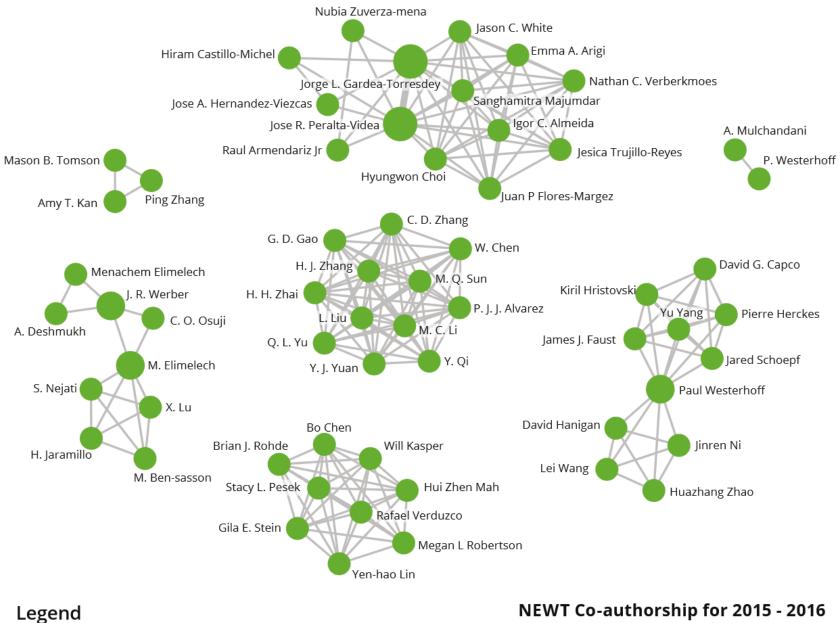
Edges ~ Co-authorship relations Edge thickness ~ Number of times co-authored



2015

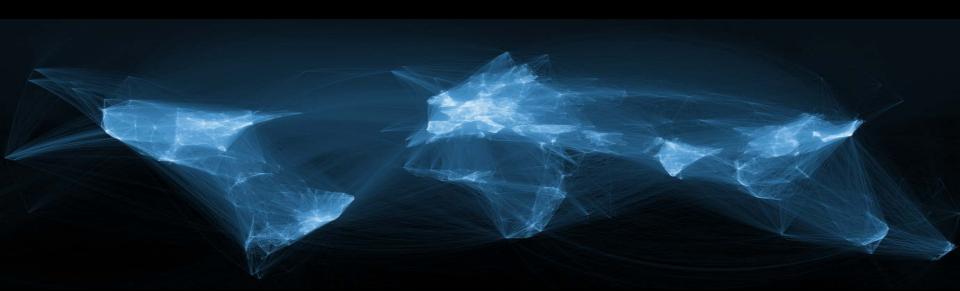
Displayed Year: 2015

SIST Engineering Research Center Evolving Co-Authorship Network CNS @ Indiana University 2016



Node area size Node area size Edge weight Number of papers published Number of papers published Number of papers co-authored

Modelling S&T Developments













Government, academic, and industry leaders discussed challenges and opportunities associated with using big data, visual analytics, and computational models in STI decision-making.

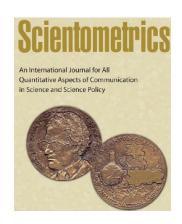
Conference slides, recordings, and report are available via http://modsti.cns.iu.edu/report



Special Issue of *Scientometrics*:

Simulating the Processes of Science, Technology, and Innovation

Bruce Edmonds, Andrea Scharnhorst, Katy Börner & Staša Milojević (Editors)



- Rogier De Langhe: Towards the Discovery of Scientific Revolutions in Scientometric Data
- Sabine Brunswicker, Sorin Matei, Michael Zentner, Lynn Zentner and Gerhard Klimeck: Creating Impact in the Digital Space: Digital Practice Dependency in Scientific Developer Communities
- Johan Bollen et al.: An Efficient System to Fund Science: From Proposal Review to Peer-to-Peer Distributions
- Petra Ahrweiler: Agent-based Simulation for Science, Technology and Innovation Policy
- David Chavalarias: What's Wrong With Science? Modeling Collective Discovery Processes With the Nobel Game
- Jeff Alstott, Giorgio Triulzi, Bowen Yan and Jianxi Luo: Mapping Technology
 Space by Normalizing Patent Technology Networks

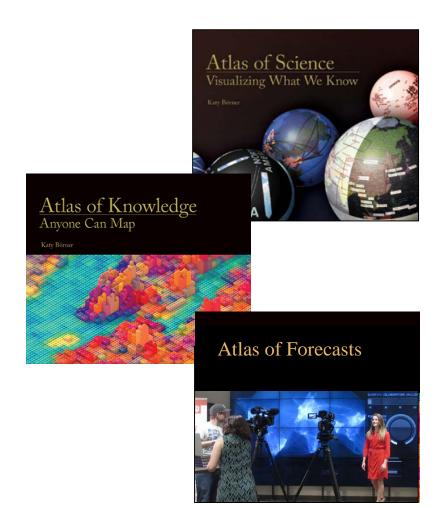
Atlas Trilogy

Börner, Katy (2010) **Atlas of Science: Visualizing What We Know**. The MIT Press. http://scimaps.org/atlas

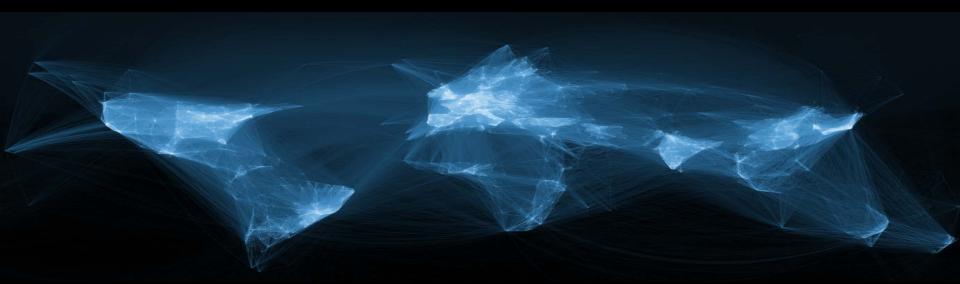
Börner, Katy (2015) **Atlas of Knowledge: Anyone Can Map**. The MIT Press. http://scimaps.org/atlas2

Börner, Katy (2020) Atlas of Forecasts: Predicting and Broadcasting Science, Technology, and Innovation. The MIT Press.

Upcoming Sackler Colloquium on "Modelling and Visualizing Science and Technology Developments" will take place in December 2017 at the Beckman Center, Irvine, CA.



Communicating S&T Development to Different Audiences





Places & Spaces: Mapping Science Exhibit, online at http://scimaps.org



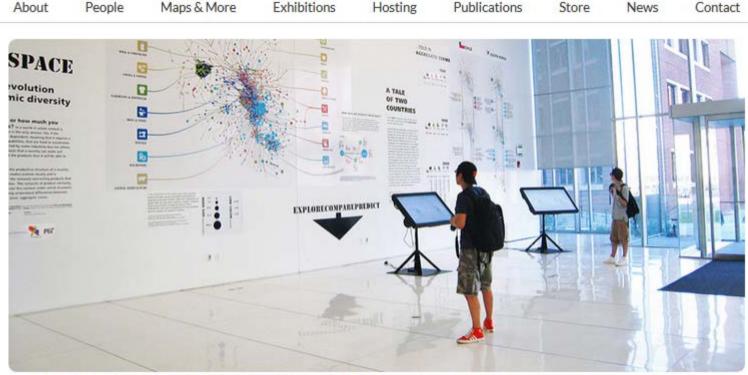
search scimaps.org

Search









Hidalgo, César A., Bailey Klinger, Albert-László Barabási, and Ricardo Hausmann. 2007. See also The Product Space map from Phase I of Places & Spaces.

Call for Macroscope Tools for the *Places & Spaces: Mapping Science* Exhibit (2017) http://scimaps.org/call

Background and Goals

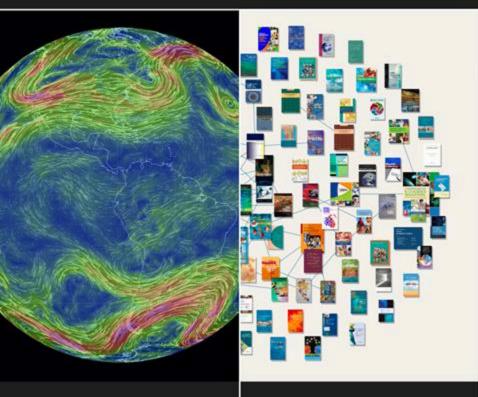
The Places & Spaces: Mapping Science exhibit was created to in communicate human activity and scientific progress on a glot that enable the close inspection of large-scale maps in public conferences; (2) novel, interactive macroscope tools that let it

Themes for the upcoming iterations/years are:

- · 11th Iteration (2015): Macroscopes for Interacting With Science
- 12th Iteration (2016): Macroscopes for Making Sense of Science
- 13th Iteration (2017): Macroscopes for Forecasting Science
- 14th Iteration (2018): Macroscopes for Economic Decision Makers
- 15th Iteration (2019): Macroscopes for Science Policy Makers

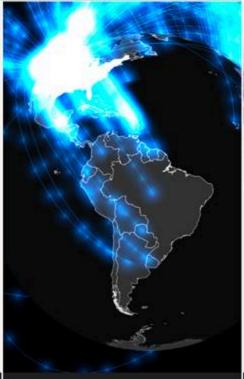
① MACROSCOPES FOR INTERACTING WITH SCIENCE





Earth

AcademyScope

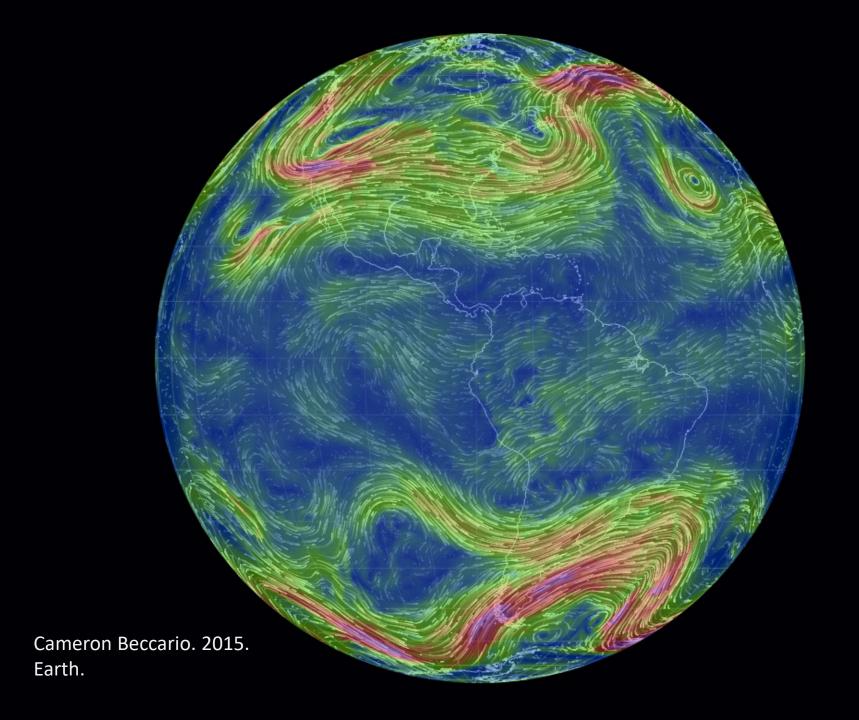


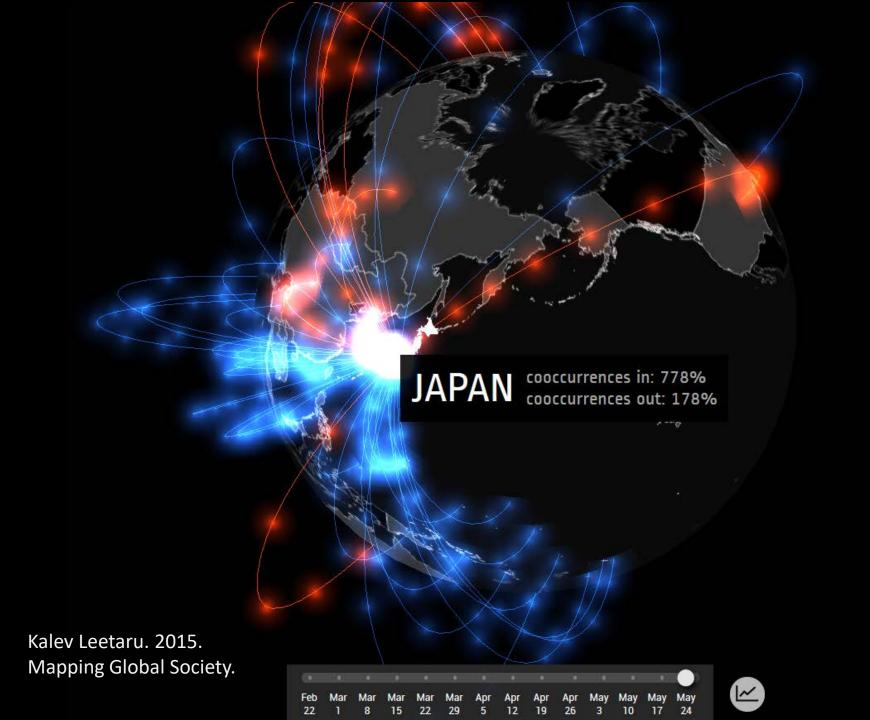
Mapping Global Society



Charting Culture

http://scimaps.org/iteration/11







Places & Spaces Exhibit at the David J. Sencer CDC Museum, Atlanta, GA January 25-June 17, 2016

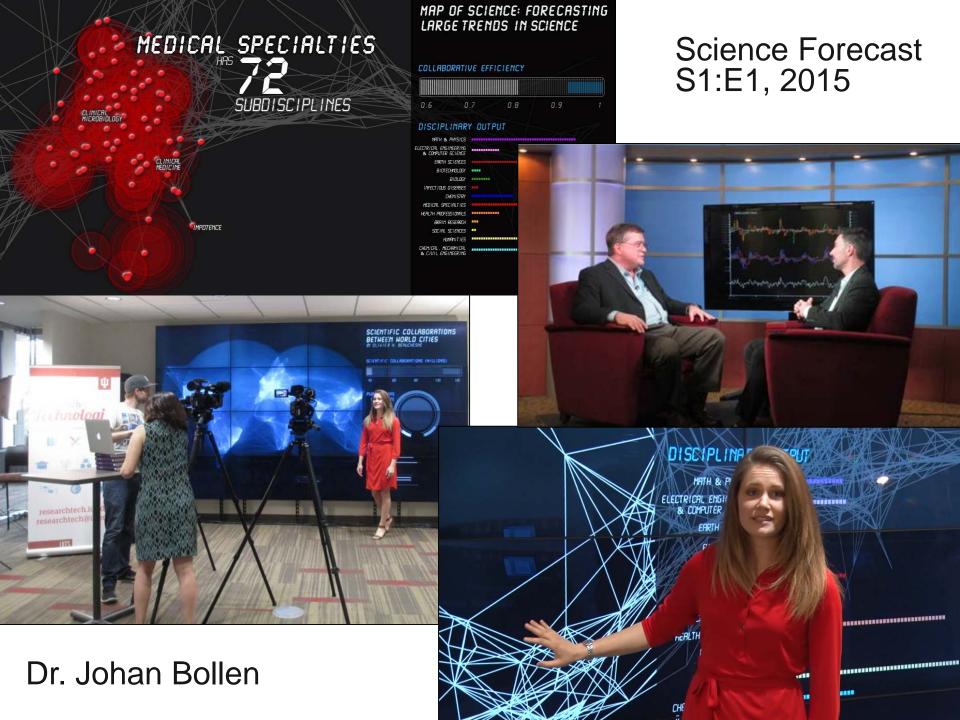
Seeing for
Action - Using
Maps and
Graphs
to Protect the
Public's Health.



Maps of Health Exhibit

David J. Sencer CDC Museum Atlanta, GA

Jan 25-Jun 17, 2016





Dr. Rob Porter





All papers, maps, tools, talks, press are linked from http://cns.iu.edu/presentations
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