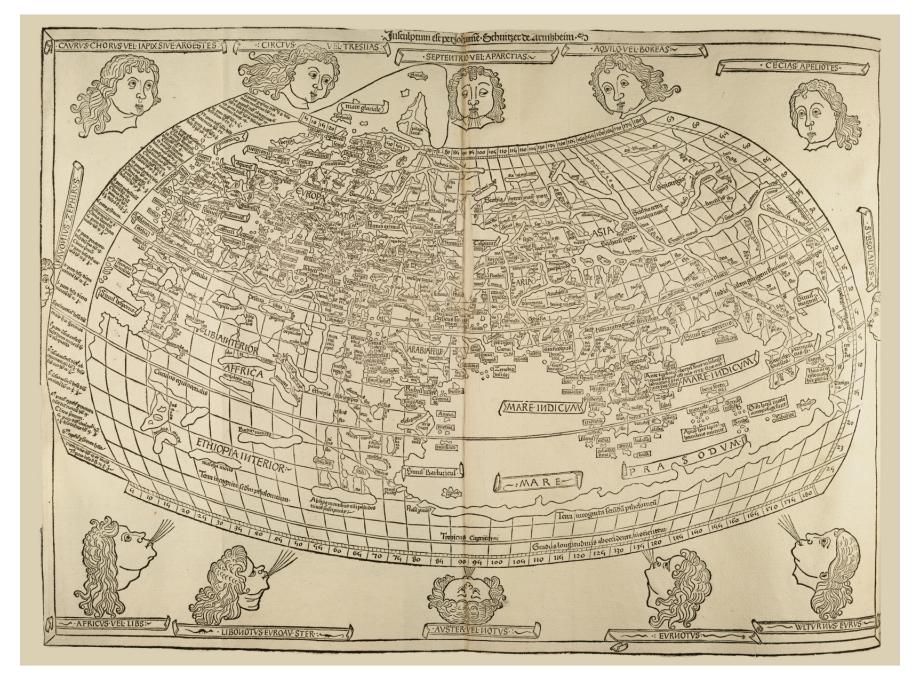
How to NOT Lie with Maps

Katy Börner @katycns

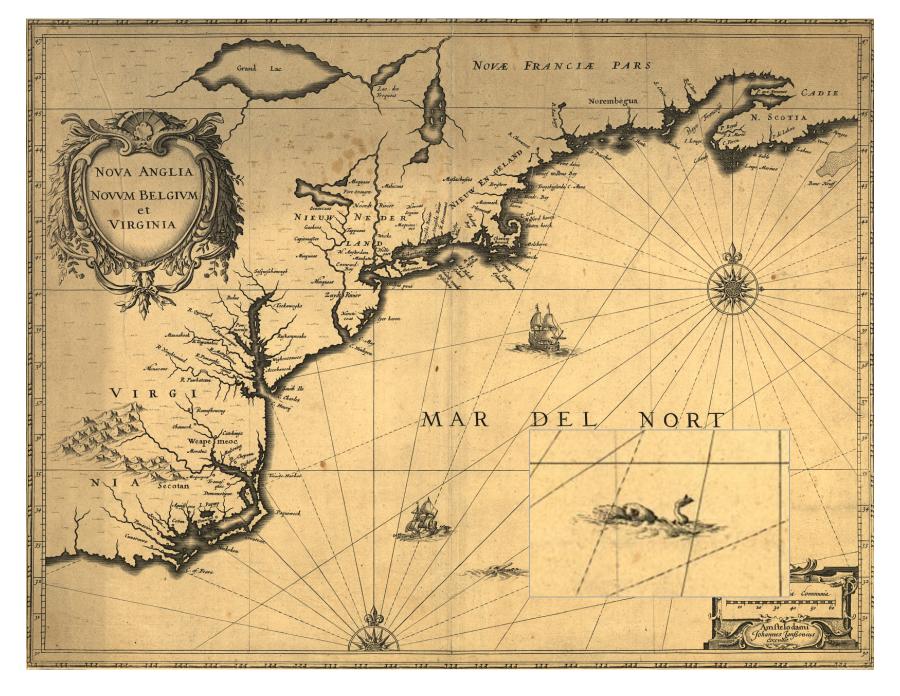
Victor H. Yngve Distinguished Professor of
Intelligent Systems Engineering & Information Science
Director, Cyberinfrastructure for Network Science Center
School of Informatics and Computing
Indiana University Network Science Institute
Indiana University, USA

Open science, dark knowledge: science in an age of ignorance"
2017 Technology Symposium

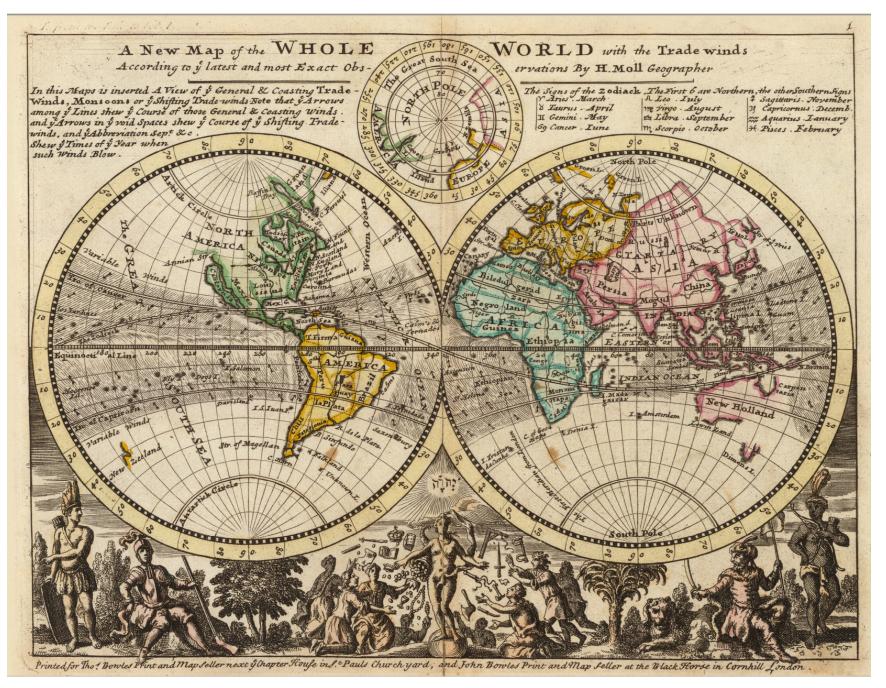
Alpbach, Austria | August 25, 2017



I.1 Cosmographia World Map - Claudius Ptolemy - 1482



1.2 Nova Anglia, Novvm Belgivm et Virginia – Johannes Janssonius - 1642



1.3 A New Map of the Whole World with Trade Winds According to the Latest and Most Exact Observations - Herman Moll - 1736



the case of this map, Science & Technology Outlook 2005-2055, the terrain we're navigating is the uncharted territory of science and technology [S&T] in the next 50 years. However, the map of the future is not a tool for prediction or, for that matter, the product of predictions. Nor is it comparable to modern navigation techniques in which we rely on a shrinking number of strong signals, like GPS coordinates, to show the right path. Rather, it's more akin to classical low-tech navigational techniques with their reliance on an array of weak signals such as wind direction, the look and feel of the water, and the shape of cloud formations. Taken together, these signals often prove more useful for navigation than high-tech methods because, in addition to aiding travelers in selecting the "right" path. the signals contextualize information and reveal interdependencies and connections between seemingly unrelated events, thus enriching our understanding of the landscape. That's precisely the intention of this map of the future of S&T-to give the reader a deeper contextual understanding of the landscape and to point to the intricacies and interdependencies between

A map is a tool for navigating an unknown terrain. In

While developing the map, the Institute for the Future (IFTF) team listened for and connected a variety of weak signals, including those generated during interviews and workshop conversations involving more than 100 eminent U.K. and U.S. experts in S&T-academicians, policymakers, journalists, and corporate researchers. The IFTF team also compiled a database of outlooks on developments that are likely to impact the full range of S&T disciplines and practice areas over the next 50 years. We also relied on IFTF's 40 years of experience in forecasting S&T developments to create the map and an accompanying set of S&T Perspectives that discuss issues emerging on the S&T horizon and are important for organizations, policymakers, and society-at-large to understand.

On this map, six themes are woven together across the 50-year horizon, often resulting in important breakthroughs. These are supported by key technolgies, innovations, and discoveries. In addition to the six themes, three meta-themes-democratized innovation, transdisciplinarity, and emergence-will overlay the future S&T landscape influencing how we think about, learn about, and practice science. Finally, S&T trends won't operate in a vacuum. Wider social, demographic, political, economic, and environmental trends will both influence S&T trends and will be influenced by them. Some of these wider trends surround the map to remind us of the larger picture.

MAP THEMES

After 20 years of basic research and development at the 100nanometer scale, the importance of nanotechnology as a source of innovations and new capabilities in everything from materials science to medicine is already well-understood. Three trends, however, will define how nanotechnology will unfold, and what impacts it will have. First, nanotechnology is not a single field with a coherent intellectual program; it's an opportunistic hybrid, shaped by a combination of fundamental research questions, promising technical applications, and venture and state capital. Second, nanotechnology is moving away from the original vision of small-scale mechanical engineering-in which assemblers build mechanical systems from individual atoms-toward one in which molecular biology and biochemistry contribute essential tools (such as proteins that build nanowires). Finally, nanotechnology will also serve as a model for transdisciplinary science. It will support both fundamental research and commercially oriented innovation: and it will be conducted not within the boundaries of conventional academic or corporate research departments, but in institutional and social milieux that emphasize heterogeneity.

Intentional Biology

For 3.6 billion years, evolution has governed biology on this planet. But today, Mother Nature has a collaborator. Inexpensive tools to read and rewrite the genetic code of life will bootstrap our ability to manipulate biology from the bottom up. We'll not only genetically reengineer existing life but actually create new life forms with purpose. Still, we will not be blind to what nature has to teach us. Evolution's elegant engineering at the smallest scales will be a rich source of inspiration as we build the bio-nanotechnology of the next 50 years.

In the next 50 years, we will be faced with broad opportunities to remake our minds and bodies in profoundly different ways. Advances in biotechnology, brain science, information technology, and robotics will result in an array of methods to dramatically alter, enhance, and extend the mental and physical hand that nature has dealt us. Wielding these tools on ourselves, humans will begin to define a variety of different "transhumanist" paths—that is, ways of being and living that extend beyond what we today consider natural for our species. In the very long term, following these paths could someday lead to an evolutionary leap for humanity.

Mathematical World

The ability to process, manipulate, and ultimately understand patterns in enormous amounts of data will allow decoding of previously mysterious processes in everything from biological to social systems. Scientists are learning that at the core of many biological phenomena-reproduction, growth, repair, and others-are computational processes that can be decoded and simulated. Using techniques of combinatorial science to uncover such patterns whether these are physical, biological, or social-will likely occupy an increasing share of computing cycles in the next 50 years. Such massive computation will also make simulation widespread. Computer simulation will be used not only to help make decisions about large complex scientific and social problems but also to help individuals make better choices in their daily lives.

Sensory Transformation

In the next ten years, physical objects, places, and even human beings themselves will increasingly become embedded with computational devices that can sense, understand, and act upon their environment. They will be able to react to contextual clues about the physical, social, and even emotional state of people and things in their surroundings. As a result, increasing demands will be placed on our visual, auditory, and other sensory abilities. Information previously encoded as text and numbers will be displayed in richer sensory formats—as graphics, pictures, patterns, sounds, smells, and tactile experiences. This enriched sensory environment will coincide with major breakthroughs in our understanding of the brain-in how we process sensory information and connect various sensory functions.

Humans will become much more sophisticated in their ability to understand, create, and manage sensory information and ability to perform such tasks will become keys to success.

Lightweight Infrastructure

A confluence of new materials and distributed intelligence is pointing the way toward a new kind of infrastructure that will dramatically reshape the economics of moving people, goods, energy, and information. From the molecular level to the macroeconomic level, these new infrastructure designs will emphasize smaller, smarter, more independent components. These components will be organized into more efficient, more flexible, and more secure ways than the capital-intensive networks of the 20th century. These lightweight infrastructures have the potential to boost emerging economies, improve social connectivity, mitigate the environmental impacts of rapid global urbanization, and offer new future paths in energy.

META-THEMES

Democratized Innovation

Before the 20th century, many of the greatest scientific discoveries and technical inventions were made by amateur scientists and independent inventors. In the last 100 years, a professional class of scientists and engineers, supported by universities, industry, and the state, pushed amateurs aside as a creative force. At the national scale, the capital-intensive character of scientific research made world-class research the property of prosperous advanced nations. In the new century, a number of trends and technologies will lower the barriers to participation in science and technology again, both for individuals and for emerging countries. The result with be a renaissance of the serious amateurs, the growth of new scientific and technical centres of excellence in developing countries, and a more global distribution of worldclass scientists and technologists.

● Transdisciplinarity

In the last two centuries, natural philosophy and natural history fractured into the now-familiar disciplines of physics, chemistry, biology, and so on. The sciences evolved into their current form in response to intellectual and professional opportunities, philanthropic priorities, and economic and state needs. Through most of the 20th century, the growth of the sciences, and academic and career pressures, encouraged ever-greater specialization. In the coming decades, transdisciplinary research will become an imperative. According to Howard Rheingold, a prominent forecaster and author, "transdisciplinarity goes beyond bringing together researchers from different disciplines to work in multidisciplinary teams. It means educating researchers who can speak languages of multiple disciplines-biologists who have understanding of mathematics, mathematicians who understand biology.

The phenomenon of self-organizing swarms that generate complex behavior by following simple rules—will likely become an important research area, and an important model for understanding how the natural world works and how artificial worlds can be designed. Emergent phenomena have been observed across a variety of natural phenomena, from physics to biology to sociology. The concept has broad appeal due to the diversity of fields and problems to which it can be applied. It is proving useful for making sense of a very wide range of phenomena. Meanwhile, emergence can be modeled using relatively simple computational tools, although those models often require substantial processing power. More generally, it is a richly suggestive as a way of thinking about designing complex, robust technological systems. Finally, emergence is an accessible and vivid a metaphor for understanding nature. Just as classical physics profited from popular treatments of Newtonian mechanics, so too will scientific study and technical reproductions of emergent phenomena likely draw benefits from the popularization of its underlying concepts.

Examining the Evolution & Distribution of Patent Classifications

Managing Growing Patent Portfolios

Organizations, businesses, and individuals rely on patents to protect their intellectual property and business models. As market competition increases, patenting innovation and intellectual property rights becomes ever more important.

Managing the staggering number of patents demands new tools and methodologies. Grouping patents by their classifications offers an ideal resolution for better understanding how intellectual borders are established and change over time.

The charts below show the annual number of patents granted from January 1, 1976 to December 31, 2002 in the United States Patent and Trademark Office (USPTO) patent archive; slow and fast growing patent classes; the top 10 fast growing patent subclasses; and two evolving patent portfolios.

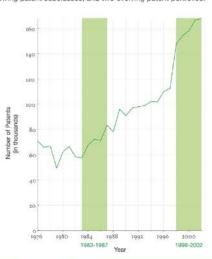
The Structure and Evolution of the Patent Space

The United States Patent and Trademark Office assigns each patent to one of more than 450 classes covering broad application domains. For example, class 514 encompasses all patents dealing with 'Drug, Bio-Affecting and Body Treating Compositions.' Classes are further broken down by subclasses that have hierarchical associations. As one example, class 455 features subclass 99 entitled "with vehicle."

The top 10 fast growing patent classes for 1998–2002 are listed together with the number of patents granted. Most come from the 'Computer and Communications' and the 'Drugs and Medical' area.

The evolving hierarchical structure of patent classes and their sizes is represented using treemaps, a space-filling visualization technique developed by Ben Shneiderman at the University of Maryland. A treemap presents a hierarchy as a collection of nested rectangles—demarcating a parent-child relationship between nodes by nesting the child within the parent rectangle. The size and color of each rectangle represent certain attributes of the nodes.

Here, each rectangle represents a class and the area size denotes the total number of patents in that class. The rectangle's color corresponds to percentage increase (green) or decrease (red) in the number of patents granted in that class from the previous interval.





Top-10 Subclasses

Class	Title	# of Patents
514	Drug, Bio-Affecting and Body Treating Compositions	18,778
438	Semiconductor Device Manufacturing:Process	17,775
435	Chemistry: Molecular Biology and Microbiology	17,474
424	Drug, Bio-Affecting and Body Treating Compositions	13,637
428	Stock Material or Miscellaneous Articles	13,314
257	Active Solid-State Devices (e.g., Transistors, Solid-State Diodes)	12,924
395	Information Processing System Organization	9,955
345	Computer Graphics Processing, Operator Interface Processing, and Selective Visual Display Systems	9,510
359	Optical: Systems and Elements	9,151
365	Static Information Storage and Retrieval	8,392
	Total	130,910

1998-2002

Patent Portfolio Analysis

A longitudinal analysis of portfolios reveals different patenting strategies. For each year (given in gray above each treemap), a treemap of all new patents granted to the assignee is shown. The number of patents is given below each treemap. The same size and color coding as above was used. In addition, proceeding the process of the patents of the p

Apple Computer, Inc.

Apple Computer, Inc.'s portfolio starts in 1980 and increases considerably in size over time. In most years, more than half of Apple Computer's patent filings were placed into four classes, namely '395 Information Processing System Organization,' '345 Computer Graphics Processing, Operator Interface Processing, and Selective Visual Display Systems,' '382 Image Analysis,' and '707 Data Processing: Database and File Management or Data Structures.' These four classes are an integral part of Apple Computer, Inc.'s patent portfolio, receiving patents every year.

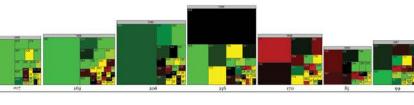
67

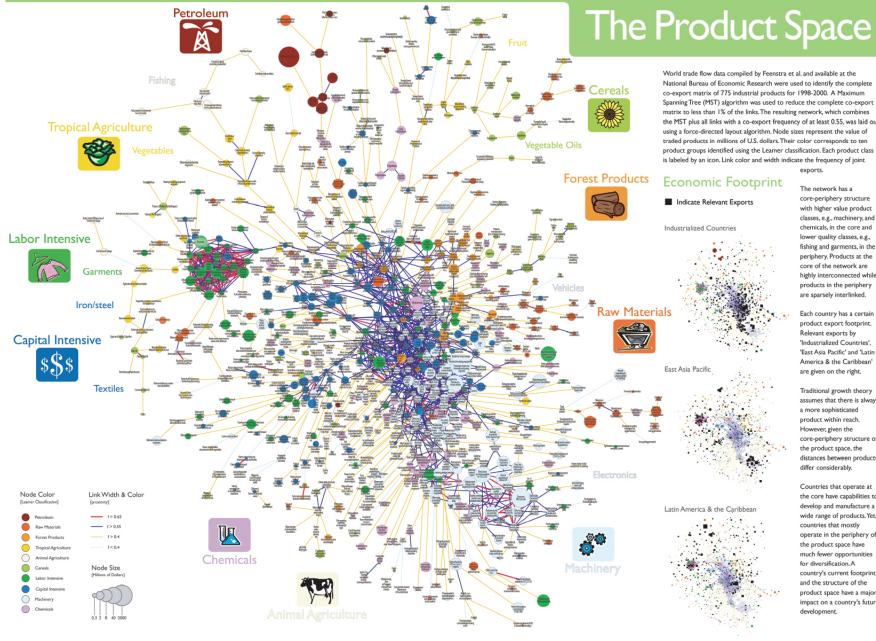
partents every year.

Jerome Lemelson

2 1 2 3 3 3 5 9 7 13 24 15

The patent portfolio of Jerome Lemelson shows a very different activity pattern. Starting in 1976, he publishes between 6–20 patents each year. However, the predominance of yellow shows that there is little continuity from previous years in regards to the classes into which patents are filed. No class dominates, Instead, more and more new intellectual space is claimed.





World trade flow data compiled by Feenstra et al. and available at the National Bureau of Economic Research were used to identify the complete co-export matrix of 775 industrial products for 1998-2000. A Maximum Spanning Tree (MST) algorithm was used to reduce the complete co-export matrix to less than 1% of the links. The resulting network, which combines the MST plus all links with a co-export frequency of at least 0.55, was laid out using a force-directed layout algorithm. Node sizes represent the value of traded products in millions of U.S. dollars. Their color corresponds to ten product groups identified using the Leamer classification. Each product class

> exports. The network has a

core-periphery structure

with higher value product classes, e.g., machinery, and chemicals, in the core and

lower quality classes, e.g., fishing and garments, in the periphery. Products at the core of the network are

highly interconnected while products in the periphery are sparsely interlinked.

Each country has a certain product export footprint. Relevant exports by 'Industrialized Countries'. 'East Asia Pacific' and 'Latin America & the Caribbean' are given on the right.

Traditional growth theory assumes that there is always a more sophisticated product within reach. However, given the core-periphery structure of the product space, the distances between products differ considerably.

Countries that operate at the core have capabilities to develop and manufacture a wide range of products. Yet, countries that mostly operate in the periphery of the product space have much fewer opportunities for diversification.A country's current footprint and the structure of the product space have a major impact on a country's future development.

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 journals. Textual descriptions of key events during the development of *Nanotechnology* are given below each frame. Most notably, leading papers in *Science* and *Nature* catalyzed the breakthrough around 2000.

CHANGING ROLES OF DIFFERENT JOURNALS

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 maps, sizes of nodes are proportional to the betweenness centrality of the respective journal 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 such as Nano Letters published by the influential American Chemical Society take the lead, and a new specialty structure with low BC value journals results.

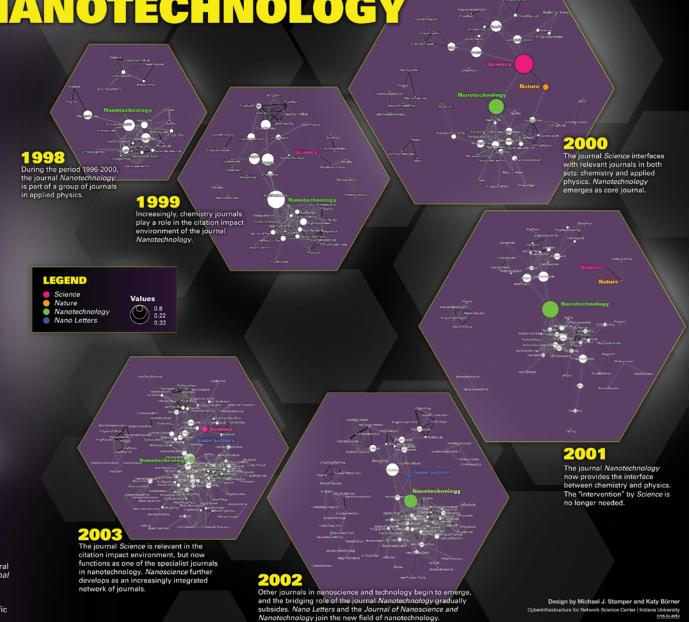


An animated sequence of this evolution is at: http://www.leydesdorff.net/journals/nanotech.

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Leydesdorff, L. and T. Schank. 2008. Dynamic Animations of Journal Maps: Indicators of Structural Change and Interdisciplinary Developments. *Journal* of the American Society for Information Science and Technology, 59(11), 1810-1818.

Price, Derek J. de Solla (1965). Networks of scientific papers. *Science*, 149, no. 3683, 510-515.



Chemical Research & Development Powers the U.S. Innovation Engine

Macroeconomic Implications of Public and Private R&D Investments in Chemical Sciences

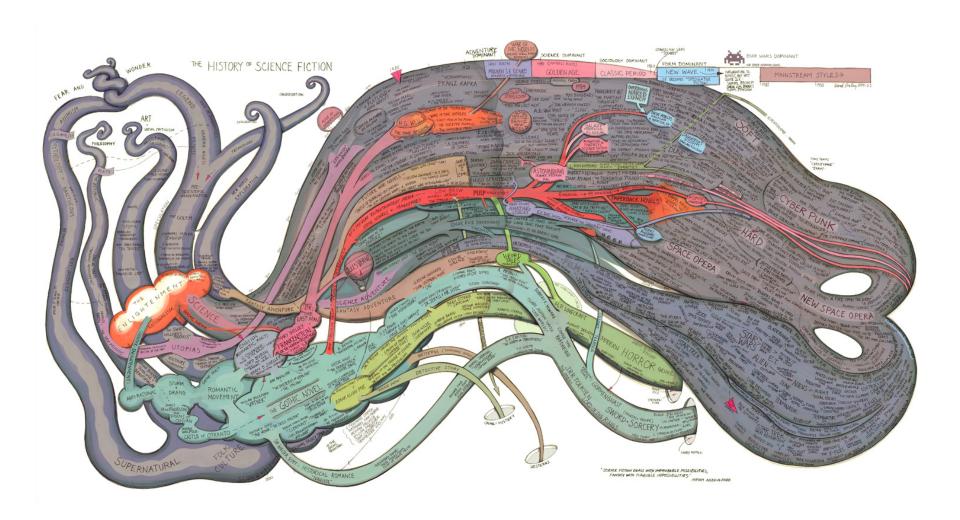
INVESTMENT IN CHEMICAL SCIENCE R&D **FEDERAL** GOVERNMENT \$8 Billion \$1 Billion **TAXES** \$5 Billion INDUSTRY FUNDING CHEMICAL INDUSTRY U.S. ECONOMY \$1B \$1B + \$5 Billion \$10 Billion \$40 Billion FOUNDATIONAL INVENTION TECHNOLOGY CHEMICAL INDUSTRY COMMERCIALIZATION **OPERATING INCOME** 600,000 - 20 YEARS -JOBS CREATED TIMELINE FROM CONCEPTION TO COMMERCIALIZATION

The Council for Chemical Research (CCR)

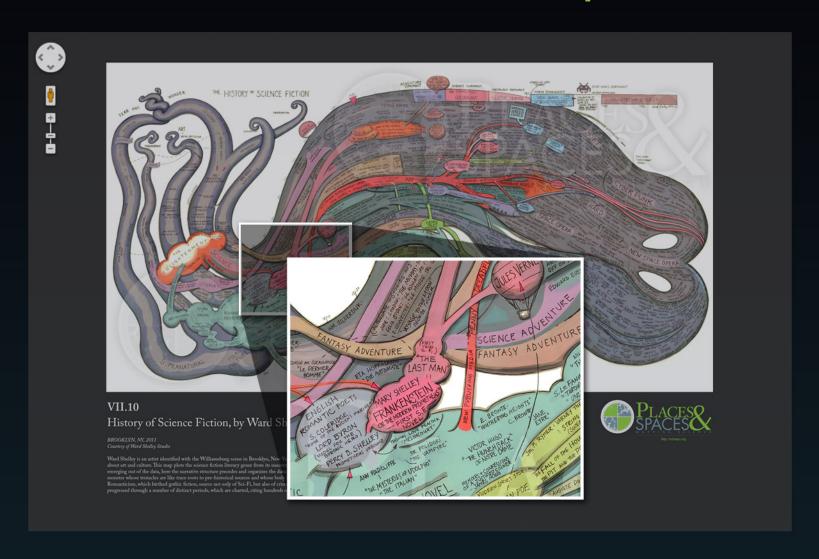
has provided the U.S. Congress and government policy makers with important results regarding the impact of Federal Research & Development (R&D) investments on U.S. innovation and global competitiveness through its commissioned 5-year two phase study. To take full advantage of typically brief access to policy makers, CCR developed the graphic below as a communication tool that distills the complex data produced by these studies in direct, concise, and clear terms.



The design shows that an input of \$1B in federal investment, leveraged by \$5B in industry investment, brings new technologies to market and results in \$10B of operating income for the chemical industry, \$40B of growth in the Gross National Product (GNP) and further impacts the US economy by generating approximately 600,000 jobs, along with a return of \$8B in taxes. Additional details, also reported in the CCR studies, are depicted in the map to the left. This map clearly shows the two R&D investment cycles; the shorter industry investment at the innovation stage to commercialization cycle; and the longer federal investment cycle which begins in basic research and culminates in national economic and job growth along with the increase in tax base that in turn is available for investment in basic research.



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(i) MACROSCOPES FOR INTERACTING WITH SCIENCE











Earth

Weather on a worldwide scale

AcademyScope

Exploring the scientific landscape

Mapping Global Society

Local news from a global perspective

Charting Culture

2,600 years of human history in 5 minutes

Iteration XI (2015): Macroscopes for Interacting with Science

http://scimaps.org/iteration/11



MAPS
vs.
MACROSCOPES



Data Visualization Literacy: Terminology & Visualization Framework

Theoretically grounded visualization framework aims to define key terminology and processes together with valid workflows and data mappings.

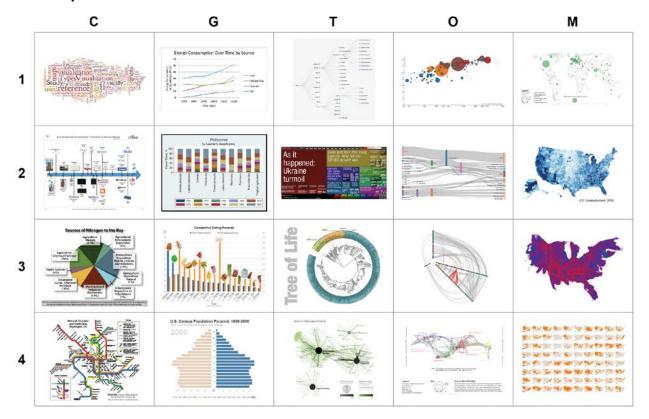
Data Visualization Literacy

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

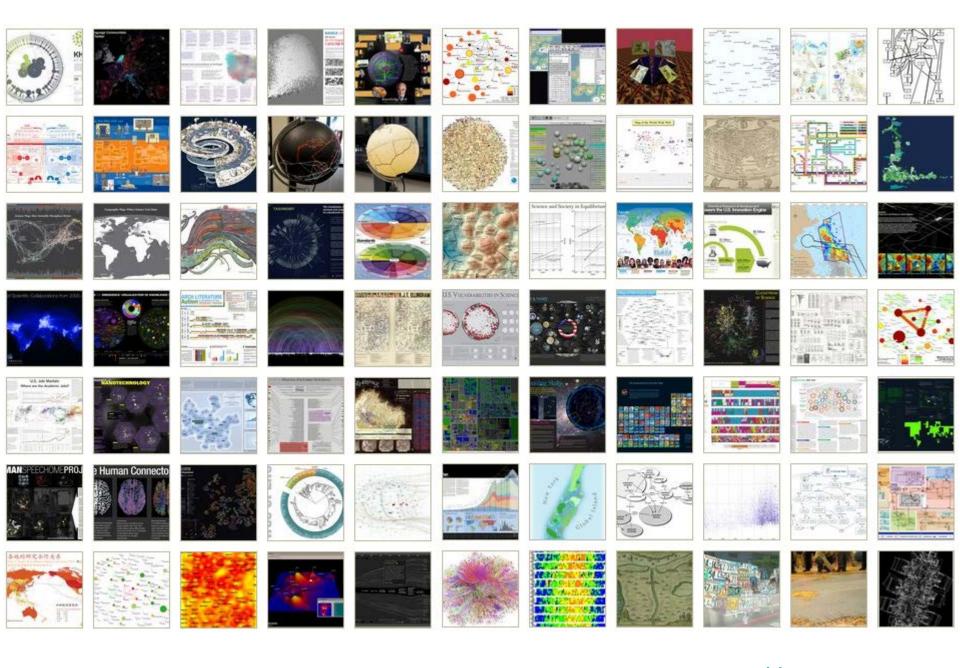
- literacy (ability to read and write text, e.g., in titles, axis labels, legend),
- visual literacy (ability to find, interpret, evaluate, use, and create images and visual media), and
- data literacy (ability to read, create, and communicate data).

Data Visualization Literacy

Is rather low: Most science museum visitors in the US cannot name, read, or interpret common data visualizations.



Börner, Katy, Joe E. Heimlich, Russell Balliet, and Adam V. Maltese. 2015. Investigating aspects of data visualization literacy using 20 information visualizations and 273 science museum visitors. Information Visualization 1-16. http://cns.iu.edu/docs/publications/2015-borner-investigating.pdf

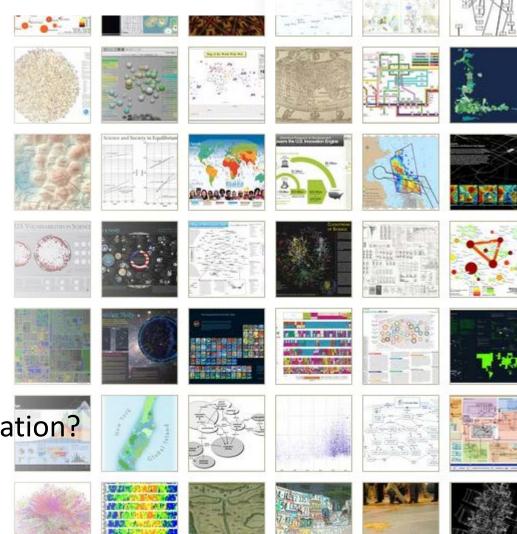


Places & Spaces: Mapping Science Exhibit, online at http://scimaps.org

How to Classify (Name & Make) Different Visualizations?

By

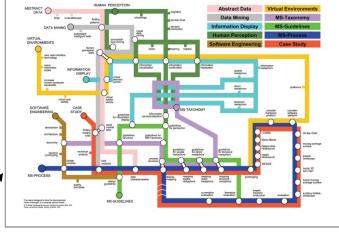
- User insight needs?
- User task types?
- Data to be visualized?
- Data transformation?
- Visualization technique?
- Visual mapping transformation?
- Interaction techniques?
- Or ?



Different Question Types



Terabytes of data



Find your way

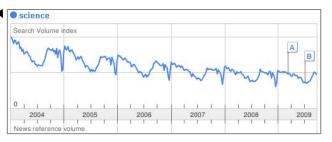
Descriptive &

Predictive

Models



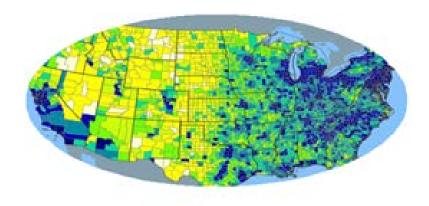
Find collaborators, friends



Identify trends

Different Levels of Abstraction/Analysis

Macro/Global Population Level



Meso/Local Group Level

Micro
Individual Level
OR
Below skin



LEVELS Tasks MICRO: Individual Level MESO: Local Level MACRO: Global Level about 1-1,000 records about 1,001-100,000 records more than 100,000 records page 10 page 6 page 8 nitan dipin dipin **TYPES** Statistical Analysis page 44 Number Productivity of scientists of Russian Knowledge versus life sciences Cartography population research and R&D costs page 135 versus GNP. page 105 page 103 WHEN: Temporal Analysis Key events page 48 Visualizing Increased in the decisiontravel and development making communication of the video processes speeds tape recorder page 95 page 83 page 85 WHERE: Geospatial Analysis page 52 Cell phone Victorian Ecological usage in poetry in footprint of Milan, Italy Europe countries page 109 page 137 page 99 WHAT: Evolving **Topical Analysis** patent Product space holdings Evolving page 56 showing of Apple co-export Computer, networks in patterns of Inc. and nanotechnology countries Jerome page 139 Technology Design Research Street page 93 Lemelson A PROPERTY OF THE PARTY OF THE WITH WHOM: Network Analysis page 60 World World-wide Electronic and Finance scholarly new media art Corporation collaboration networks network networks Atlas of Knowledge page 133 page 87 page 157 Anyone Can Map

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

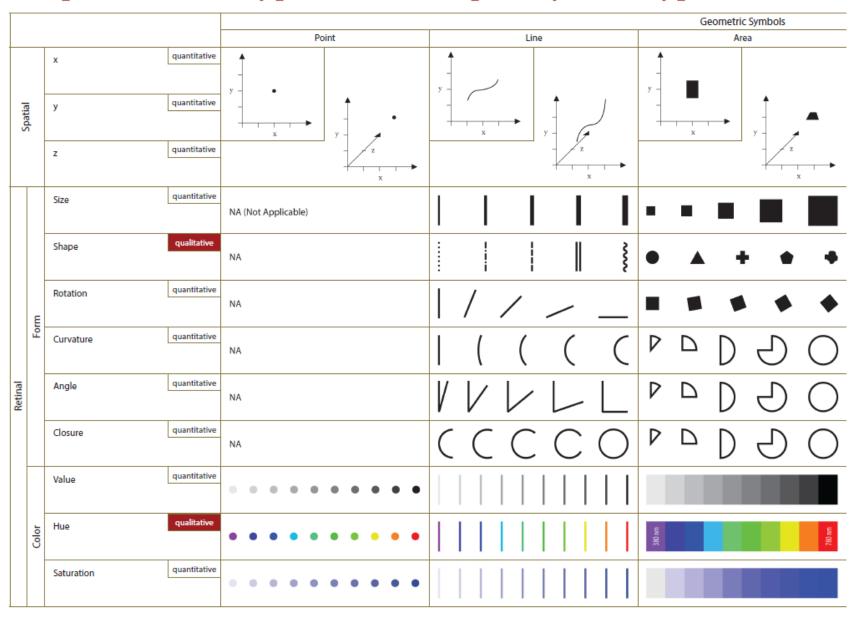
Visualization Framework

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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Graphic Variable Types Versus Graphic Symbol Types

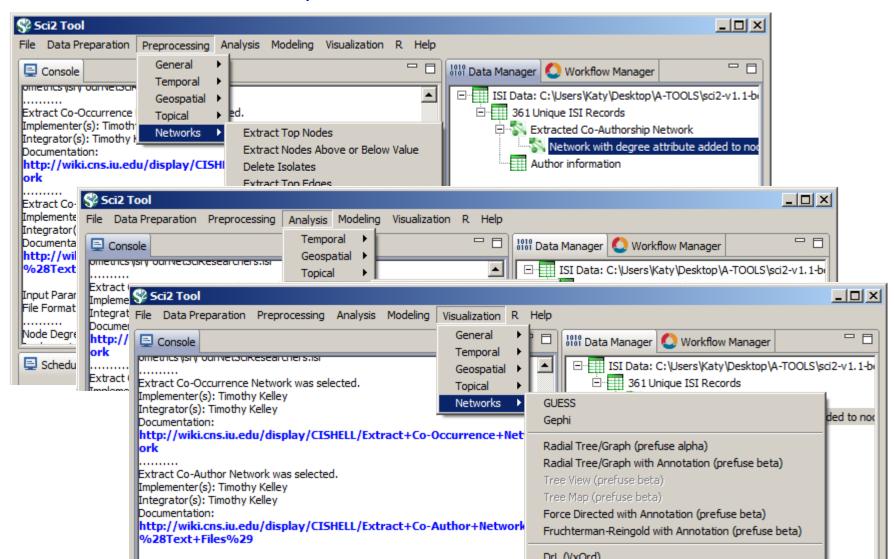


Graphic Variable Types Versus Graphic Symbol Types Grometric Symbols Linguistic Symbols Fictorial Symbols questilative quantitative size. NA Nat Applicable Text That curature ome table cells are left blank to encou questions clours substation specing combally **Fallery** quantitative crimitation credient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



Sci2 Tool Interface Components Implement Vis Framework

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



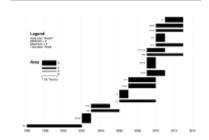
Load **One** File and Run **Many** Analyses and Visualizations

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

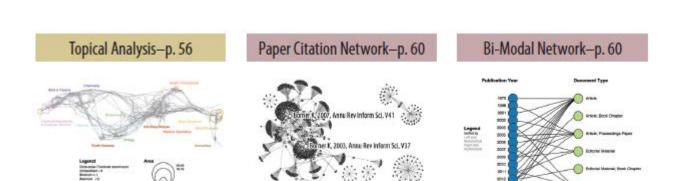


Geospatial Analysis—p. 52



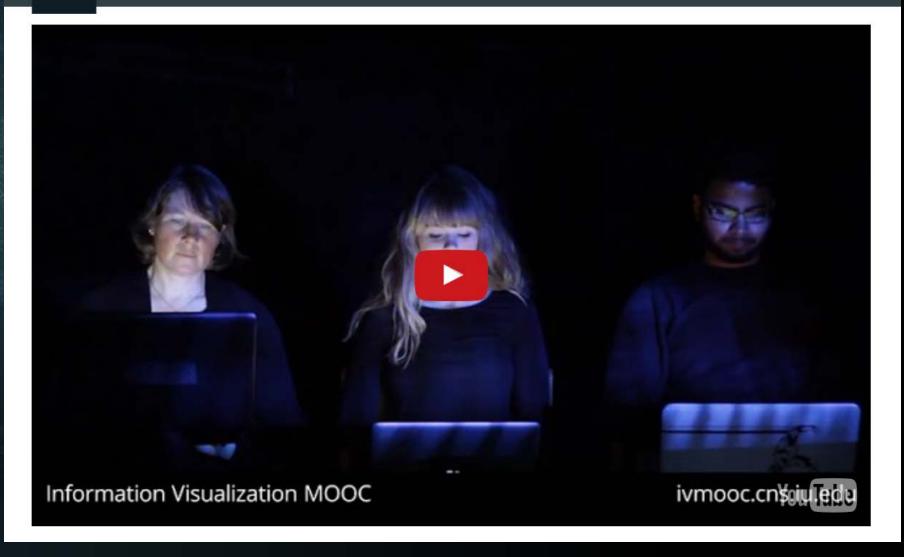
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Co-author and many other bi-modal networks.





Register for free: http://ivmooc.cns.iu.edu. Class restarts Jan 9, 2018.

Data Visualization Literacy: Outlook

Data Visualization Literacy: Research and Tools that Advance Public Understanding of Scientific Data

NSF Org: DRL

Division Of Research On Learning

Initial Amendment Date: June 13, 2017

Latest Amendment Date: June 13, 2017

Award Number: 1713567

Award Instrument: Standard Grant

Program Manager: Arlene M. de Strulle

DRL Division Of Research On Learning

EHR Direct For Education and Human Resources

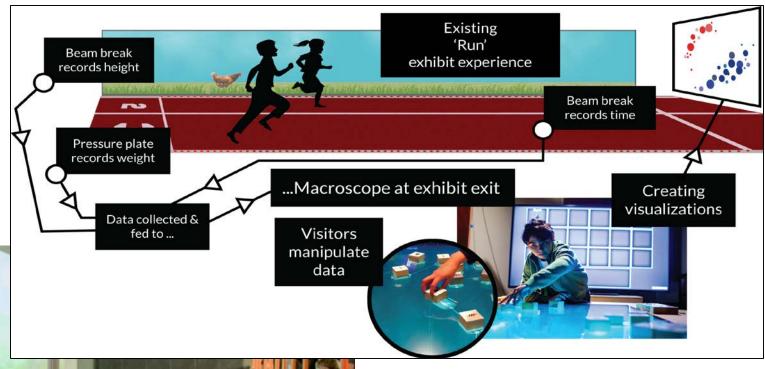
Start Date: August 1, 2017

End Date: July 31, 2021 (Estimated)

Awarded Amount to Date: \$1,355,236.00

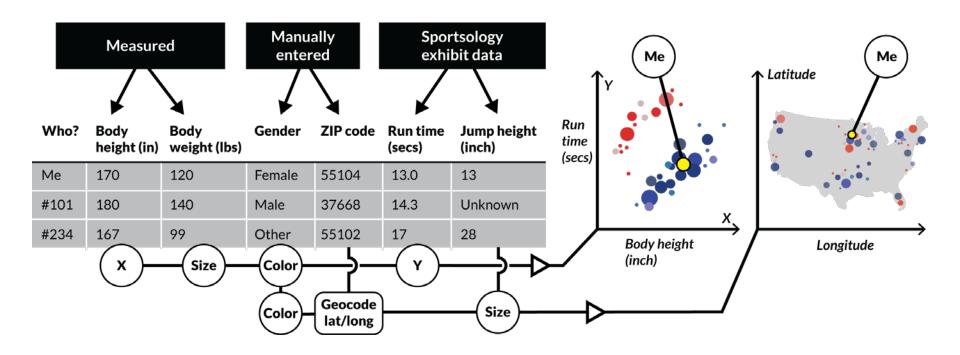
Investigator(s): Katy Borner katy@indiana.edu (Principal Investigator)

Kylie Peppler (Co-Principal Investigator) Bryan Kennedy (Co-Principal Investigator) Stephen Uzzo (Co-Principal Investigator) Joe Heimlich (Co-Principal Investigator)



Section 1. Control of the control of

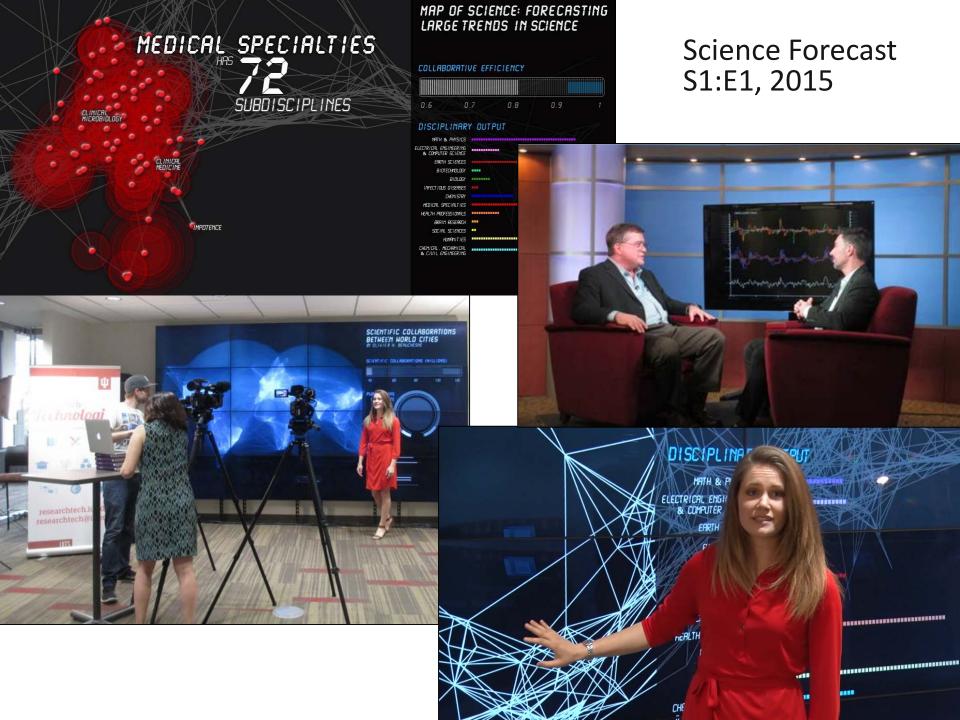
Sketch of the *Run* exhibit including data collection (top) and macroscope addon that lets interested visitors explore more complex data visualizations using table-top displays.



xMacroscope general setup and activity—Raw data on left is converted to visualization on right by dragging and dropping (or connecting) column headers to axes, paint buckets, size, and shape.

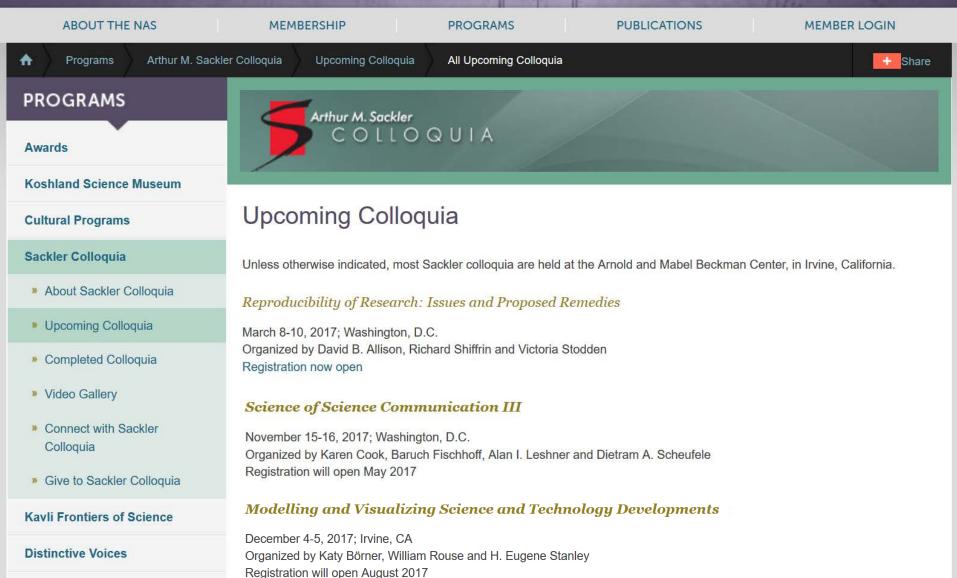


CSWS Session: Visualizing STEAM Data in Support of Smart Decision Making November 15-17, 2017, Tokyo, Japan. http://scws2017.org









References

Börner, Katy, Chen, Chaomei, and Boyack, Kevin. (2003). **Visualizing Knowledge Domains.** In Blaise Cronin (Ed.), *ARIST*, Medford, NJ: Information Today, Volume 37, Chapter 5, pp. 179-255. http://ivl.slis.indiana.edu/km/pub/2003-borner-arist.pdf

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). http://www.pnas.org/content/vol101/suppl_1

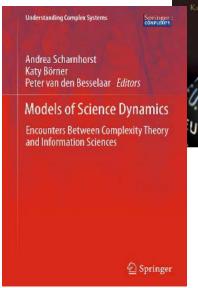
Börner, Katy (2010) **Atlas of Science: Visualizing What We Know**. The MIT Press. http://scimaps.org/atlas

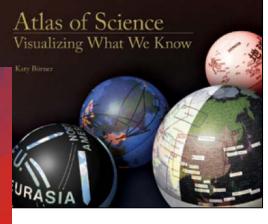
Scharnhorst, Andrea, Börner, Katy, van den Besselaar, Peter (2012) **Models of Science Dynamics**. Springer Verlag.

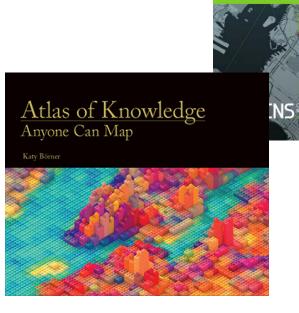
Katy Börner, Michael Conlon, Jon Corson-Rikert, Cornell, Ying Ding (2012) VIVO: A Semantic Approach to Scholarly Networking and Discovery. Morgan & Claypool.

Katy Börner and David E Polley (2014) **Visual Insights: A Practical Guide to Making Sense of Data**. The MIT Press.

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









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

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