

Connecting the Dots: Museum Visitors and Data Visualization Literacy

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January 25, 2018 | 1:00 PM Exploratorium Journal Club



CNS Tools and Services are used around the globe.



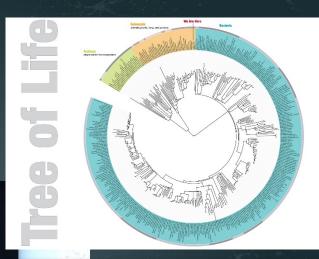
Our mission is to advance datasets, tools, and services for the study of biomedical, social and behavioral science, physics, and other networks. A specific focus is research on the structure and evolution of science and technology (S&T) and the communication of results via static and interactive maps of science.







scimaps.org





100

MAPS

in large format, full color, and high resolution.

24



MACROSCOPE MAKERS including one whose job title is "Truth and Beauty Operator."

354

DISPLAY VENUES AND EVENTS

from the Cannes Film Festival to the World Economic Forum.

5



visited in 2016 including Ghent, Belgium, and Valencia, Spain. 215

MAPMAKERS

from fields as disparate as art, urban planning, engineering, and the history of science.

8

MACROSCOPES

for touching all kinds of data.

214

PRESS ITEMS

including articles in Nature, Science, USA Today, and Wired.

4,378,916

WEBSITE VISITS

to scimaps.org since 2005.









Macroscopes

Macroscopes are tools that help people focus on patterns in data that are too large or complex to see unaided.

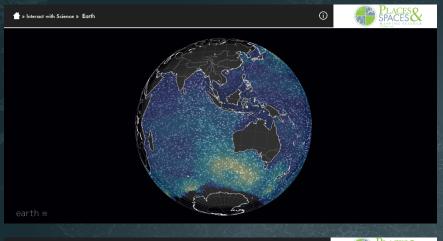


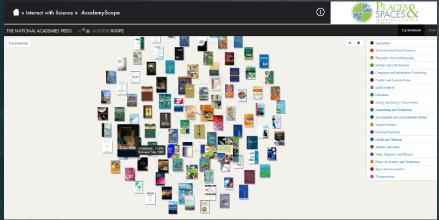


Why move from maps to macroscopes?

- Data visualizations are increasingly interactive.
- Manipulating visualizations can help people understand them better.
- We want visitors to take an active role in using and understanding data.







Earth Cameron Beccario

earth.nullschool.net

AcademyScope

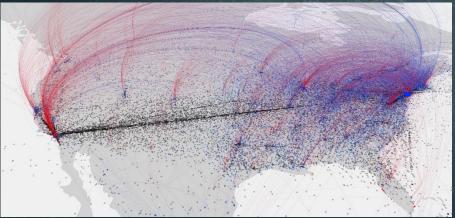
National Academy of Sciences and Cyberinfrastructure for Network Science Center

nap.edu/academy-scope/#top-downloads



Mapping Global News Kalev Leetaru

gdeltproject.org

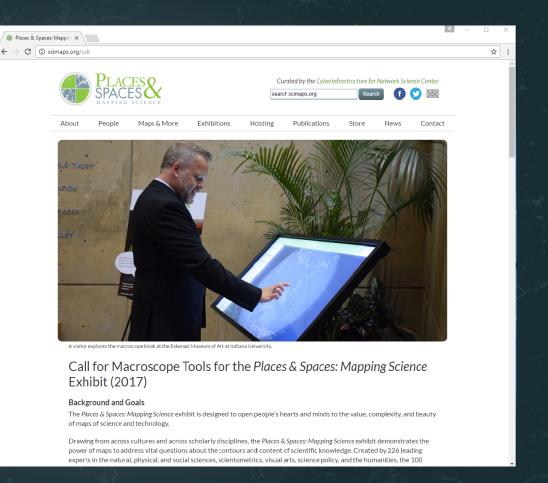


Charting Culture
Maximilian Schich and
Mauro Martino

cultsci.net

Join us!

- Submissions due Jan 31, 2018
- Recommend others or submit your own work
- For details: scimaps.org/call



Pathways

Sense-Making of Big Data

The project examined the data visualization literacy of over 900 youth and adult visitors across five US science museums. The New York Hall of Science and Science Museum of Minnesota are both involved as partner institutions, providing financial support, facilities, and collaborative research. Data collection took place at the New York Hall of Science, the Marian Koshland Science Museum, COSI in Columbus Ohio, and Wonderlab Museum in Bloomington, IN.



Pathways: Sense-Making of Big Data. NSF ISE DRL-1223698 Award (Katy Börner, Adam V. Maltese, Joe E. Heimlich, Stephen Miles Uzzo, Paul Martin, and Sasha Palmquist, \$250,000) 2012.07.01 - 2015.06.30.



Participants from the November 2013 Pathways Workshop at the Science Museum of Minnesota



Jax and the Big Data Beanstalk, a Science Museum of Minnesota theater piece funded by the NSF, introduces museum visitors to big data visualizations and science maps

What is Big Data?

More than two-thirds of visitors interviewed said that they had not previously heard the phrase "Big Data."

—Sense Making of Big Data, Heimlich, Tranby, Wojton 2014

"Important information. Something everybody relates to, but doesn't understand."

—Project participant

"[lt] gives me anxiety. I don't know and I don't like it."

—Project participant

Big Data Insight Needs Sort

Looking for generalizable ways visitors engage with and then make meaning of big data sets.

> Eggs Crackers Lasagna Apples Tofu Almonds



Meaning making through data representation construction and deconstruction

For better comprehension

Build a sandwich . . .

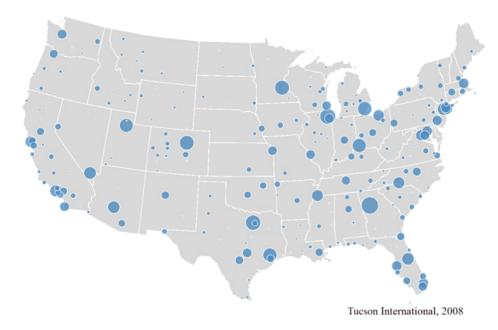




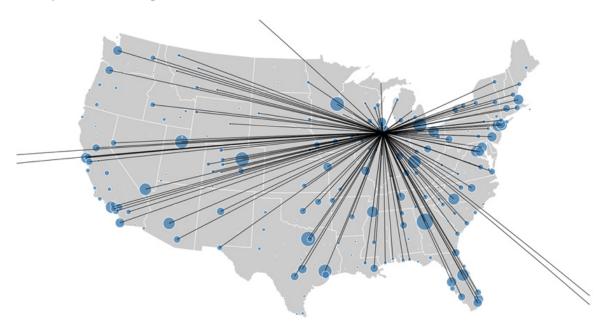
... instead of peeling an onion.



Circle size equals number of flights per day.

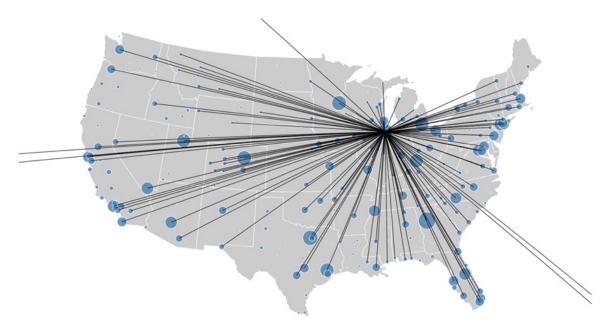


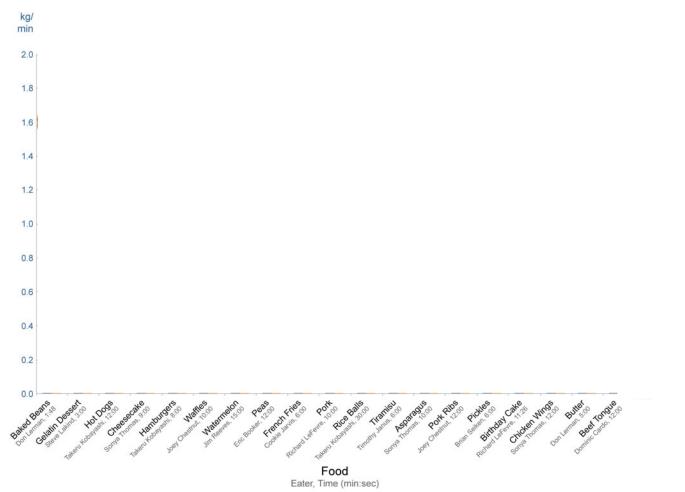
Links represent flight connections.

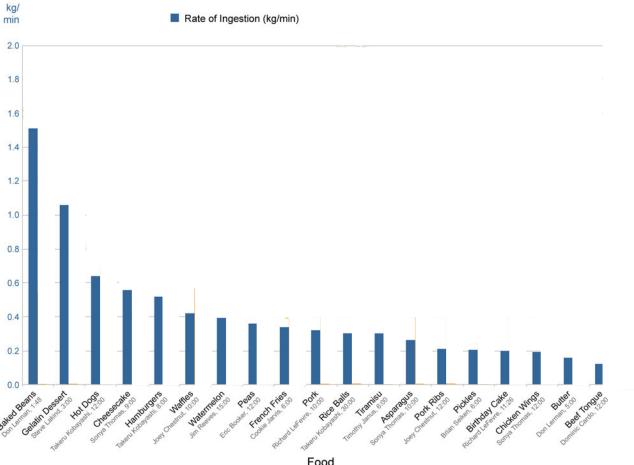


Airports reachable from Chicago O'Hare International Airport in 2008

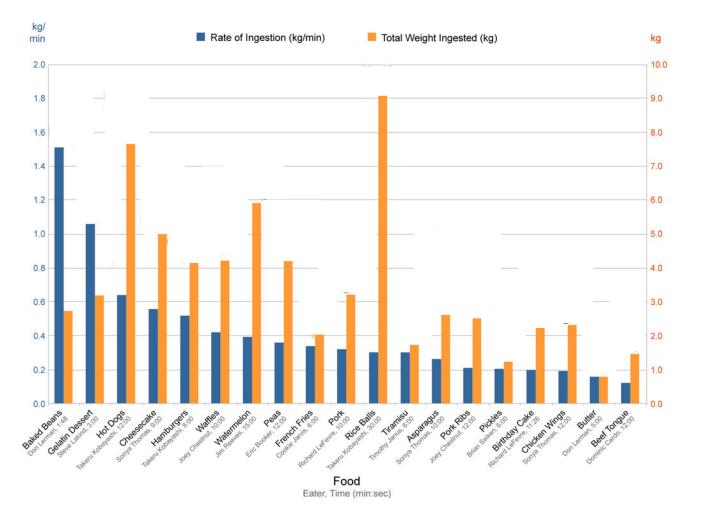
Circle size equals number of flights per day.

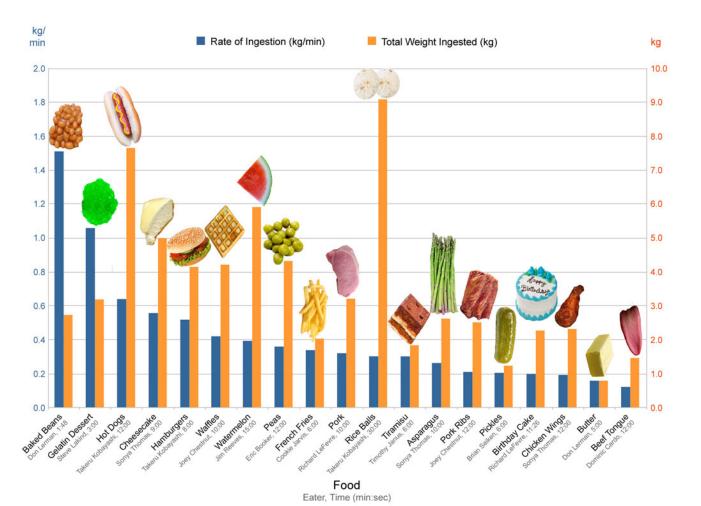


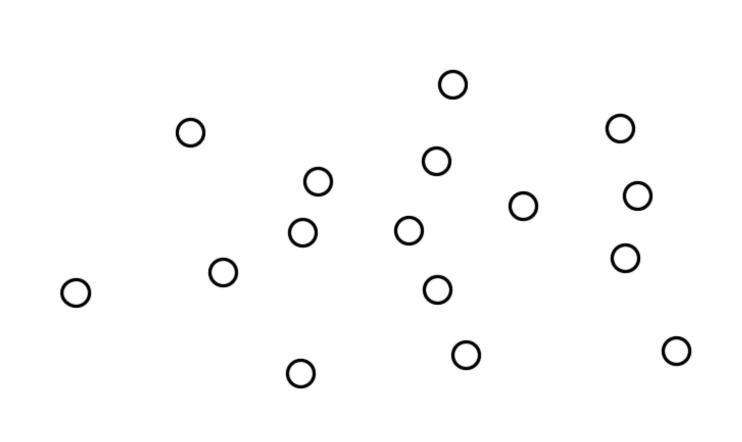


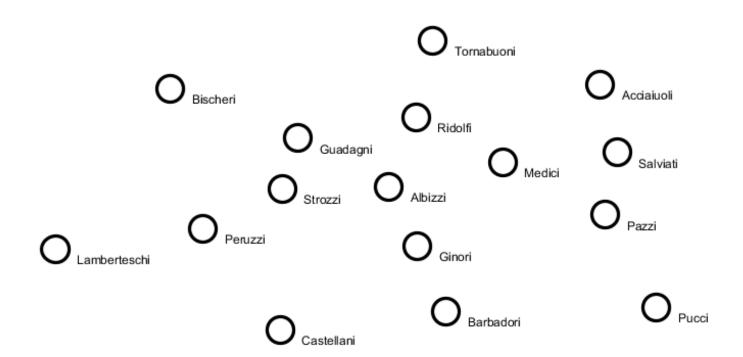


Food Eater, Time (min:sec)

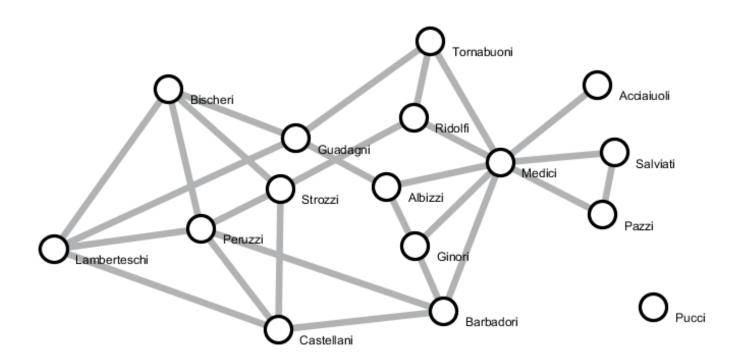




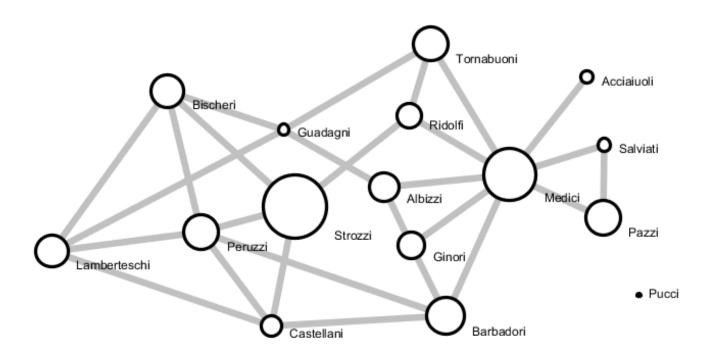




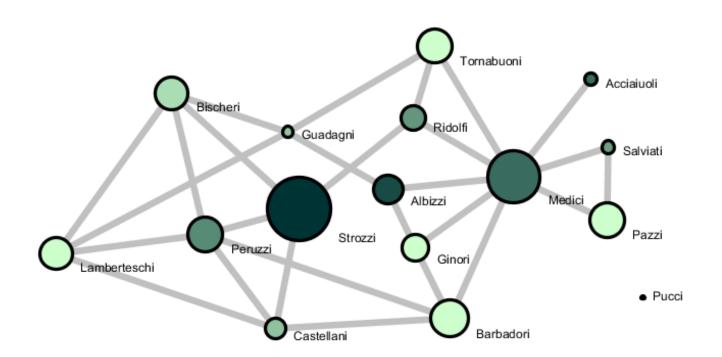
Label: Family Name



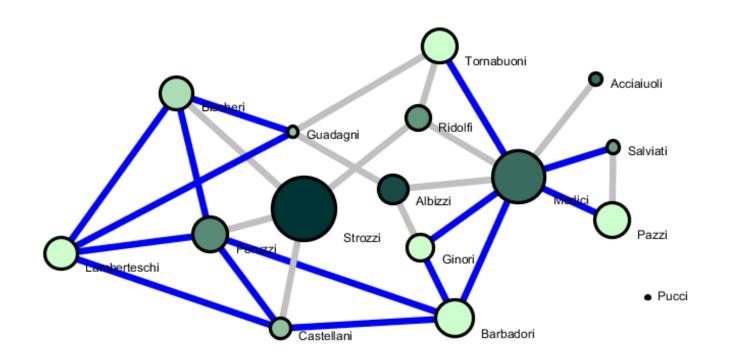
Label: Family Name



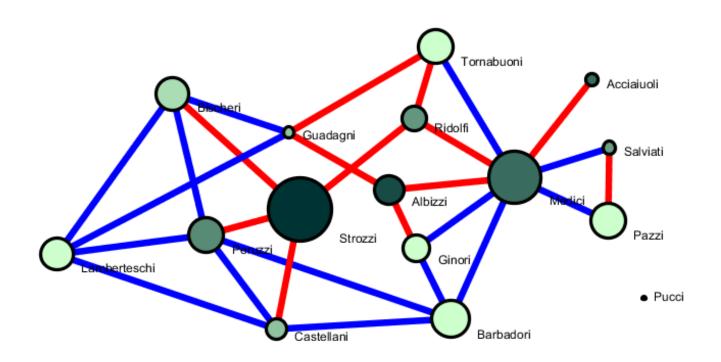
Circle size: Family's net wealth (in thousands of lira) in year 1427.



Circle color: Number of seats on the civic council held 1282-1344.

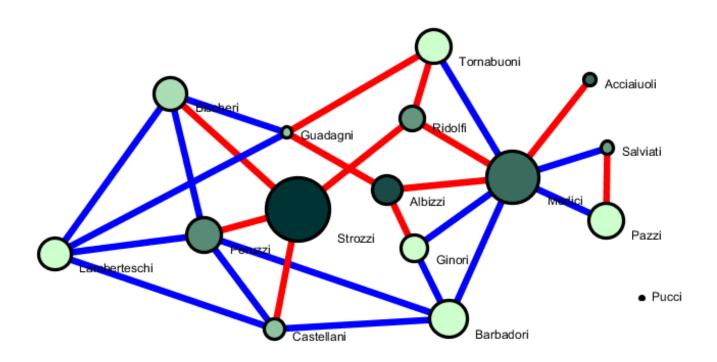


Link color: Business ties, e.g., loans, credits, and joint partnerships.



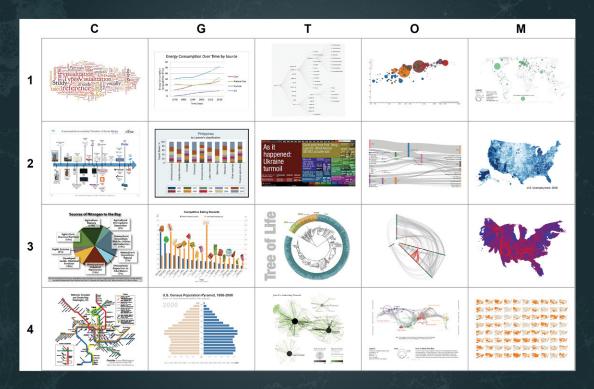
Link color: Marriage alliances.

Padgett's Florentine Families



Link color: Marriage alliances.

Data Visualization Literacy: Can 273 Science Museum Visitors Read 20 Information Visualizations?



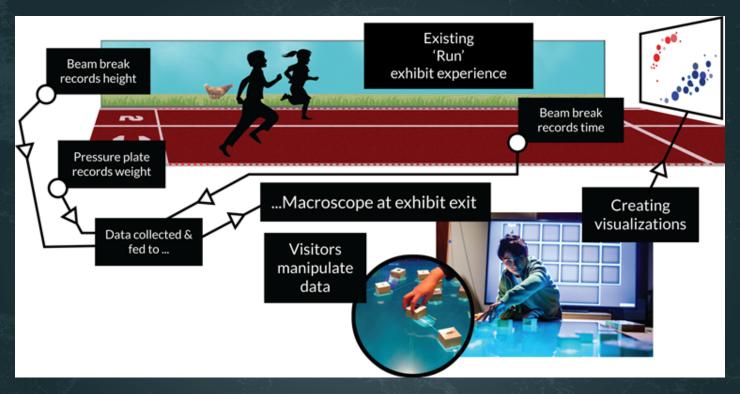
Börner, Katy, Joe E. Heimlich, Russell Balliet, and Adam V. Maltese. (Submitted). "Investigating Aspects of Data Visualization Literacy Using 20 Information Visualizations and 273 Science Museum Visitors". Information Visualization.

Visualization Types (Reference Systems)

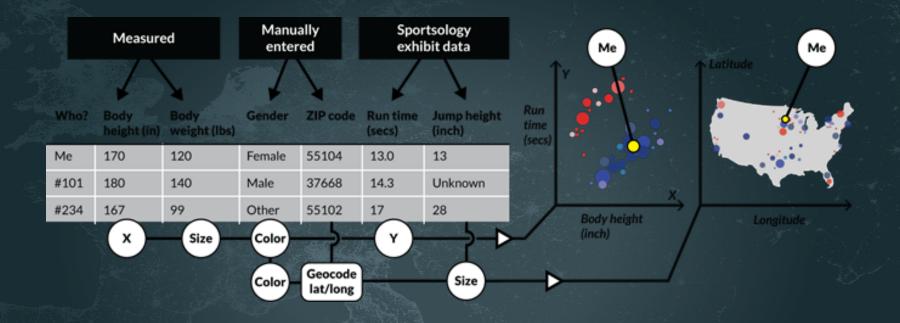
- **1. Charts:** No reference system—e.g., Wordle.com, pie charts
- **Tables:** Categorical axes that can be selected, reordered; cells can be color coded and might contain proportional symbols. Special kind of graph.
- **3. Graphs:** Quantitative or qualitative (categorical) axes. Timelines, bar graphs, scatter plots.
- **4. Geospatial maps:** Use latitude and longitude reference system. World or city maps.
- 5. Network graphs: Node position might depends on node attributes or node similarity. Tree graphs: hierarchies, taxonomies, genealogies. Networks: social networks, migration flows.

Intro to AISL project

- Research into how to define and test for data visualization literacy
- Develop the xMacroscope, a platform for research and for exhibit development



Sketch of the *Run* exhibit including data collection (top) and macrocope add-on at exhibit exit that lets interested visitors explore more complex data visualizations (lower right).



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

Who is willing to share what information, when, and with whom?

Data Visualization Literacy Literature Review: Sharing Personal Information E. Elaine T. Horr, Ph.D. Joe E. Heimlich, Ph.D.

October 2017







Resources

IVMOOC

Information Visualization MOOC

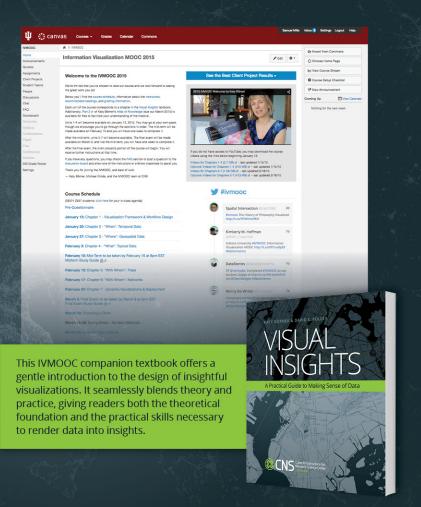
The Information Visualization MOOC provides an overview about the state of the art in information visualization, teaching the process of producing effective visualizations that take the needs of users into account.

The inaugural IVMOOC, which launched in January 2013, attracted participants from more than 100 countries. It is one of the first MOOCs offered by IU and the first to offer an opportunity for students to work in teams with real clients. All registrants gain free access to the Scholarly Database and the Sci2 Tool.

The course can be taken for three Indiana University credits as part of the Online Data Science Program offered by the School of Informatics and Computing.

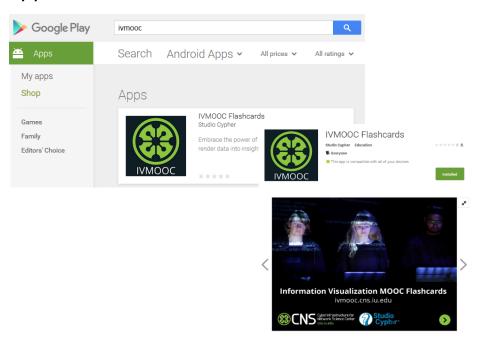
The course will return in January 2016. Learn more at ivmooc.cns.iu.edu.

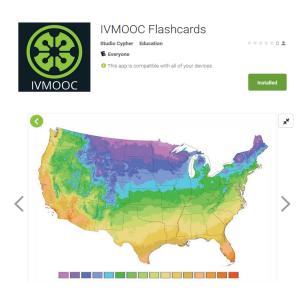




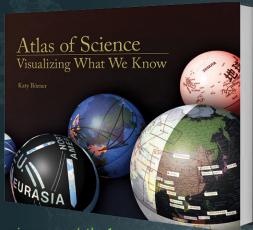
IVMOOC App – More than 60 visualizations

The "IVMOOC Flashcards" app can be downloaded from Google Play and Apple iOS stores.



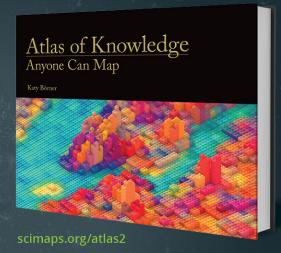


Enjoy the first two books in Katy Börner's 3-Part *Atlas* series



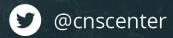
scimaps.org/atlas1

Atlas of Science, featuring more than thirty full-page science maps, fifty data charts, a timeline of science-mapping milestones, and 500 color images, serves as a sumptuous visual index to the evolution of modern science and as an introduction to "the science of science"—charting the trajectory from scientific concept to published results.



The Atlas of Knowledge introduces a theoretical visualization framework meant to empower anyone to systematically render data into insights. It aims to teach "timeless" knowledge that holds true over a lifetime while referring to an extensive set of references for "timely" advice on what tool and workflow is currently the best for answering a specific question.





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.

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