# Reproducibility in Scientometrics: Data Enclaves, Open Code, and Open Education



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



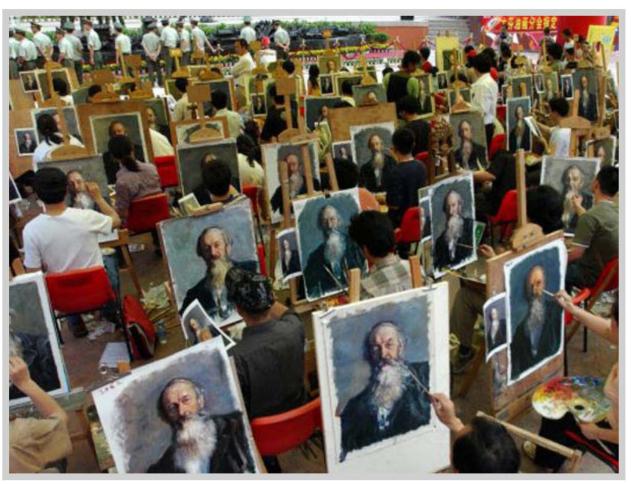
Workshop on "Reproducible Scientometrics Research" at ISSI 2017 in Wuhan, China

Oct 17, 2017



# Reproducibility of Research: Issues and Proposed Remedies

http://www.nasonline.org/programs/sackler-colloquia/completed\_colloquia/Reproducibility\_of\_Research.html



March 8-10, 2017; Washington, D.C.
Organized by David B. Allison, Richard Shiffrin and Victoria Stodden

(Help | Advanced search)

### Statistics > Other Statistics

### Scientific progress despite irreproducibility: A seeming paradox

Richard M. Shiffrin, Katy Borner, Stephen M. Stigler

(Submitted on 5 Oct 2017)

It appears paradoxical that science is producing outstanding new results and theories at a rapid rate at the same time that researchers are identifying serious problems in the practice of science that cause many reports to be irreproducible and invalid. Certainly the practice of science needs to be improved and scientists are now pursuing this goal. However, in this perspective we argue that this seeming paradox is not new, has always been part of the way science works, and likely will remain so. We first introduce the paradox. We then review a wide range of challenges that appear to make scientific success difficult. Next, we describe the factors that make science work-in the past, present, and presumably also in the future. We then suggest that remedies for the present practice of science need to be applied selectively so as not to slow progress, and illustrate with a few examples. We conclude with arguments that communication of science needs to emphasize not just problems but the enormous successes and benefits that science has brought and is now bringing to all elements of modern society.

Comments: 3 figures

Subjects: Other Statistics (stat.OT) Cite as: arXiv:1710.01946 [stat.OT]

(or arXiv:1710.01946v1 [stat.OT] for this version)

### Submission history

From: Katy Börner [view email]

[v1] Thu, 5 Oct 2017 10:10:56 GMT (286kb)

### Download:

PDF only

(cc) ZERO

Current browse context:

stat.OT

< prev | next > new | recent | 1710

Change to browse by:

stat

References & Citations

NASA ADS

Bookmark (what is this?)









https://arxiv.org/abs/1710.01946

### **Data Enclaves**

# http://cns.iu.edu/workshops/event/161114.html

Workshop web site features slides

and links to relevant efforts

Web of Science as a Research Dataset

Date:

November 14-15, 2016

**Meeting Place:** 

Social Science Research Commons (SSRC), Indiana University Woodburn Hall, Room 200 1100 East Seventh Street Bloomington, IN 47405

Web Indiana University Campus Map »

### Organizers:



Katy Börner

Victor H. Yngve Distinguished Professor of Information Science Director, CI for Network Science Center Curator, Mapping Science Exhibit

School of Informatics and Computing Indiana University, Bloomington katy@indiana.edu



**Eamon Duede** 

Executive Director Knowledge Lab University of Chicago eduede@uchicago.edu



**James Pringle** 

Head of Industry Development Clarivate Analytics



# Web of Science™ as a Research Dataset









Katy Börner,1 Valentin Pentchev,2 Matthew Hutchinson,2 James Pringle,3 Jason Rollins,3 Yadu N. Babuji,4 & Eamon Duede

1katy@indiana.edu CNS, SOIC & Network Science Institute, Indiana University, Bloomington, IN, US <sup>2</sup>maahutch@iu.edu & vpentche@iu.edu IUNI, Indiana University, Bloomington, USA

³jason.rollins@clarivate.com & james.pringle@clarivate.com Clarivate Analytics, USA fyadunand@uchicago.edu Computation Institute, Knowledge Lab, UofChicago, USA

<sup>5</sup>eduede@uchicago.edu Computation Institute, Knowledge Lab, Committee on the Conceptual and Historical Studies of Science, Uof Chicago, USA

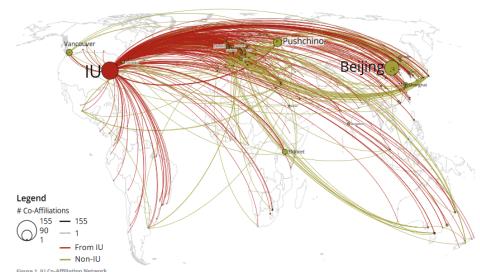
#### Introduction

The Clarivate Analytics Web of Science (WoS) has served as a research dataset for more than 9,000 scholarly articles in the past 15 years alone—across a wide a range of fields and disciplines from toxicology to computer science to economics. Scientists and scholars have been particularly interested in the WoS citation network, a massive graph containing billions of links that can proxy the structure and dynamics of not only scholarly communication, but knowledge diffusion, the evolution of fields, and the career lifecycles of individuals and institutions. To power these investigations, scholars are increasingly employing a number of compute-intensive methodologies, sophisticated big data infrastructures, and so called collaborative "discovery science" tools and techniques. Suddenly, in addition to deep, domain specific expertise, world-class computational knowhow appears to be a new prerequisite for analysis of scholarly data at the scale represented by WoS. While cloud-based computing and tools are more prevalent and accessible than ever before, harnessing these technologies remains both a challenge and opportunity for researchers and data providers (i.e., Clarivate Analytics and similar commercial data vendors and non-commercial aggregators). While the opportunities made possible by scholarly data at the size and scope of WoS for discovery and innovation are limited only by imagination, two general prospects come readily to mind. First, access to these data coupled with the appropriate computational and analytical capabilities opens up a wide range of funding and subsequent publishing opportunities in high impact venues. Second, data providers can pursue new business opportunities, including novel data access models, new types of analytic products, and new kinds of academic/industry partnerships. In this poster paper, we briefly explore 1) the new computational infrastructures that are being developed to enable collaborative research that leverages scholarly datasets such as WoS that are both big and proprietary: 2) some recent findings that have been made possible by these infrastructures; and, 3) new commercial offerings that have been enabled and demanded in response to increasing reliance on the

#### **New Computational Infrastructures**

Research leveraging big, scholarly datasets like WoS presents researchers with challenges related to the data's size, inherently relational format, and sensitive (proprietary) nature. To overcome these challenges, researchers have developed a new generation of enclave supported, high performance, and cloud-based, collaborative research environments that are both elastic enough to provide substantial computational resources when needed while remaining secure enough to protect data providers'

### IU International Co-Affiliation Network, 2004-2013 CNS @ Indiana University



commercial interests. Moreover, these environments have evolved to allow geographically distributed researchers to collaborate on research projects in the fast paced, iterative style that has come to dominate research in the era of Big Data—namely, "discovery science".

The Indiana University Network Science Institute (IUNI) acquired the complete set of Clarivate Analytics' Web of Science XML raw data (Web of Knowledge version 5). The data was parsed and stored in a well-documented Postgresql database, see entity-relationship diagram, database schema, and data dictionary on http://iuni.iu.edu/resources/web-of-science. The code used to parse the WoS XML format and to save data in the Postgresql database was made available freely on GitHub, see https://glthub. lu.edu/CNS/generic\_parser. All data can be accessed via the IUNI WoS Data Enclave, a secure repository that uses IU's Karst high-throughput computing cluster designed to deliver large amounts of processing capacity over long periods of time. Access to the XML data and the PostgreSQL database is granted to a user's Karst account. IU faculty, staff, and qualifying sponsored affiliates can request accounts on Karst to use the data for academic research and without any sharing of data. A simple web browser based query interface to the WoS dataset was implemented to support custom queries for specific terms, journals, or authors. Datasets can be downloaded in CSV data format compatible with data mining and visualization tools such as Gephi or the Sci2 Tool (http://sci2.cns.iu.edu) (Sci2 Team, 2009). More about the IUNI WoS Data Enclave can be found at http://lunl.lu.edu/resources/web-of-science.

#### Cloud Kotta

One platform specifically developed with WoS in mind is Knowledge Lab's Cloud Kotta (CK). CK is a secure data enclave and analytics platform that serves the research needs of social sciences (Babuii 2016). By hosting CK in the Amazon Web Services cloud, the developers were able to take advantage of virtually limitless compute, cost-effective storage and the ability to implement a fine-grained security model ensuring the authorized collaborators could access both data and compute resources from any where in the world (Babuji 2016). Moreover, CK supports multiuser, rapid ideation and research iteration through a novel Python library that enables specific functions in an analysis code, written in a Jupyter Notebook to be seamlessly and securely submitted to the CK execution fabric (Babuji 2017). By allowing researchers to develop and share analysis code interactively over secure data like WoS, CK has removed the need for deep computational infrastructure expertise. The complete WoS XML dataset was ingested into a relational database housed in CK using a custom parser that has been made freely available on GitHub (see: https://glthub.com/alexander-bellkov/wos\_parser). The Cloud Kotta WoS database schema can

be found on CK's documentation pages (see: http:// docs.cloudkotta.org/datagulde/wos.html). More about Cloud Kotta can be found at http://docs.

#### **New Computational Infrastructures**

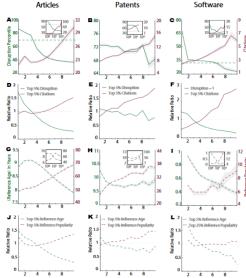
#### Fostering Global Collaboration

Among others, IU started to use the IUNI WoS data to understand existing and foster global research collaborations. The world map in Figure 1 shows the co-affiliations of authors that listed "Indiana Univ" and at least one other non-U.S. institution as affiliation on 1,590 scholarly papers published in 2004-2013. There are 344 affiliation locations (not counting IU) and 641 co-affiliation links. Nodes denote author locations and are area size coded by degree with the exception of IU, which has 1.592 co-affiliation links. Links denote co-affiliations, e.g., an author with three affiliations IU, X, Y will add three links; the two links that connect IU with X and Y are drawn in red while the link between X and Y is given in green. Links are size coded by the number of co-affiliations with the top-three being Beijing, China (155), Eldoret, Kenya (115), and Pushchino, Russia (90)

#### Impact vs. Disruptiveness

Researchers at the University of Chicago's Knowledge Lab and Northwestern University's NICO have used WoS data going back to 1900 to study the relationship between team size and impact and the relationship between team size and disruptiveness. This work, currently under review, finds striking differences between the scientific output of large and small teams. Looking across all fields represented in WoS, small teams are shown to disrupt science, patents, and software with new ideas and opportunities, while large teams contribute to existing ones. Figure 2 shows the relationship between impact and disruptiveness of articles (left panel) indexed by WoS, patents (middle), and software (right). In all three spaces, there is a strong, inverse relationship between citations and disruptiveness as team size increases.

Articles **Patents** 



#### Figure 2. Tracing Inventive Team

#### New Commercial Offerings

The value of the Web of Science as a search and discovery tool is well established at thousands of research institutions worldwide. But the commercial opportunities for the use of its high-quality metadata outside of the platform for big data studies are still emerging. When researchers need to study broadscale trends in science, technology, and innovation, they very often turn to the Web of Science as the most comprehensive citation source to provide over 100 years of consistent, global publication data. Increasingly, user requests for this data take the form of custom reports, curated data sub-sets, and largescale raw XML delivery. Clarivate Analytics is actively looking at compelling ways to meet these customer demands with new commercial products and data delivery choices. These options must balance scale and ease of use, with security and control over access to the proprietary WoS data. The lessons learned in the development of Cloud Kotta and IUNI WoS Data Enclave will very likely be instructive here, as they have proven their utility and leverage a mix of custom code built on proven commercial cloud services Both self-service data access and secure use of analytical tools in a cloud "sandbox" seem like attractive features of these environments that could make commercial sense to meet the evolving expectation of Web of Science customers.

#### Acknowledgements

This work as partially supported by and contributes to research for IBM, Facebook, Jump Trading, AWS, Clarivate Analytics, the National Institutes of Health under awards P01 AG039347 and U01CA198934 and the National Science Foundation under awards NCSE-1538763, EAGER 1566393, and NCN CP Supplement 1553044. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation. This work uses Web of Science data by Clarivate Analytics.

Babuji, Y. N., Chard, K., Gerow, A., & Duede, E. (2016). Cloud Kotta: Enabling Secure and Scalable Data Analytics in the Cloud. IEEE Big Data 2016.

Babuji, Y. N., Chard, K., Gerow, A., & Duede, E. (2016). A Secure Data Enclave and Analytics Platform for Social Scientists. IEEE eScience 2016.

Babuji, Y. N., Chard, K., & Duede, E. (2017). Enabling Interactive Analytics of Secure Data using Cloud Kotta. Science Cloud Workshop: ACM International Symposium on High-Performance Parallel and Distributed Computing 2017 (Forthcoming)

Sci2 Team. (2009). Science of Science (Sci2) Tool. Indiana University and SciTech Strategies, http://scl2.

Poster at ISSI 2017, <a href="http://cns.iu.edu/docs/publications/2017-borner-ISSI.pdf">http://cns.iu.edu/docs/publications/2017-borner-ISSI.pdf</a>

# Data Enclave at IU <a href="http://iuni.iu.edu/resources/web-of-science">http://iuni.iu.edu/resources/web-of-science</a>

### Web of Science

The IUNI Science of Science Hub acquired the complete set of Thomson Reuters' Web of Science XML raw data (Web of Knowledge version 5). This data can be used by any employee of Indiana University for academic research and without any sharing of data. Data details and information on how to access the data are given below.

### **Data Details**

- 1. WOK XML ONE TIME DELIVERY: Science Citation Index Expanded from 1900-2013
- 2. WOK XML ONE TIME DELIVERY: Social Sciences Citation Index from 1900-2013
- 3. WOK XML ONE TIME DELIVERY: Arts & Humanities Citation Index from 1975-2013
- 4. WOK XML ONE TIME DELIVERY: Book Citation Index -- Science from 2005-2013
- 5. WOK XML ONE TIME DELIVERY: Book Citation Index -- Social Sciences & Humanities from 2005-2013
- 6. WOK XML ONE TIME DELIVERY: Conference Proceedings Citation Index -- Science & Technical from 1990-2013

### Documentation

- Entity Relationship Diagram (pdf)
- Web of Knowledge Schemas (pdf)
- Data Dictionary (pdf)
- Web of Knowledge User Guide (pdf)

### **Data Access**

The data can be accessed as RAW XML data or in a Postgresql database via a Karst Data Enclave. Custom datasets can be requested in standard data formats (.csv..net).

To request data access, please complete the **IUNI Web of Science Data (WoS) Access Request Form** 

# **Open Code**

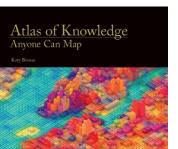


http://sci2.cns.iu.edu

#### **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 Learning decides being b 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
<ul> <li>categorize/cluster</li> <li>order/rank/sort</li> <li>distributions (also outliers, gaps)</li> <li>comparisons</li> <li>trends (process and time)</li> <li>geospatial</li> <li>compositions (also of text)</li> <li>correlations/relationships</li> </ul>	<ul> <li>nominal</li> <li>ordinal</li> <li>interval</li> <li>ratio</li> </ul>	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

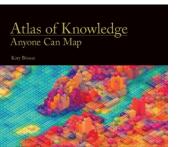


# Visualization Framework

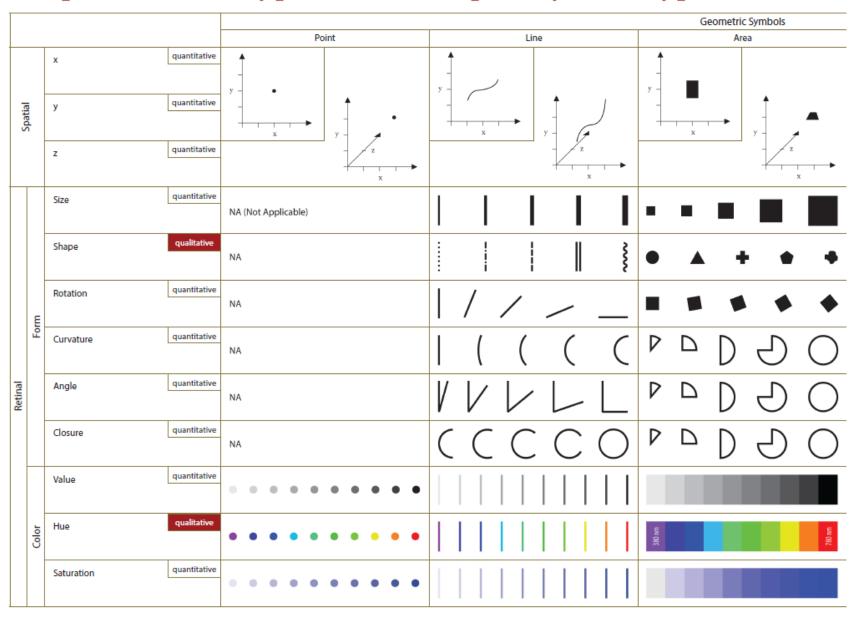
Basic Task Types								
Bertin, 1967	Wehrend & Lewis, 1996	Few, 2004	Yau, 2011	Rendgen & Wiedemann, 2012	Frankel, 2012	Tool: Many Eyes	Tool: Chart Chooser	Börner, 2014
selection	categorize			category				categorize/ cluster
order	rank	ranking					table	order/rank/ sort
	distribution	distribution					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
<ul> <li>categorize/cluster</li> <li>order/rank/sort</li> <li>distributions (also outliers, gaps)</li> <li>comparisons</li> <li>trends (process and time)</li> <li>geospatial</li> <li>compositions (also of text)</li> <li>correlations/relationships</li> </ul>	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



# 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

Sec.

detim

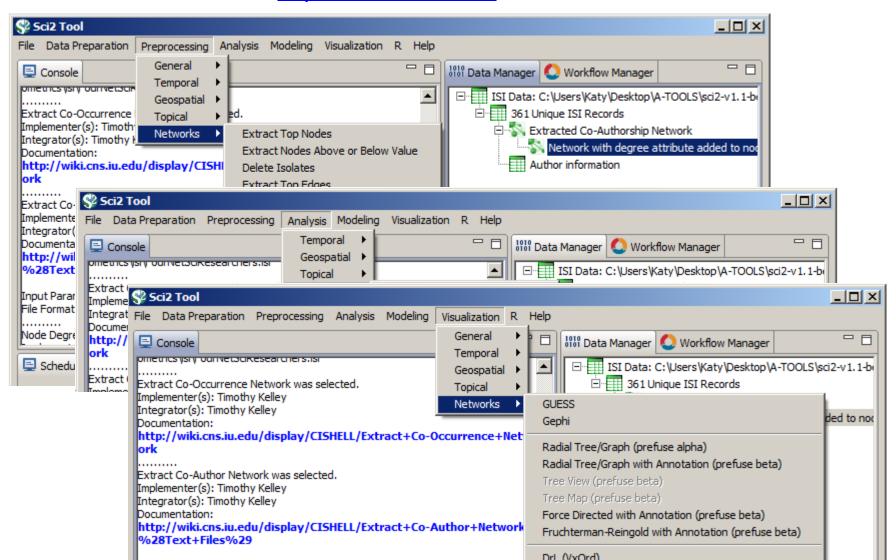
Blinking paint

feet.



# Sci2 Tool Menu

Download tool for free at <a href="http://sci2.cns.iu.edu">http://sci2.cns.iu.edu</a>

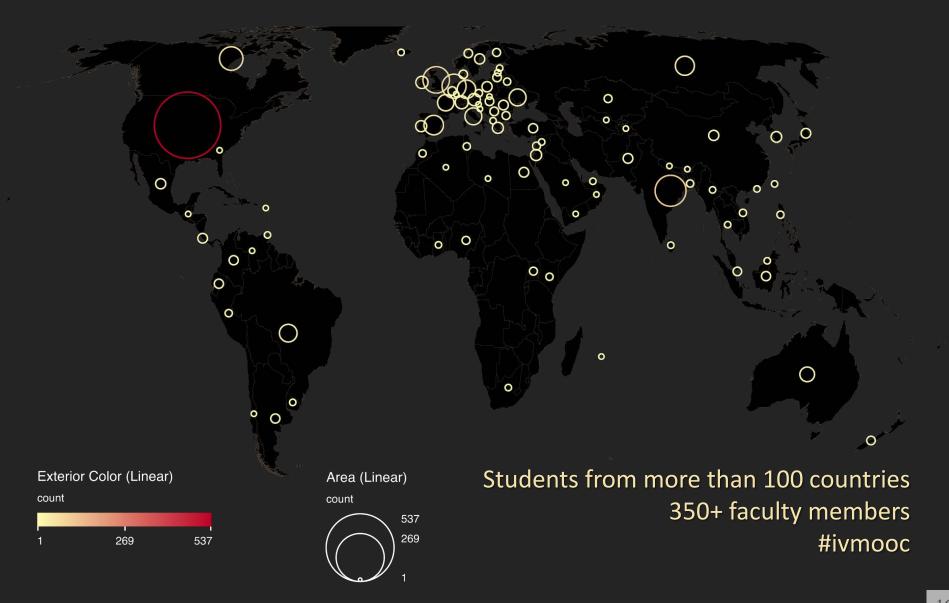


# **Open Education**



Register for free: <a href="http://ivmooc.cns.iu.edu">http://ivmooc.cns.iu.edu</a>. Class restarts Jan 9, 2018.

# The Information Visualization MOOC ivmooc.cns.iu.edu





SCWS Session: Visualizing STEAM Data in Support of Smart Decision Making November 15-17, 2017, Tokyo, Japan. <a href="http://scws2017.org">http://scws2017.org</a>



# **Upcoming Colloquia**

Unless otherwise indicated, most Sackler colloquia are held at the Arnold and Mabel Beckman Center, in Irvine, California.

### Science of Science Communication III

November 16-17, 2017; Washington, D.C.

Organized by Karen Cook, Baruch Fischhoff, Alan I. Leshner, and Dietram A. Scheufele

Register Now **→** 

Sign up to receive information about live webcast here

### Modeling and Visualizing Science and Technology Developments

December 4-5, 2017; Irvine, CA

Organized by Katy Börner, William Rouse, H. Eugene Stanley, and Paul Trunfio

Register Now **>>** 

### Economics, Environment, and Sustainable Development

January 17-18, 2018: Irvine, CA

Organized by: Simon Levin, Stephen Carpenter, Gretchen Daily, Sir Partha Dasgupta, Paul Ehrlich, Geoffrey Heal, Catherine Kling, Jane Lubchenco, and Stephen Polasky

http://www.nasonline.org/programs/sackler-colloquia/upcoming-colloquia