

# Registering, Visualizing, and Exploring Biomedical Data

Katy Börner @katycns

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



*NIMH Workshop on Advanced Statistical Methods and Dynamic Data Visualizations for Mental Health Studies*

July 30, 2021



# Overview

Mapping Science: An Exhibit

Mapping SPOKE: 3M Nodes and 30M Edges

HuBMAP: Toward a Human Reference Map

Data Visualization Literacy Framework

Empower Yourself!

# Mapping Science Exhibit

<http://scimaps.org>







101st Annual Meeting of the Association of American Geographers, Denver, CO.  
April 5th - 9th, 2005 (First showing of Places & Spaces)



University of Miami, Miami, FL.  
September 4 - December 11, 2014.



The David J. Sencer CDC Museum, Atlanta, GA.  
January 25 - June 17, 2016.



Duke University, Durham, NC.  
January 12 - April 10, 2015

<http://scimaps.org>



# Places & Spaces: Mapping Science Exhibit

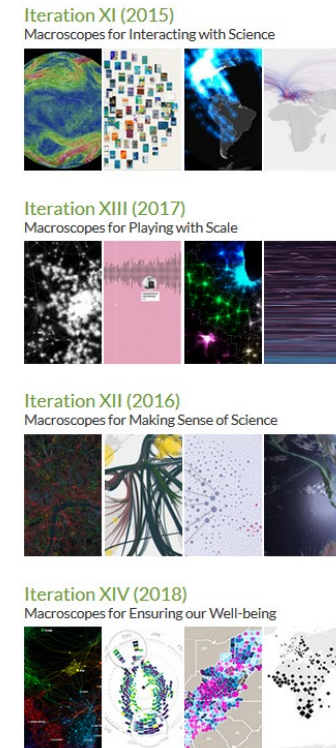
1<sup>st</sup> Decade (2005-2014)

## Maps



2<sup>nd</sup> Decade (2015-2024)

## Macroscopes

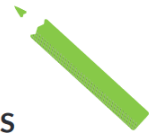


100

MAPS  
in large format, full color, and  
high resolution.

248

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



43



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

20

MACROSCOPES  
for touching all kinds of data.

382

DISPLAY VENUES  
from the Cannes Film Festival  
to the World Economic Forum.

354



PRESS ITEMS  
including articles in *Nature*,  
*Science*, *USA Today*, and *Wired*.

<http://scimaps.org>

# Map of Scientific Collaborations from 2005-2009



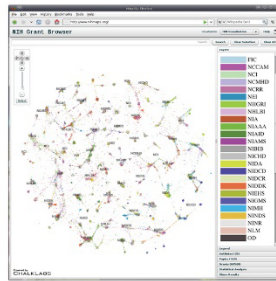
Computed Using Data from Elsevier's Scopus



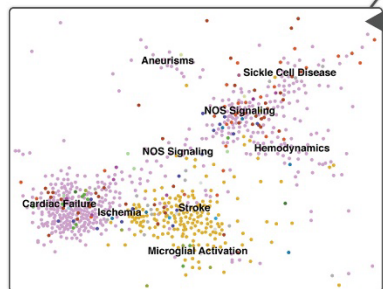
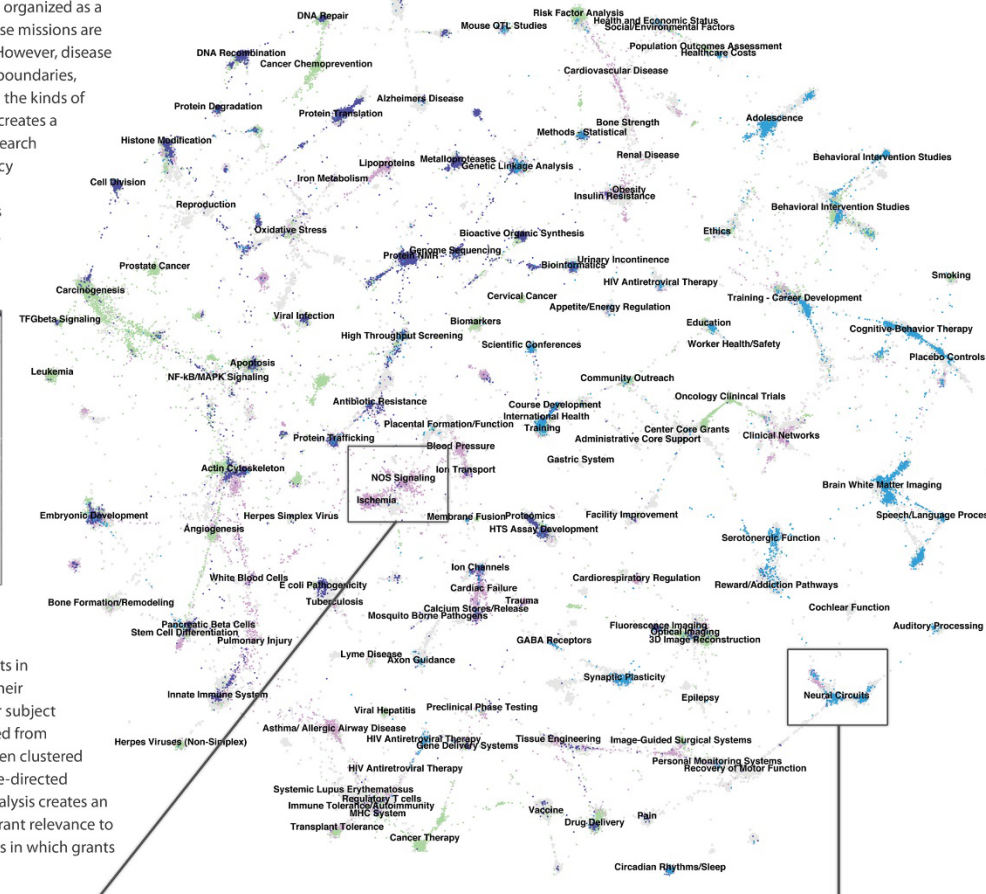
# A Topic Map of NIH Grants 2007

Bruce W. Herr II (Chalklabs & IU), Gully Burns (ISI), David Newman (UCI), Edmund Talley (NIH)

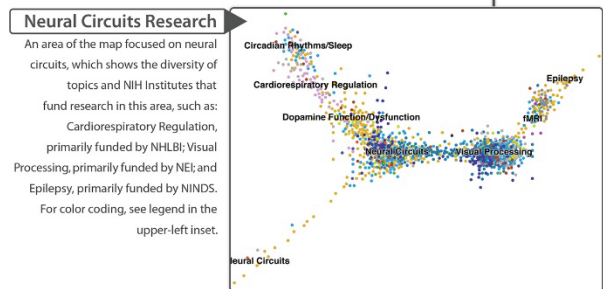
The National Institutes of Health (NIH) is organized as a multitude of Institutes and Centers whose missions are primarily focused on distinct diseases. However, disease etiologies and therapies flout scientific boundaries, and thus there is tremendous overlap in the kinds of research funded by each Institute. This creates a daunting landscape for decisions on research directions, funding allocations, and policy formulations. Shown here is devised an interactive topic map for navigating this landscape, online at [www.nihmaps.org](http://www.nihmaps.org). Institute abbreviations can be found at [www.nih.gov/icd](http://www.nih.gov/icd).



Topic modeling, a statistical technique that automatically learns semantic categories, was applied to assess projects in terms used by researchers to describe their work, without the biases of keywords or subject headings. Grant similarities were derived from their topic mixtures, and grants were then clustered on a two-dimensional map using a force-directed simulated annealing algorithm. This analysis creates an interactive environment for assessing grant relevance to research categories and to NIH Institutes in which grants are localized.



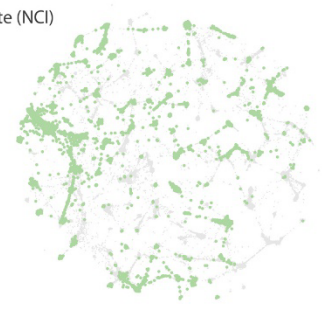
**Cardiac Diseases Research**  
An area of the map focused on cardiovascular function and dysfunction. Cardiac Failure (primarily funded by NHLBI) is typically clustered next to Stroke (NINDS), since these are the two major medical emergencies associated with ischemia, which results from a restricted blood supply. Also localized in this area are grants focused on Nitric Oxide (NOS) Signaling, a major biochemical pathway for vasodilation, and grants on Hemodynamics, Sickle Cell Disease, and Aneurysms.



**Neural Circuits Research**  
An area of the map focused on neural circuits, which shows the diversity of topics and NIH Institutes that fund research in this area, such as: Cardiorespiratory Regulation, primarily funded by NHLBI; Visual Processing, primarily funded by NEI; and Epilepsy, primarily funded by NINDS. For color coding, see legend in the upper-left inset.

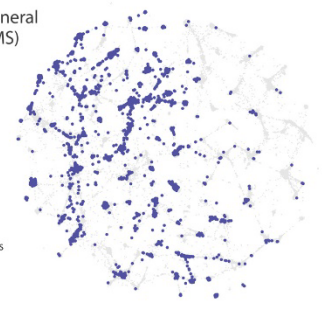
## National Cancer Institute (NCI)

- TOP 10 TOPICS
- 1 Oncology Clinical Trials
  - 2 Cancer Treatment
  - 3 Cancer Therapy
  - 4 Carcinogenesis
  - 5 Risk Factor Analysis
  - 6 Cancer Chemotherapy
  - 7 Metastasis
  - 8 Leukemia
  - 9 Prediction/Prognosis
  - 10 Cancer Chemoprevention



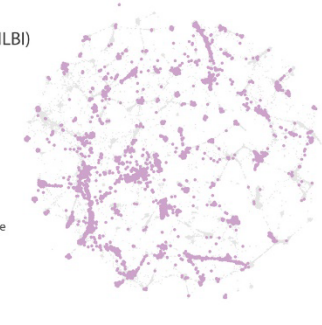
## National Institute of General Medical Sciences (NIGMS)

- TOP 10 TOPICS
- 1 Bioactive Organic Synthesis
  - 2 X-ray Crystallography
  - 3 Protein NMR
  - 4 Computational Models
  - 5 Yeast Biology
  - 6 Metalloproteases
  - 7 Enzymatic Mechanisms
  - 8 Protein Complexes
  - 9 Invertebrate/Zebrafish Genetics
  - 10 Cell Division



## National Heart, Lung, and Blood Institute (NHLBI)

- TOP 10 TOPICS
- 1 Cardiac Failure
  - 2 Pulmonary Injury
  - 3 Genetic Linkage Analysis
  - 4 Cardiovascular Disease
  - 5 Atherosclerosis
  - 6 Hemostasis
  - 7 Blood Pressure
  - 8 Asthma/ Allergic Airway Disease
  - 9 Gene Association
  - 10 Lipoproteins



## National Institute of Mental Health (NIMH)

- TOP 10 TOPICS
- 1 Mood Disorders
  - 2 Schizophrenia
  - 3 Behavioral Intervention Studies
  - 4 Mental Health
  - 5 Depression
  - 6 Cognitive-Behavior Therapy
  - 7 AIDS Prevention
  - 8 Genetic Linkage Analysis
  - 9 Adolescence
  - 10 Childhood



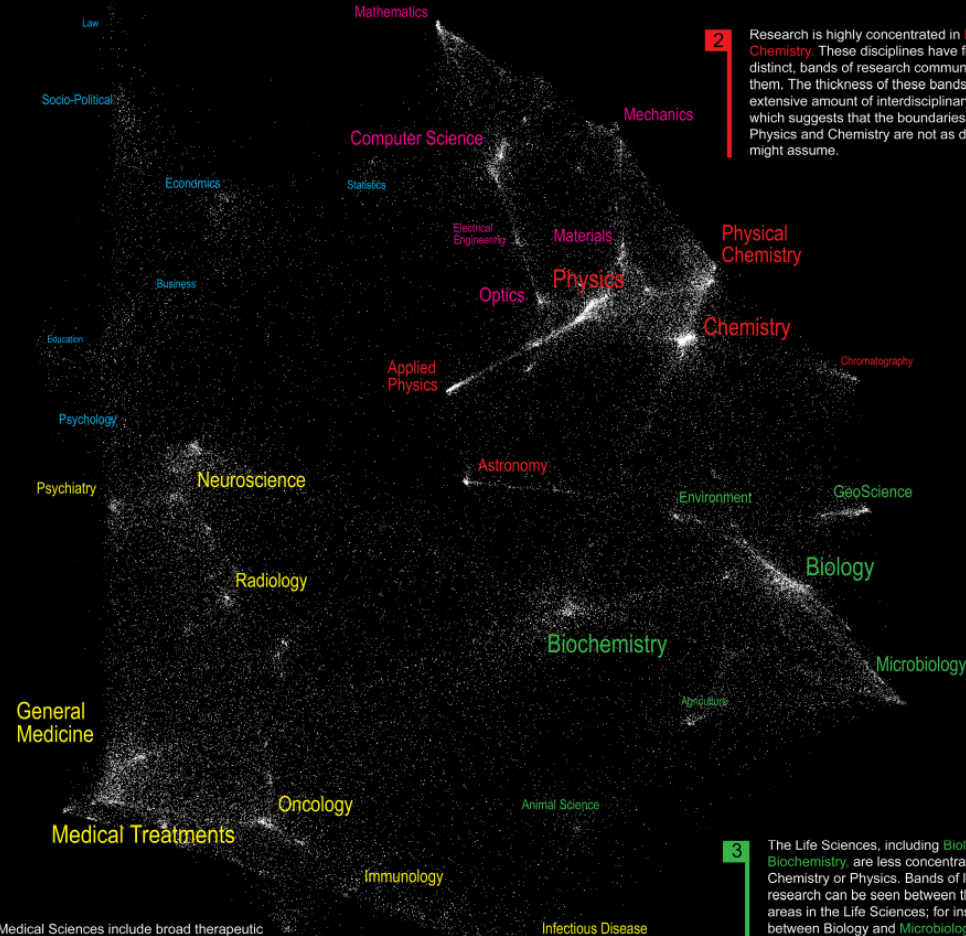


# The Structure of Science

**5** 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.

**1** **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. Although applied, these disciplines are highly concentrated with distinct bands of research communities that link them. Bands indicate interdisciplinary research.

**2** Research is highly concentrated in **Physics** and **Chemistry**. 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 might assume.



**4** The Medical Sciences include broad therapeutic studies and targeted areas of **Treatment** (e.g. central nervous system, cardiology, gastroenterology, etc.) Unlike Physics and Chemistry, the medical disciplines are more spread out, suggesting a more multi-disciplinary approach to research. The transition into Life Sciences (via Animal Science and Biochemistry) is gradual.

**3** 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 between Biology and **Environmental Science**. Biochemistry is very interesting in that it is a large discipline that has visible links to disciplines in many areas of the map, including Biology, Chemistry, Neuroscience, and General Medicine. It is perhaps the most interdisciplinary of the sciences.

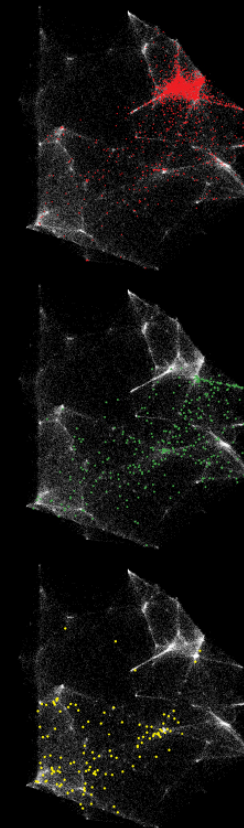
We are all familiar with traditional maps that show the relationships between countries, provinces, states, and cities. Similar relationships exist between the various disciplines and research topics in science. This allows us to map the structure of science.

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.

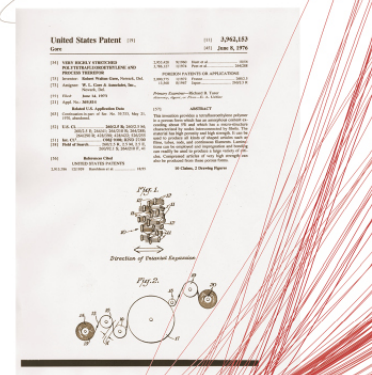
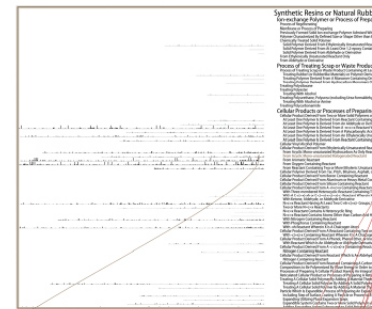


# Impact

The United States Patent and Trademark Office does scientists and industry a great service by granting patents to protect inventions. Inventions are categorized in a taxonomy that groups patents by industry or use, proximate function, effect or product, and structure. At the time of this writing there are 160,523 categories in a hierarchy that goes 15 levels deep. We display the first three levels (13,529 categories) at right in what might be considered a textual map of inventions.

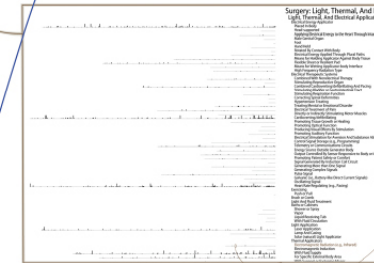
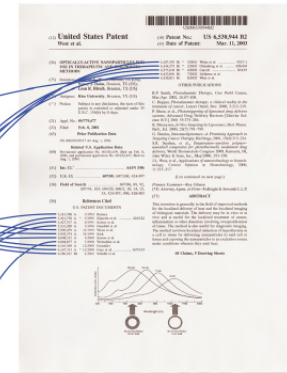
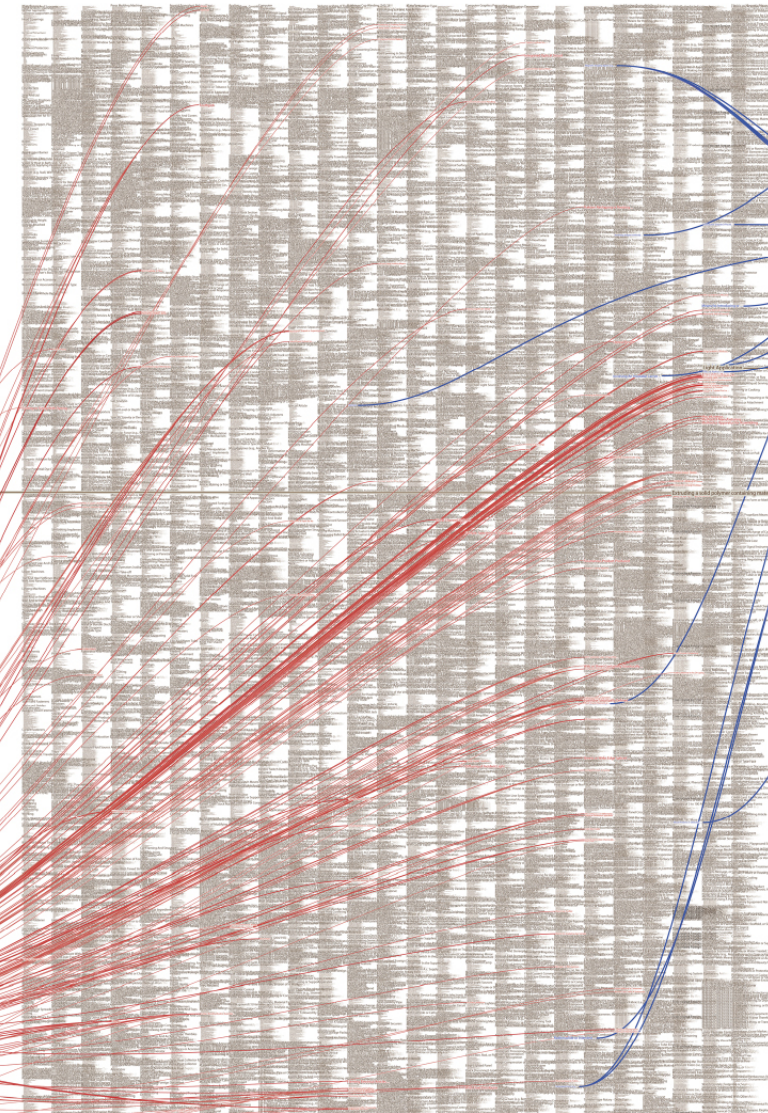
Patent applications are required to be unique and non-obvious, partially by revealing any previous patents that might be similar in nature or provide a foundation for the current invention. In this way we can trace the impact of a single patent, seeing how many patents and categories it affects.

The patent on Goretex—a lightweight, durable synthetic fiber—is an example of one that has had significant impact. The box below enlarges the section of the hierarchy where it is filed, and the red lines (arranged to start along a time line from 1981 to 2006) point to the 130 categories that contain 182 patents, from waterproof clothing to surgical cosmetic implants, that mention Goretex as "prior art."



# The US Patent Hierarchy

# Prior Art



New patents often build on older ideas from many different categories. Here, blue lines originate in the sixteen categories that contain patents cited as prior art for a patent on "gold nanoshells." Gold nanoshells are a new invention: tiny gold spheres (with a diameter ten million times smaller than a human hair) that can be used to make tumors more visible in infrared scans; they have even helped cause complete remission of tumors in tests with laboratory mice. The blue lines show that widely separated categories provided background for this invention.

Keeping categories understandable is an important part of maintaining any taxonomy, including the patent hierarchy. Categories are easier to understand, search, and maintain if they contain elements that comfortably fit the definition of the category. The box above shows tiny bar charts, part of a *Taxonomy Validator* that reveals whether elements fit their categories. Categories may need to be redefined, and sometimes need to be split when they get too vague or large; a problem shared by many classification systems in this information-rich century. But how can we tell which ones to eliminate, add or revise—or how to revise them—in the complex, abstract sociolinguistic spaces we partition into ontologies?

Something as simple as a bar chart helps people see how entities in a category relate to that category. Here, each bar encodes a "distance to prototype": how much each patent differs from an idealized "prototype patent" for that category. A measure like this can be based on statistics, computational linguistics, or even human insight. Thus a category with mostly small bars is a good one, and a generally ragged one needs scrutiny or reorganization; but one that has only two or three tall bars may mean that only those few elements don't belong.

Even simple visuals can make thinking easier by providing better distilled data to the eye: vastly more data than working memory can hold as words. They focus people on exactly the right issues, and support them with the comprehensive overviews they need to make more informed judgements.



# Science related Wikipedian ACTIVITY

This visualization explores the activity of science, math, and technology (SMT) related articles in the English-language Wikipedia (<http://en.wikipedia.org>). The central image shows 659,388 articles (circles). Overlaid is a 37 x 37 grid of relevant half-inch sized images.

Blue, green, and yellow circles represent the 3,599 math, 6,474 science, and 3,164 technology related articles respectively. The larger the size of a circle the higher the likelihood it is that type of article. The four corners show activity patterns of SMT articles.

**Article Edit Activity**  
Articles are size coded based on how frequently they have been edited from Feb. 6, 2001 to April 6, 2007. More consideration is given to current and major edits. Larger circles have been edited more frequently than smaller circles.

**2007 Major Edits**  
Articles are size coded based on how many major edits they received from January 1st, 2007 to April 6th, 2007. Larger circles have received more edits than smaller circles. The highest number of major edits was 2,627.

For the central image, each article is size coded based on the likelihood that it is math, science, or technology related.

- 0%
- 50%
- 100%

All five images are color coded based on type. Transparency is used for legibility, and creates different colors when nodes overlap.

- Math
- Science
- Technology

**Article Popularity**  
Articles are size coded based on the number of Wikipedia articles referencing it. Larger circles are receiving more links from other articles than smaller circles. The highest number of references to an article was 142,602.

**Number of Bursts**  
Articles are size coded based on the number of bursts, i.e. sudden increases, of edit activity that occurred during the article's lifetime. Larger circles have had more bursts in activity than smaller circles. The most bursts an article had was 9.





# Diseasome

## The Human Disease Network

Explore online at <http://diseasome.eu>

### Statistics

# of Nodes: 516  
 # of Edges: 1188  
 Density: 0,0089  
 Average Degree: 9,20  
 Diameter: 15  
 Average Shortest Path: 6,5

### Disorder Class

- Cancer
- Endocrine
- Ear, Nose, Throat
- Ophthalmological
- Neurological
- Hematological
- Cardiovascular
- Muscular
- Immunological
- Dermatological
- Nutritional
- Connective Tissue Disorder
- Renal
- Psychiatric
- Metabolic
- Bone
- Skeletal
- Developmental
- Gastrointestinal
- Respiratory
- Multiple
- Unclassified

### Top 5 Diseases

1. Deafness
2. Leukemia
3. Colon Cancer
4. Retinitis Pigmentosa
5. Diabetes Mellitus

### Top 5 Genes

1. TP53
2. PAX6
3. FGFR2
4. RTN
5. MSH2

### Description

The map presents a network of 516 diseases linked by 1188 known disorder-gene associations, indicating the common genetic origin of many diseases.

#### GENE NETWORK CLUES

This map offers a rapid visual reference of the genetic links between disorders and a valuable global perspective for physicians, genetic counselors, and biomedical researchers alike. This view appears only when the network is zoomed, revealing to their associated genes, together the understanding of the roots of disease, and the functions of particular genes.

#### NETWORK VISUALIZATION TECHNIQUES APPLIED

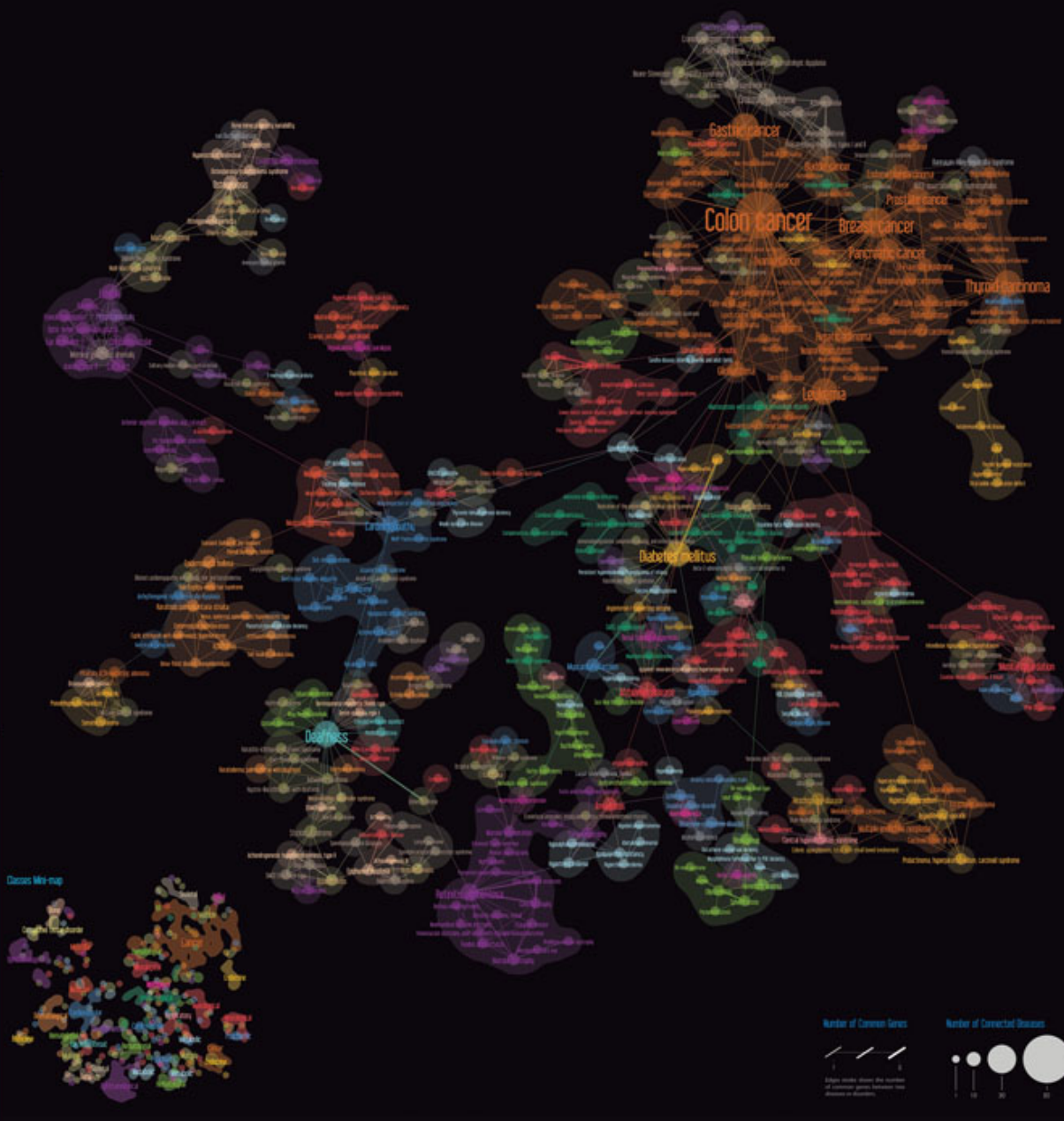
The map was done using the force-directed layout algorithm ForceAtlas in Gephi. Node sizes correspond to the disorder class to which the disease belongs, and the size is proportional to its node degree, the overall number of links. Link's width is proportional to the number of genes that are implicated in both disorders and colored with the average color between source and target nodes. Isolated diseases are not shown and only the giant component has been kept. The Clusters Mini-map shows more readable disorder classes and shows largest visual clusters.

The Disorder Class Interactions graph below shows the interaction level between disorder classes, representing the number of shared genes, up to 80.

#### References

The Human Disease Network  
 Bastin & Heymann 2009, *PLoS ONE*, 4(10): e7000

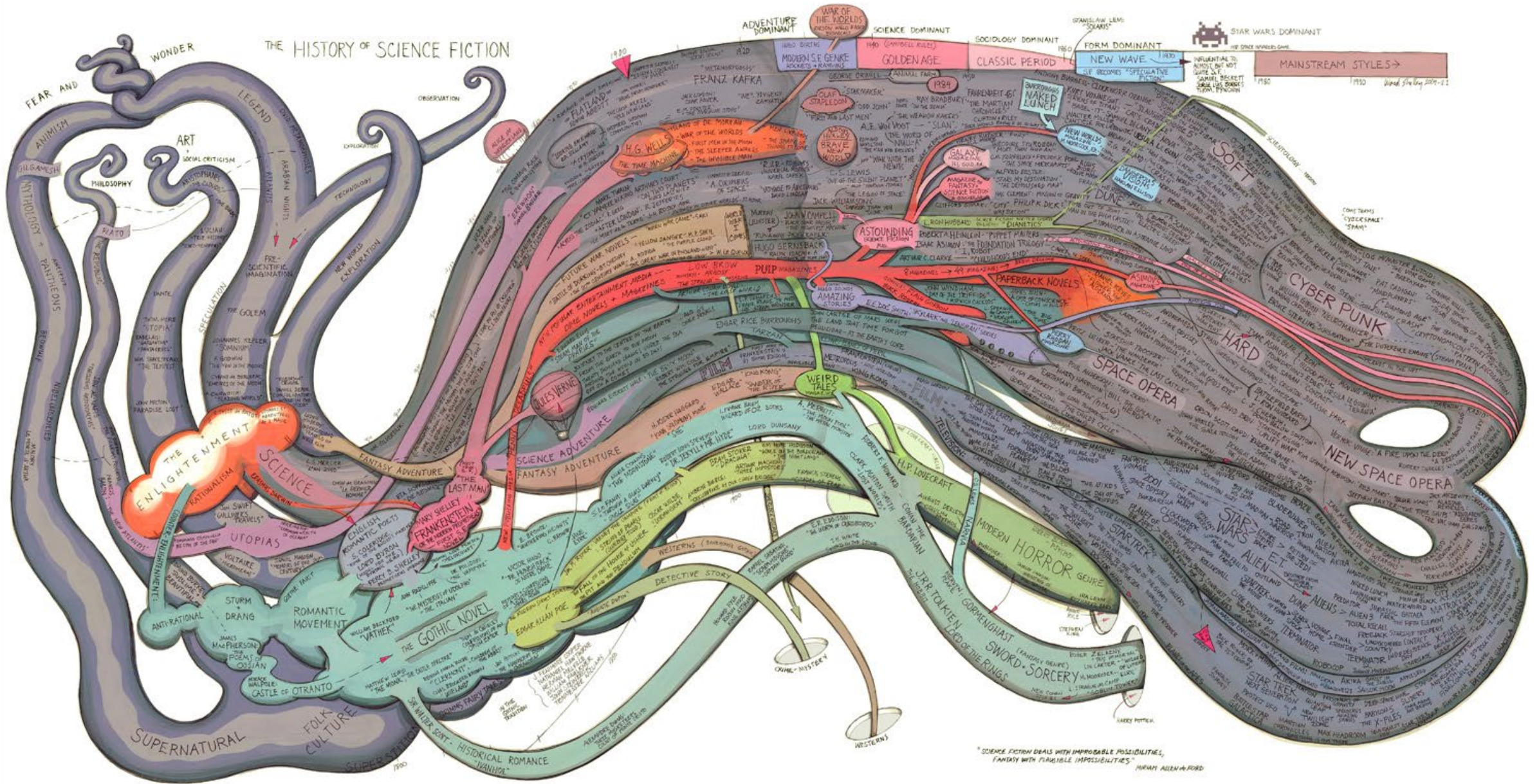
### Disorder Class Interactions



Disorder Mini-map







VII.10 History of Science Fiction - Ward Shelley - 2011



# Check out our **Zoom Maps** online!



VII.10  
History of Science Fiction, by Ward Shulley

BROOKLYN, NY, 2011  
Courtesy of Ward Shulley Studio

Ward Shulley is an artist identified with the Williamsburg scene in Brooklyn, New York. He is a writer, artist, and curator. This map plots the science fiction literary genre from its nascent beginnings in the late 18th century, through the 19th-century Romanticism, to the 20th-century modernism and postmodernism. The map is a complex network of colored lines and text, representing the evolution of science fiction literature. It includes various sub-genres like 'SCIENCE ADVENTURE', 'FANTASY ADVENTURE', 'CYBER PUNK', and 'SPACE OPERA'. A zoomed-in inset shows a detailed section of the map focusing on 'ENGLISH ROMANTIC POETS' and 'SCIENCE ADVENTURE', featuring names like Mary Shelley, Frankenstein, and Jules Verne.

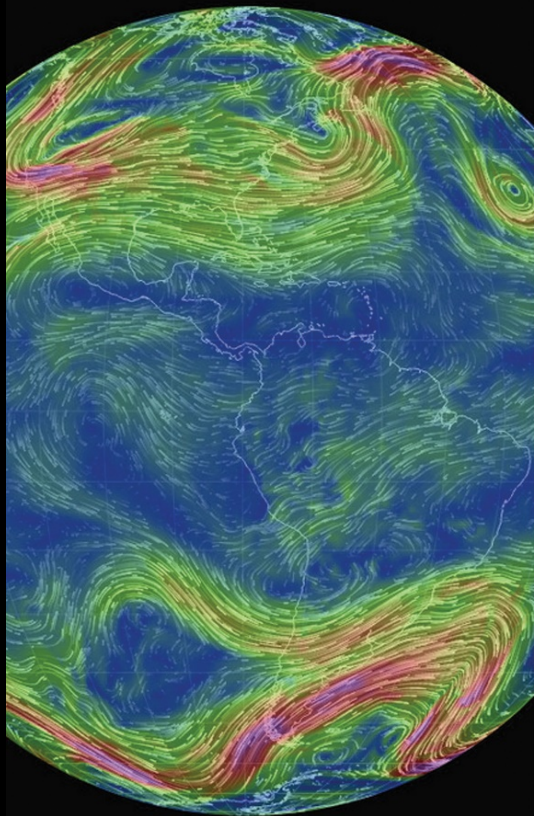
PLACES & SPACES  
MAPPING & DESIGN

Visit [scimaps.org](http://scimaps.org) and check out all our maps in stunning detail!





# MACROSCOPES FOR INTERACTING WITH SCIENCE



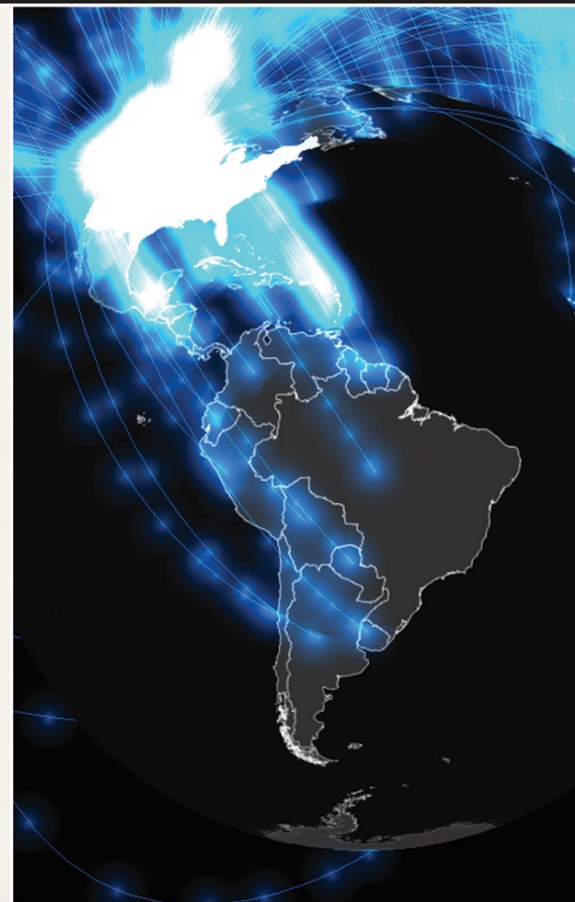
**Earth**

*Weather on a worldwide scale*



**AcademyScope**

*Exploring the scientific landscape*



**Mapping Global Society**

*Local news from a global perspective*

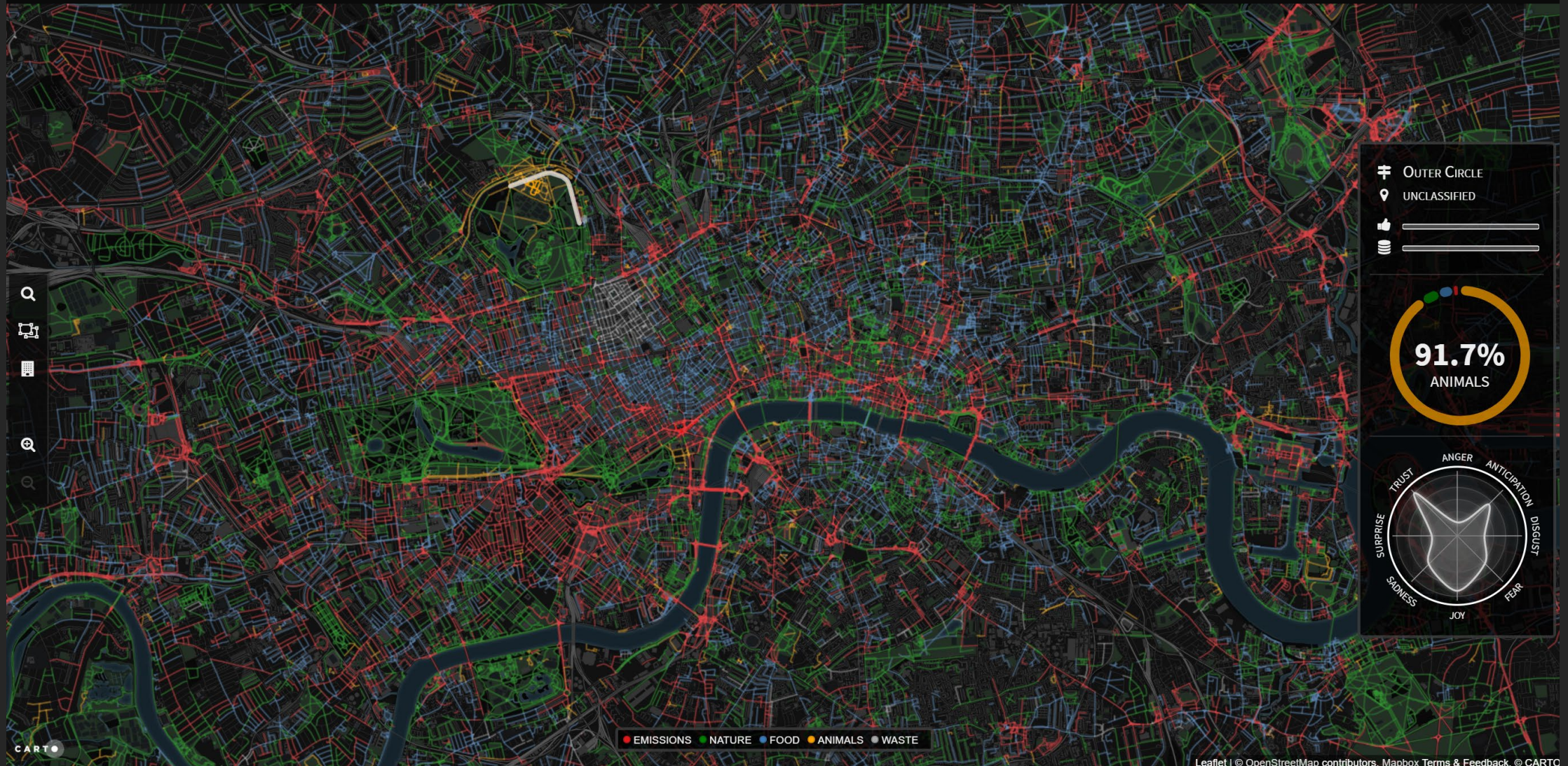


**Charting Culture**

*2,600 years of human history in 5 minutes*



SMELLY MAPS

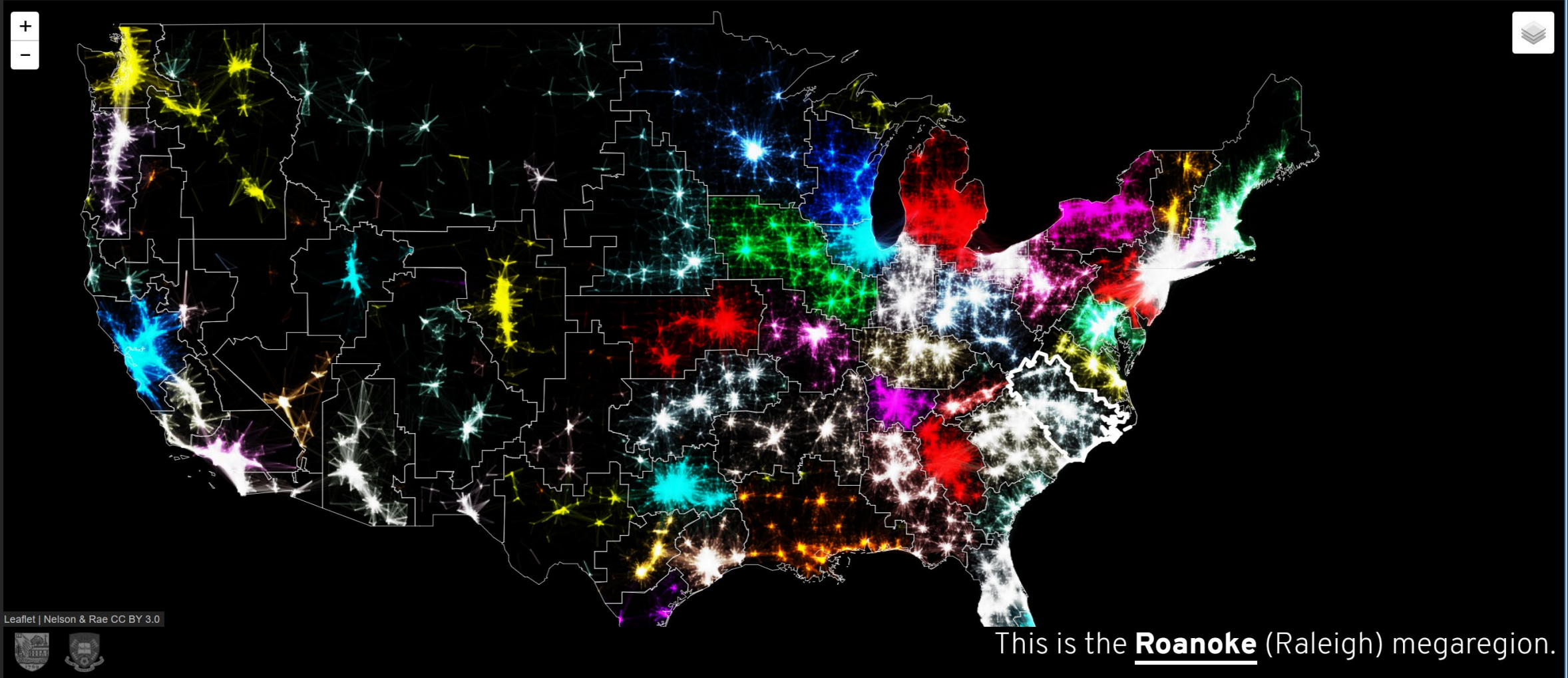


Smelly Maps – Daniele Quercia, Rossano Schifanella, and Luca Maria Aiello – 2015



# THE MEGAREGIONS OF THE US

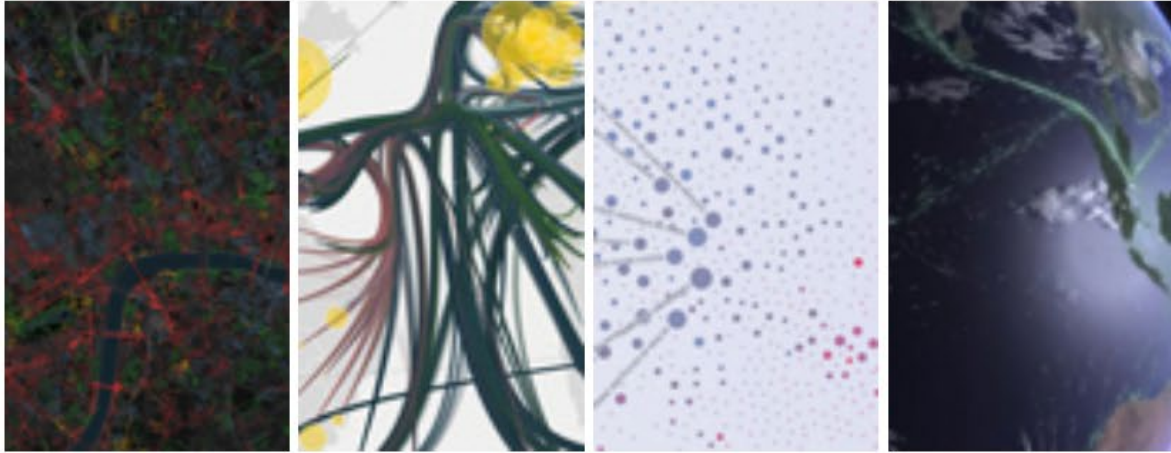
Explore the new geography of commuter connections in the US.  
Tap to identify regions. Tap and hold to see a single location's commuted.





## Iteration XII (2016)

Macrosopes for Making Sense of Science



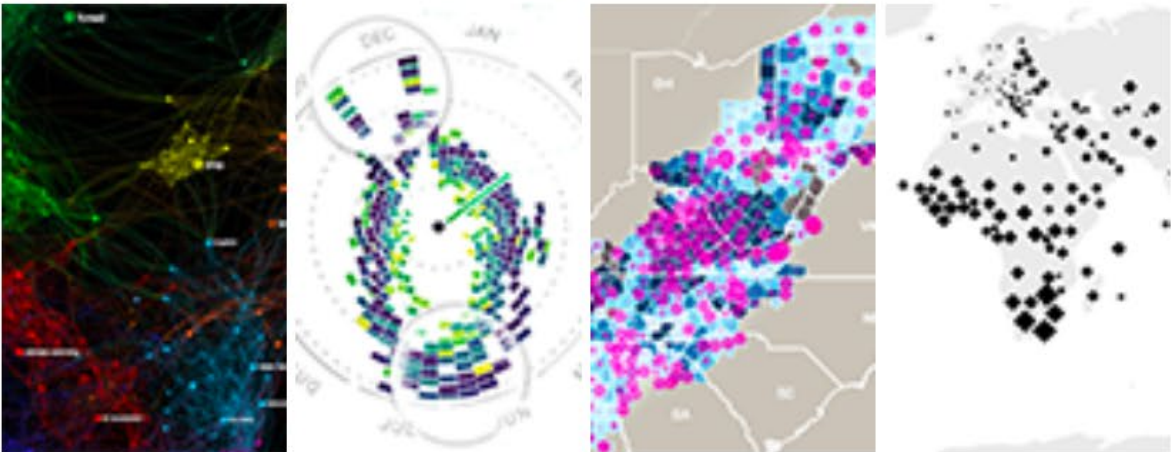
## Iteration XIII (2017)

Macrosopes for Playing with Scale



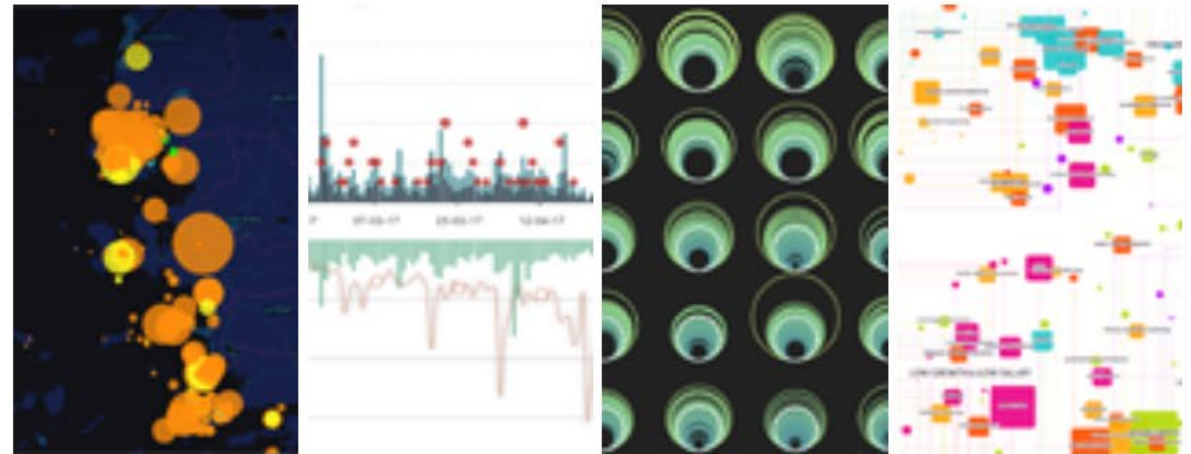
## Iteration XIV (2018)

Macrosopes for Ensuring our Well-being



## Iteration XV (2019)

Macrosopes for Tracking the Flow of Resources



# Acknowledgments

## Exhibit Curators



The exhibit team: Lisel Record, Katy Börner, and Todd Theriault.

<http://scimaps.org>

Plus, we thank the more than 250 authors of the 100 maps and 16 interactive macroscopes.

## Exhibit Advisory Board



**Gary Berg-Cross**  
Cognitive psychologist (PhD, SUNY-Stony Brook). Potomac, MD, USA



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Director of the **Advanced Visualization Laboratory** at the National Center for Supercomputing Applications, University of Illinois at Urbana-Champaign, IL, USA



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**Francis Harvey**  
Professor of Visual Communication in Geography at the Leibnitz Institute for Regional Geography, Leipzig University, Germany



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Professor, **The Graduate Center**, City University of New York; Director, **Software Studies Initiative** (big data, digital humanities, visualization)



**André Skupin**  
Associate Professor of Geography at San Diego State University, California



**Moritz Stefaner**  
Freelance designer on the crossroads of data visualization, information aesthetics, and user interface design in Germany



**Olga Subirós**  
Curator of Big Bang Data and Founder of **Olga Subirós Studio** in Barcelona, Spain



**Stephen Uzzo**  
Vice President of Science and Technology for the **New York Hall of Science**



**Benjamin Wiederkehr**  
Founding Partner and Managing Director of **Interactive Things** in Zürich, Switzerland



# Visualizations of the Scalable Precision Medicine Knowledge Engine (SPOKE)

<https://spoke.ucsf.edu>



Scalable Precision  
Medicine Knowledge  
Engine

Search... 

Data & Tools

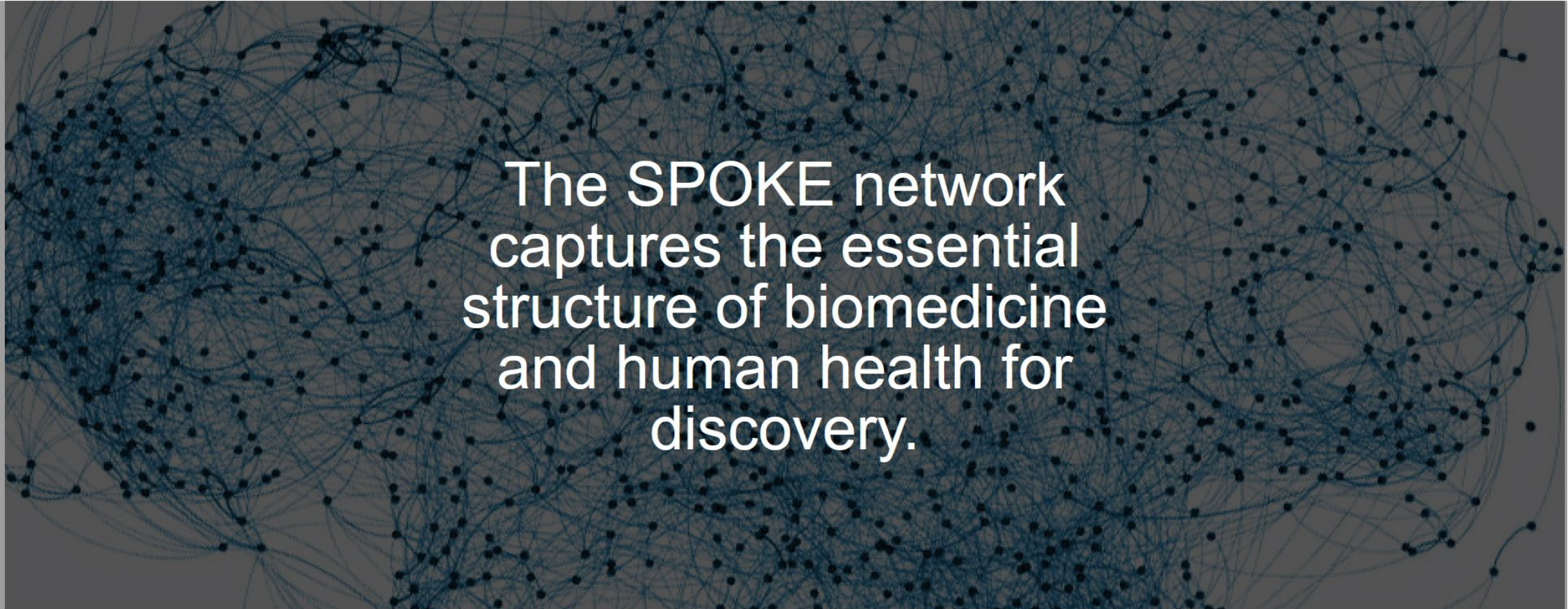
Neighborhood Explorer

Funding

Applications

People

Publications



The SPOKE network  
captures the essential  
structure of biomedicine  
and human health for  
discovery.

<https://spoke.ucsf.edu>



## Lead Investigators



**Sergio Baranzini, PhD**  
Principal Investigator



**Sui Huang, MD, PhD (ISB)**



**Sharat Israni, PhD**



**Mike Keiser, PhD**

## *SPOKE* investigative teams

The SPOKE team members are from the following organizations. *Team members listed below are from UCSF, except when indicated.*

- [Google](#)
- [Indiana University \(IU\)](#)
- [Institute for Systems Biology \(ISB\)](#)
- [Lawrence Livermore National Lab \(LLNL\)](#)
- [Stanford University](#)
- [University of California, San Diego \(UCSD\)](#)
- [University of California, San Francisco \(UCSF\)](#)

## Technical & Planning Team

**Rafael Gonçalves, PhD (Stanford)**

**Adil Harroud, MD**

**Elaine Meng, MD**

**Scoter Morris, PhD**

**Charlotte Nelson, PhD**

**Boris Oskotsky, PhD**

**Angela Rizk-Jackson, PhD**

**Peter Rose, PhD (UCSD)**

**Brett Smith (ISB)**

**Karthik Soman, PhD**

**Xiaoyuan Zhou, PhD**

## Collaborators

**Katy Börner, PhD (IU)**

**William Brown, PhD, DrPH**

**Ramanathan V. Guha, PhD (Google)**

**Mark Musen, MD, PhD (Stanford)**

**Camille Nebeker, EdD, MS (UCSD)**

**Roger Pearce, PhD (LNL)**





Scalable Precision  
Medicine Knowledge Engine

# Envisioning SPOKE: 3M Nodes and 30M Edges

The Scalable Precision Medicine Oriented Knowledge Engine (SPOKE) graph federates about 19 open datasets into a public data commons of health relevant knowledge. This site lets users explore the massive SPOKE knowledge graph.

The site was designed for two user groups: (1) novice users interested to understand the coverage and quality of SPOKE data and (2) expert users interested to analyze and optimize the interlinked knowledge graphs in SPOKE.

The overview visualization shows the different entity type and their diverse interlinkages. Detail

SPOKE is a fully interactive tool for exploring the interconnections between data.

[Explore SPOKE](#)



NIH National Center  
for Advancing  
Translational Sciences



CNS Cyberinfrastructure for  
Network Science Center





A:0 1 2 3 4 5 6 7 8 9 10 11 12 A

B B

C C

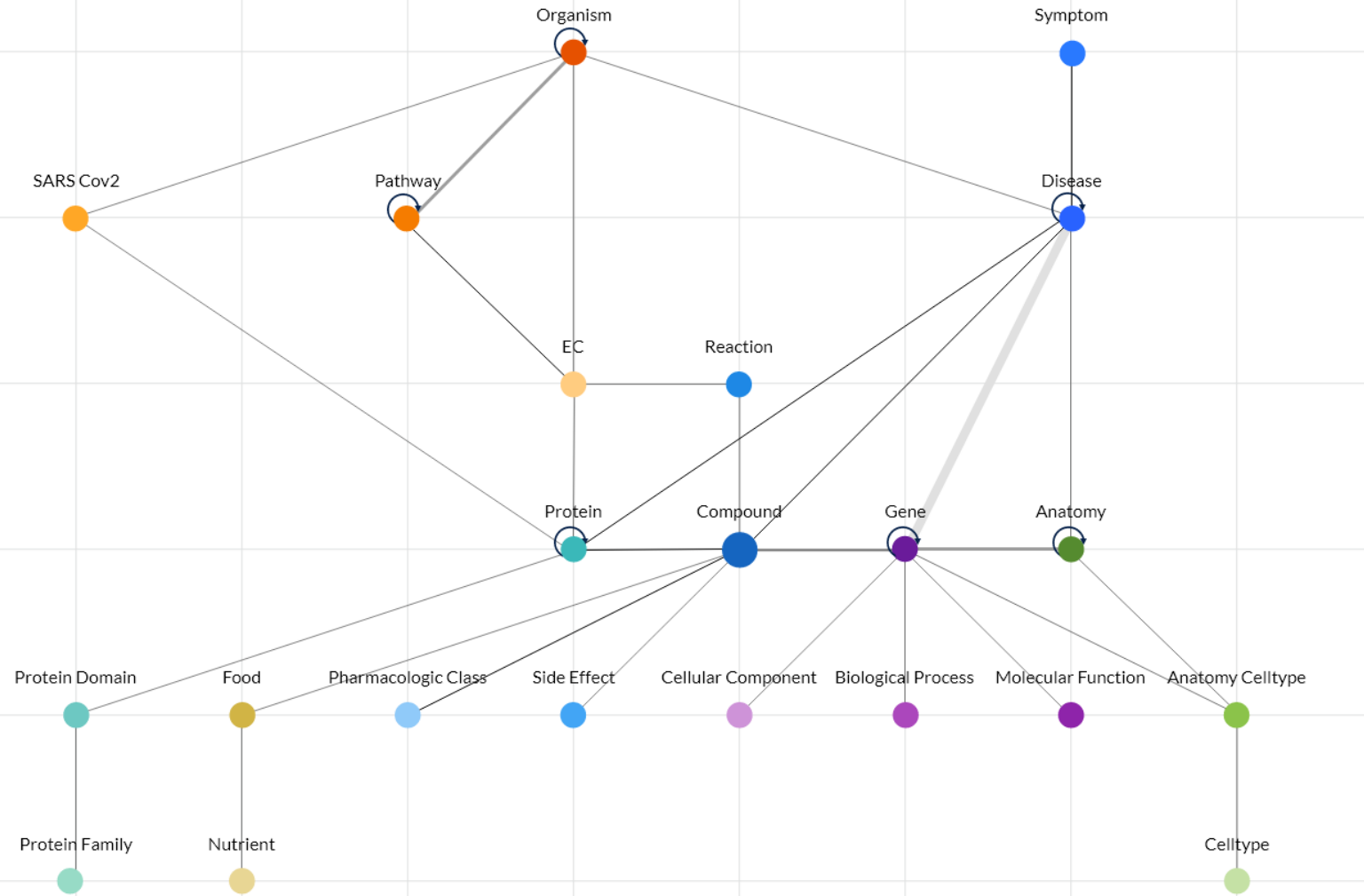
D D

E E

F F

G G

H H

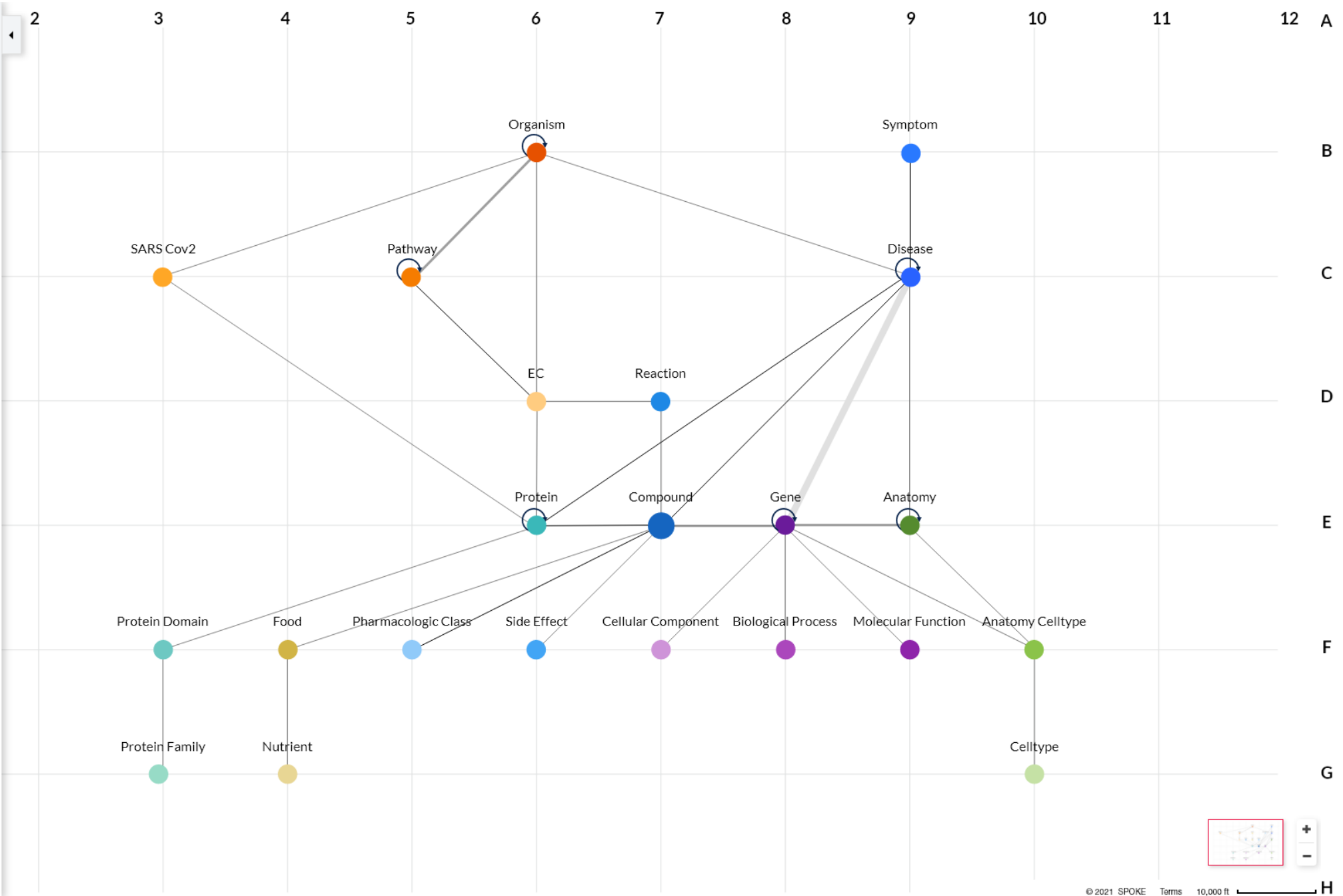




potato

heart

- alcoholic cardiomyopathy
- anterolateral myocardial infarction
- atrial fibrillation
- beta thalassemia
- brugada syndrome
- cardiac arrest
- cardiac tuberculosis
- coronary artery disease
- diabetic neuropathy
- diastolic heart failure
- ...

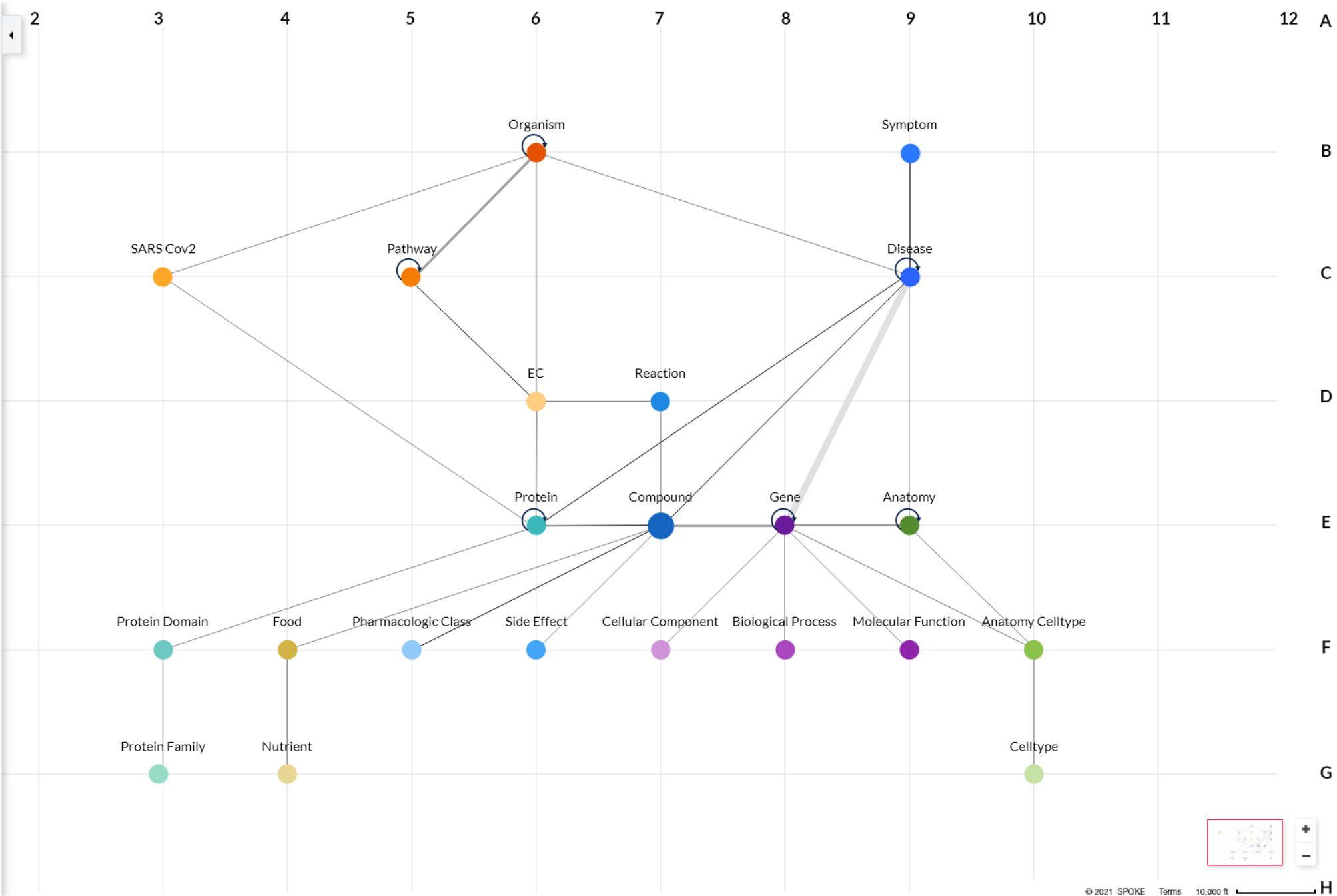


Send feedback





- potato
- heart
- alcoholic cardiomyopathy
- anterolateral myocardial infarction
- atrial fibrillation
- beta thalassemia
- brugada syndrome
- cardiac arrest
- cardiac tuberculosis
- coronary artery disease**
- diabetic neuropathy
- diastolic heart failure
- ...



Send feedback



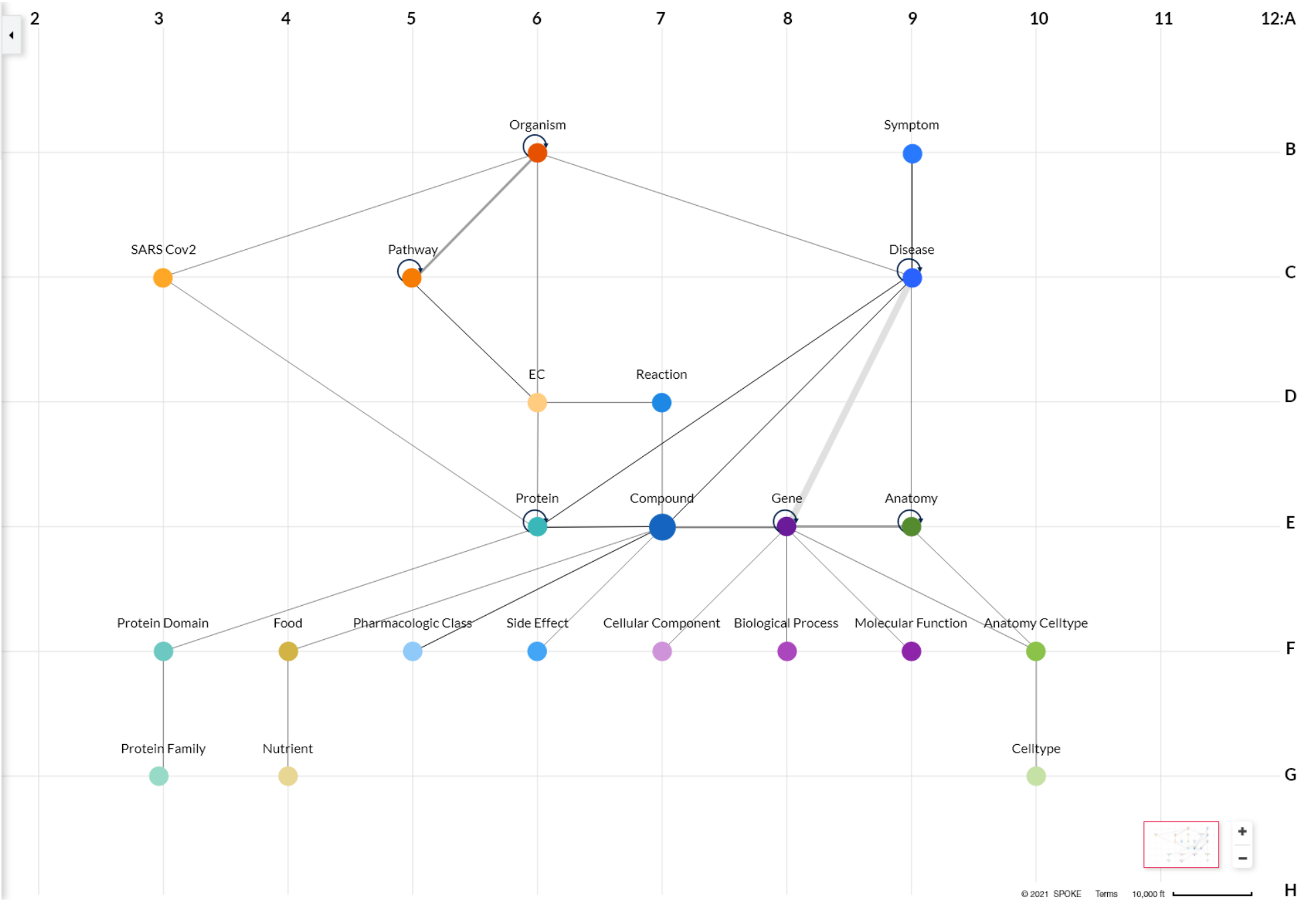
potato ✕

coronary artery disease ✕

SEARCH

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Show Details



Send feedback





potato ✕

coronary artery disease ✕

[SEARCH](#)

Favorite
Similar
Send to phone
Share

[Show Details](#)



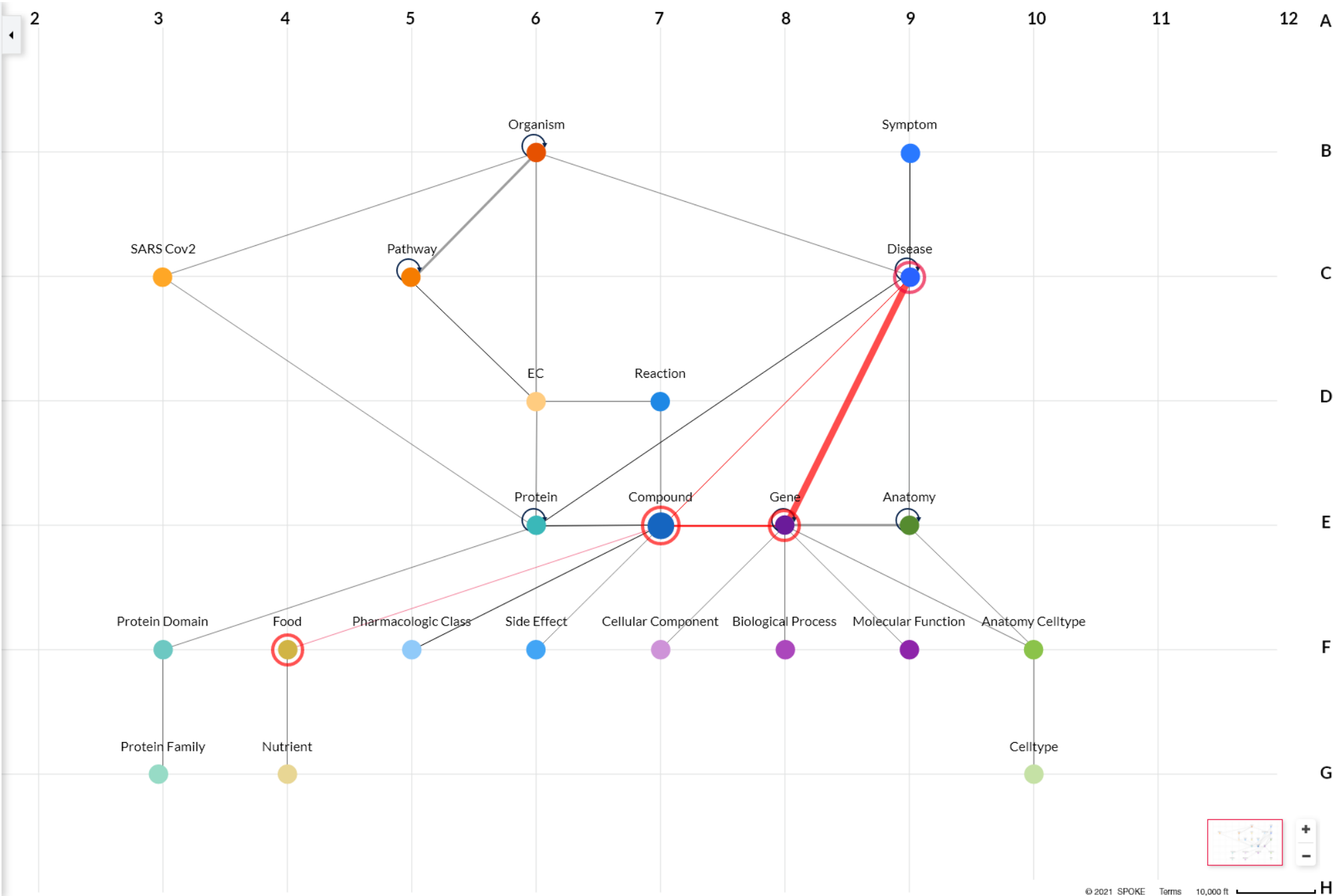
Send feedback

potato ✕

coronary artery disease ✕

SEARCH

Favorite Similar Send to phone Share



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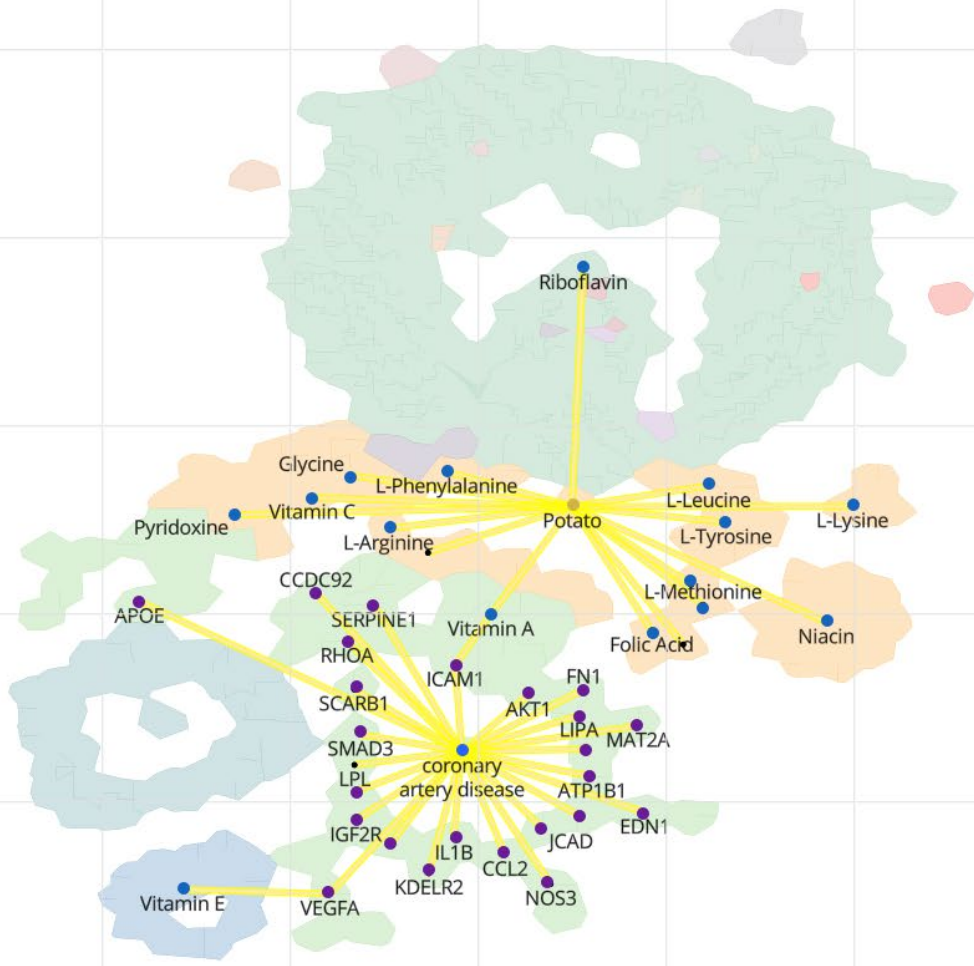
potato ✕

coronary artery disease ✕

SEARCH

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● Disease ● Compound ● Food ● Gene

Send feedback

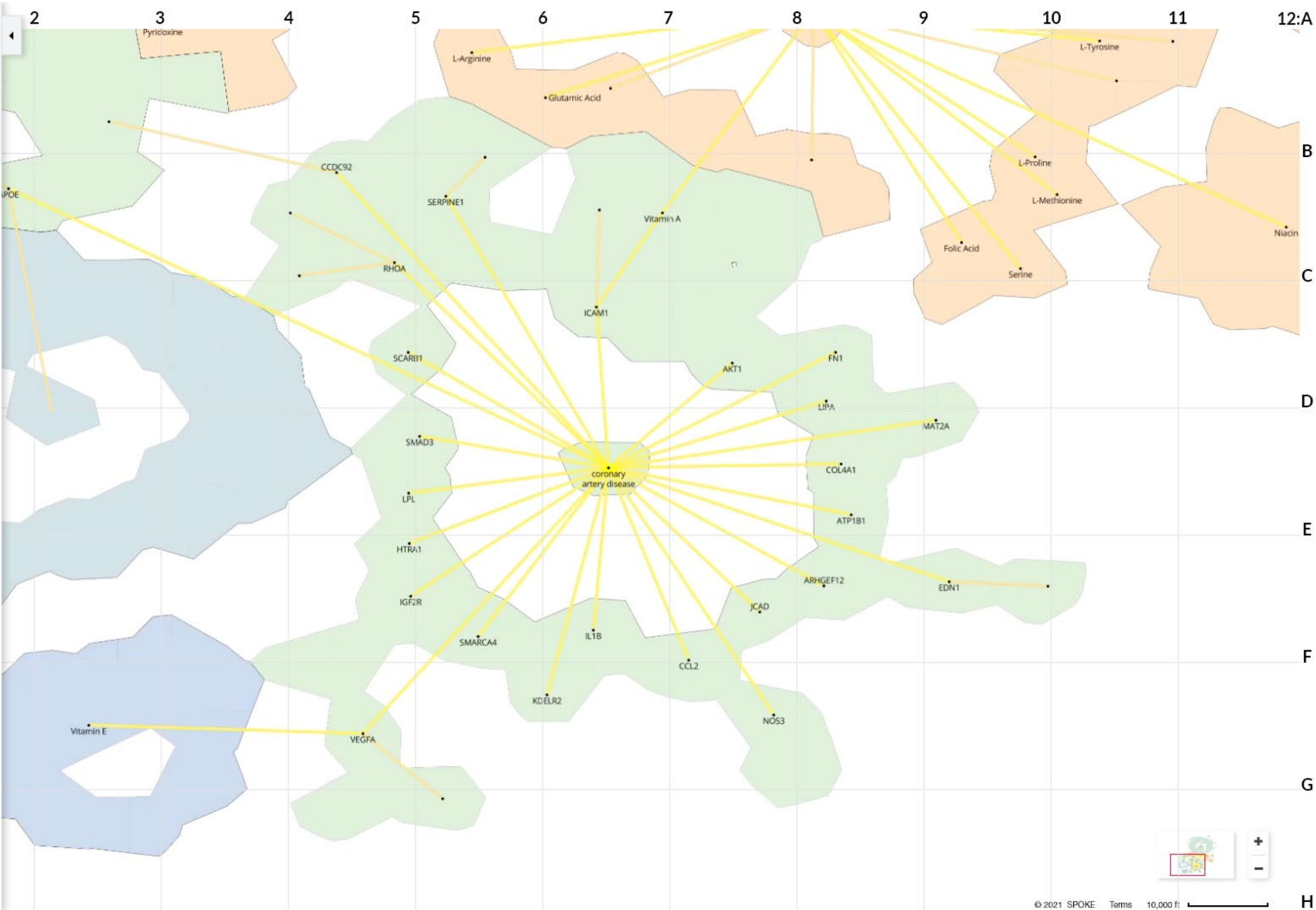
© 2021 SPOKE Terms 10,000 ft

potato ✕

coronary artery disease ✕

Favorite Similar Send to phone Share

Show Overview





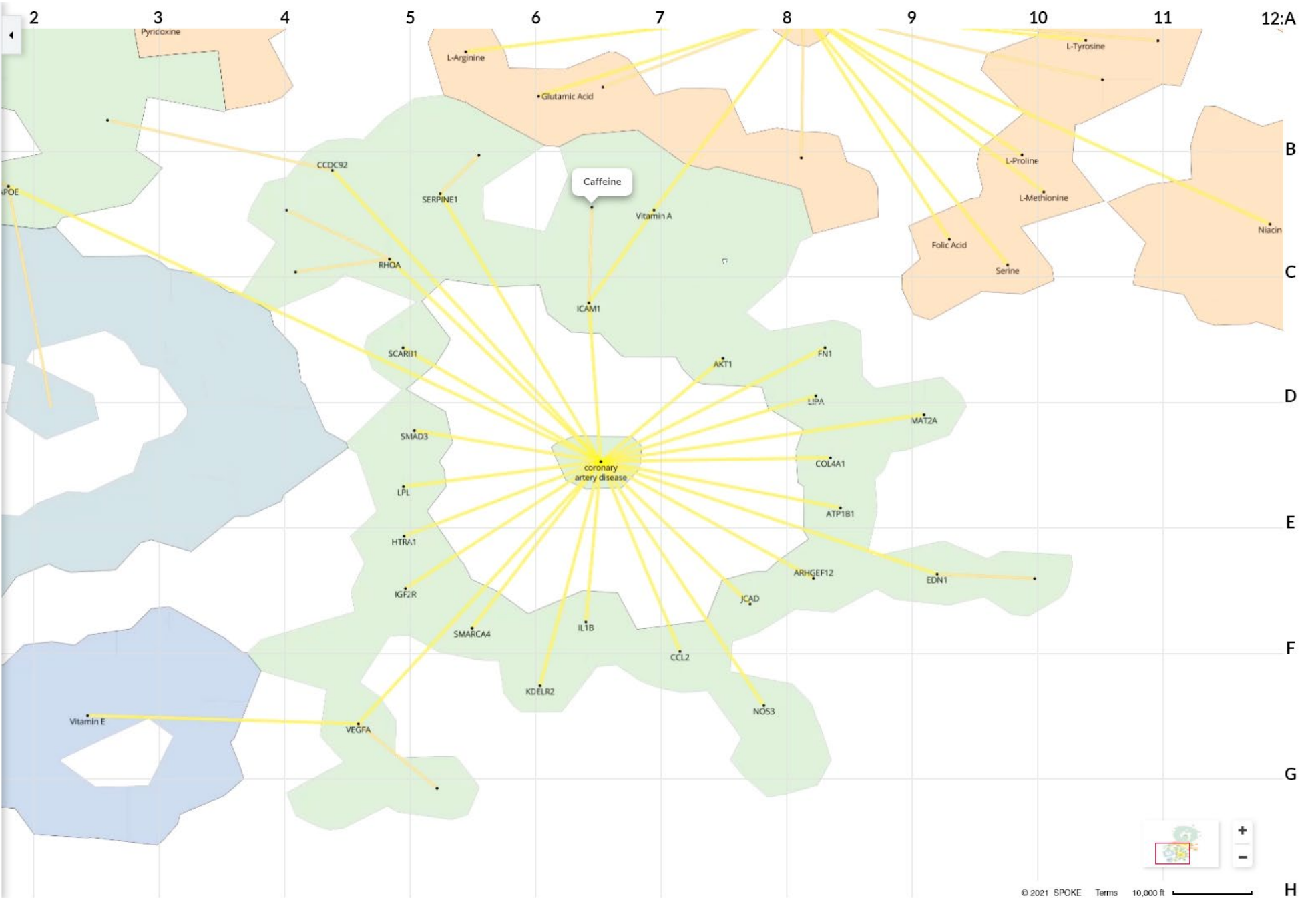
potato ×

coronary artery disease ×

SEARCH

Favorite Similar Send to phone Share

Show Overview



Send feedback

# Call for Papers: Special Issue on Multi-level Graph Representations for Big Data in Science

CG&A seeks submissions for this upcoming special issue.

The July/Aug 2022 special issue in *IEEE Computer Graphics and Applications* on “Multi-Level Graph Representations for Big Data in Science”

Articles due for review:  
December 29, 2021

## Guest editors:

- Katy Börner, Indiana University, Bloomington, US
- Stephen G. Kobourov, University of Arizona, Tucson, US

<https://www.computer.org/digital-library/magazines/cg/call-for-papers-special-issue-on-multi-level-graph-representations-for-big-data-in-science>

For centuries, cartographic maps have guided human exploration. While being rather imperfect initially, they helped explorers find promised lands and return home safely. Recent advances in data, algorithms, and computing infrastructures make it possible to map humankind’s collective scholarly knowledge and technology expertise by using topic maps on which “continents” represent major areas of science (e.g., mathematics, physics, or medicine) and zooming reveals successively more detailed subareas. Basemaps of science and technology are generated by analyzing citations links between millions of publications and/or patents. “Data overlays” (e.g., showing all publications by one scholar, institution, or country or the career trajectory of a scholar as a pathway) are generated by science-locating relevant publication records based on topical similarity. Despite the demonstrated utility of such maps, current approaches do not scale to the hundreds of millions of data records now available. The main challenge is designing efficient and effective methods to visualize and interact with more than 100 million scholarly publications at multiple levels of resolution.

This special issue invites researchers in cartography, data visualization, science of science, graph drawing, and other domains to submit novel and promising new research on graph mining and layout algorithms and their application to the development of science mapping standards and services. Topics of interest include:

- Science of science user needs and applications
- Efficient multi-level graph algorithms
- Network visualizations
- Effective user interfaces to large-scale data visualizations

## Deadlines

**Submissions due:** 29 December 2021

Preliminary notification: 2 March 2022

Revisions due: 6 April 2022

Final notification: 11 May 2022

Final version due: 25 May 2022

Publication: July/August 2022



# HuBMAP: Mapping 30+ Trillion Cells

Michael P. Snyder, et al. 2019. The human body at cellular resolution: The NIH Human Biomolecular Atlas Program. *Nature*. 574, p. 187-192.

<https://www.nature.com/articles/s41586-019-1629-x.pdf>



# HuBMAP

## Vision

Catalyze the development of an open, global framework for comprehensively mapping the human body at cellular resolution.

## Goals

1. Accelerate the development of the next generation of tools and techniques for constructing high resolution spatial tissue maps
2. Generate foundational 3D tissue maps
3. Establish an open data platform
4. Coordinate and collaborate with other funding agencies, programs, and the biomedical research community
5. Support projects that demonstrate the value of the resources developed by the program

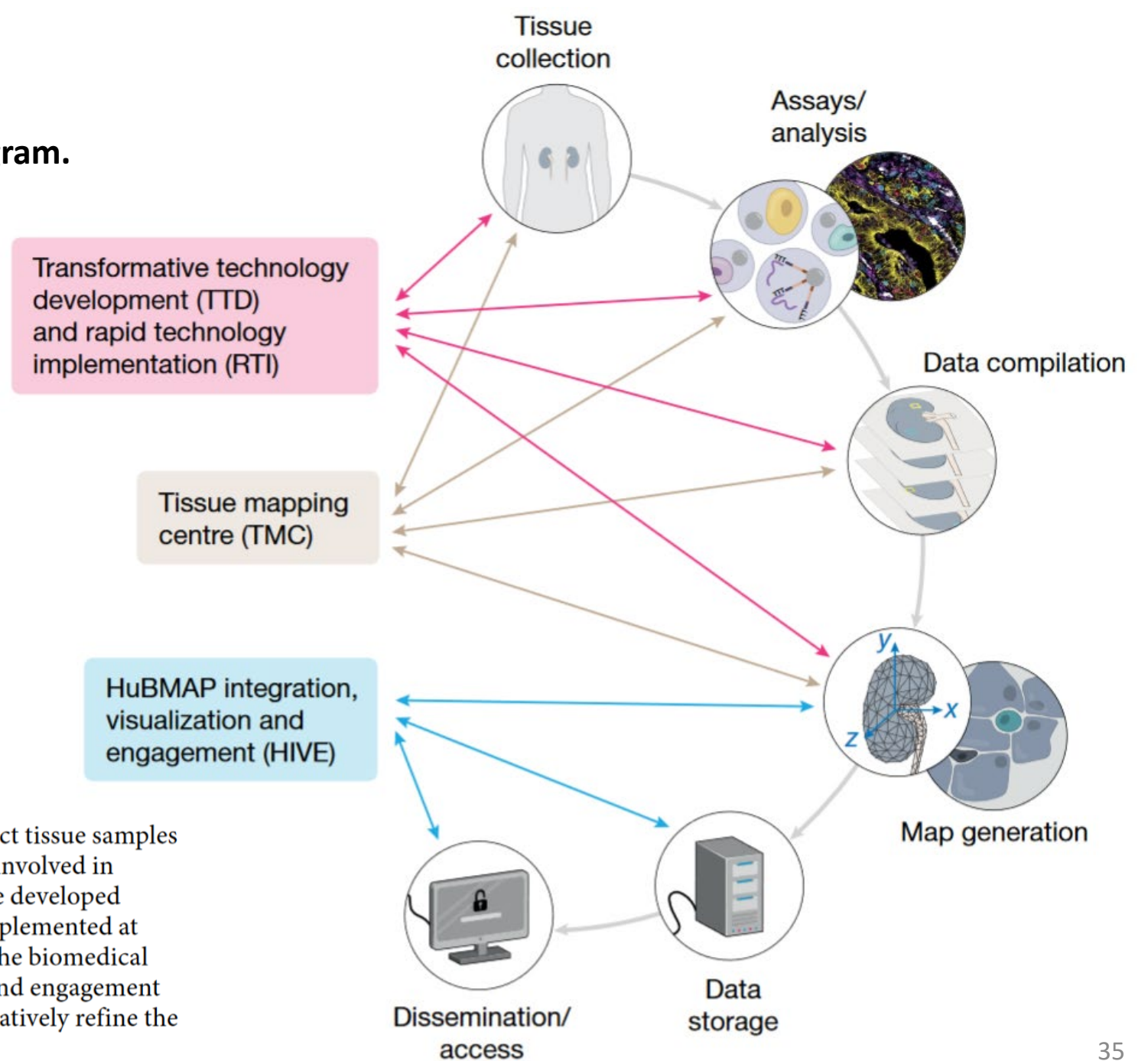


<https://commonfund.nih.gov/HuBMAP>



# The Human Body at Cellular Resolution: The NIH Human Biomolecular Atlas Program.

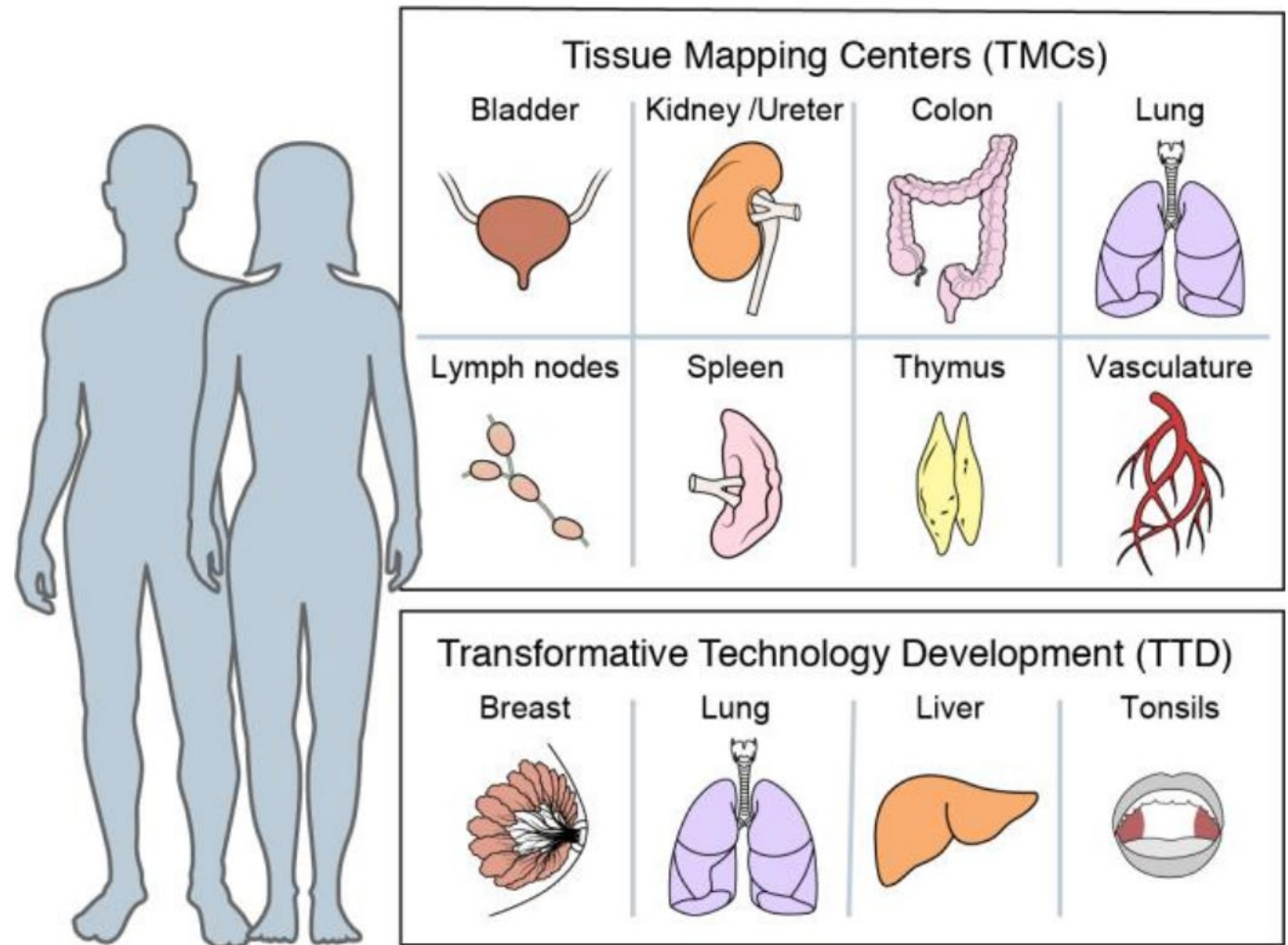
Snyder et al. *Nature*. 574, p. 187-192.



**Fig. 1 | The HubMAP consortium.** The TMCs will collect tissue samples and generate spatially resolved, single-cell data. Groups involved in TTD and RTI initiatives will develop emerging and more developed technologies, respectively; in later years, these will be implemented at scale. Data from all groups will be rendered useable for the biomedical community by the HuBMAP integration, visualization and engagement (HIVE) teams. The groups will collaborate closely to iteratively refine the atlas as it is gradually realized.

## The Human Body at Cellular Resolution: The NIH Human Biomolecular Atlas Program.

Snyder et al. *Nature*. 574, p. 187-192.

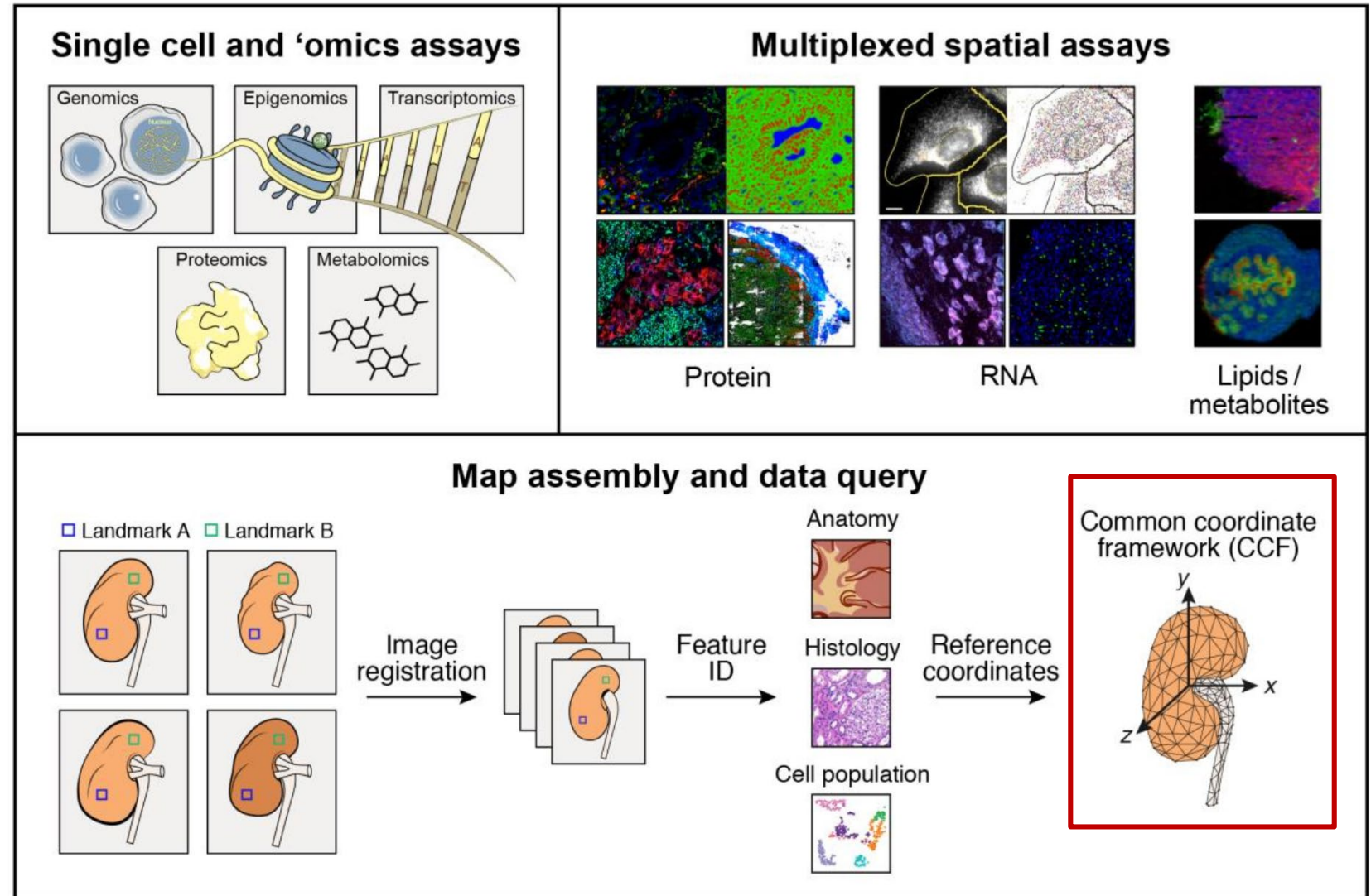


**Fig. 2 | Key tissues and organs initially analysed by the consortium.**

Using innovative, production-grade ('shovel ready') technologies, HuBMAP TMCs will generate data for single-cell, three-dimensional maps of various human tissues. In parallel, TTD projects (and later RTI projects) will refine assays and analysis tools on a largely distinct set of human tissues. Samples from individuals of both sexes and different ages will be studied. The range of tissues will be expanded throughout the program.



**The Human Body at Cellular Resolution: The NIH Human Biomolecular Atlas Program.**  
 Snyder et al. *Nature*. 574, p. 187-192.



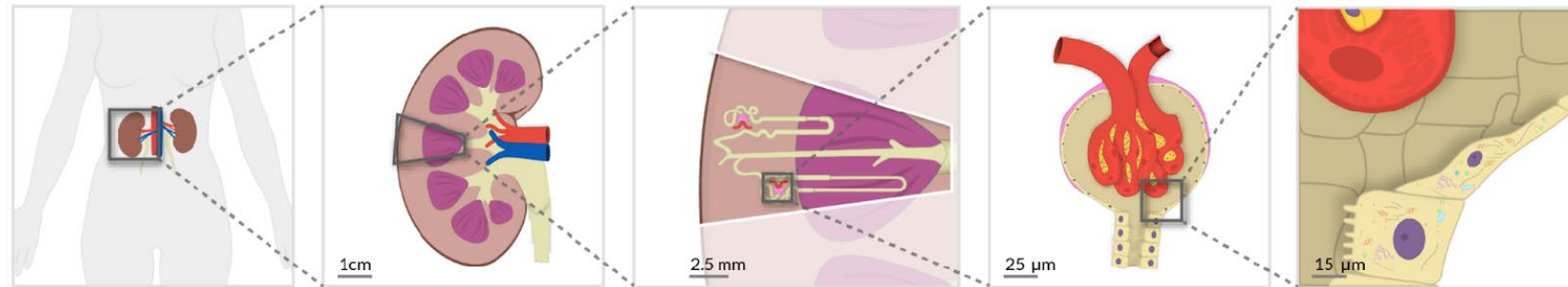
**Fig. 3 | Map generation and assembly across cellular and spatial scales.** HuBMAP aims to produce an atlas in which users can refer to a histological slide from a specific part of an organ and, in any given cell, understand its contents on multiple 'omic levels—genomic, epigenomic, transcriptomic, proteomic, and/or metabolomic. To achieve these ends, centres will apply a combination of imaging, 'omics and mass spectrometry

techniques to specimens collected in a reproducible manner from specific sites in the body. These data will be then be integrated to arrive at a high-resolution, high-content three-dimensional map for any given tissue. To ensure inter-individual differences will not be confounded with collection heterogeneity, a robust CCF will be developed.

# CCF Requirements

The CCF must capture major **anatomical structures, cell types, and biomarkers** and their interrelations across **multiple levels of resolution**.

It should be **semantically explicit** (using existing ontologies, e.g., Uberon, CL) and **spatially explicit** (e.g., using 3D reference organs for registration and exploration).



## Body

- Body
- Kidney (Left, Right)
- Aorta
- Renal artery
- Renal vein
- Ureter

## Organ

- Renal capsule
- Renal pyramid
- Renal cortex
- Renal medulla
- Renal calyx
- Renal pelvis

## Functional Tissue Unit

- Nephron
- Renal corpuscle
- Proximal convoluted tubule
- Loop of Henle
- Distal convoluted tubule
- Connecting tubule
- Collecting duct

## FTU Sub-structure(s)

- Bowman's capsule
- Glomerulus
- Efferent arteriole
- Afferent arteriole

## Cellular

- Parietal epithelial cell
- Capillary endothelial cell
- Mesangial cell
- Podocyte



# ASCT+B Tables

Anatomical Structures (AS), Cell Types (CT), and Biomarkers (B) or ASCT+B tables aim to capture the partonomy of anatomical structures, cell types, and major biomarkers (e.g., gene, protein, lipid or metabolic markers).

ASCT Table

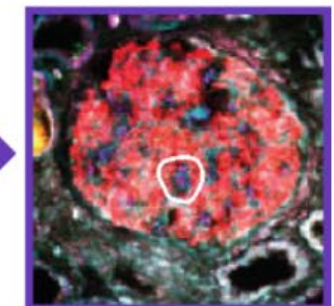
Structure/Region	Sub structure/Sub region	Cell Type
Renal Corpuscle	Bowman's (glomerular) Capsule/parietal layer	Parietal epithelial Cell
	Bowman's (glomerular) Capsule/visceral layer	Podocyte
	Glomerular Tuft	Capillary Endothelial Cell Mesangial Cell
Tubules	Proximal Tubule	Proximal Tubule Epithelial Cell (general)
		Proximal Convoluted Tubule Epithelial Cell Segment 1
		Proximal Tubule Epithelial Cell Segment 2
		Proximal Tubule Epithelial Cell Segment 2
		Proximal Tubule Epithelial Cell Segment 2
	Loop of Henle, Thin Limb	Descending Thin Limb Cell (general)
		Ascending Thin Limb Cell (general)
	Loop of Henle, Thick Limb	Thick Ascending Limb Cell (general)
	Distal Convolution	Cortex-TAL Cell
		Medulla-TAL Cell
		TAL-Macula Densa Cell
Distal Convoluted Tubule Cell (general)		
Connecting Tubule	DCT Type 1 Cell	
	DCT Type 2 Cell	
	Connecting Tubule Cell (general)	
	CNT-Principal Cell	

Ontology

**Anatomical Structures Partonomy**  
 kidney  
 kidney capsule  
 cortex of kidney  
 outer cortex of kidney  
 renal medulla

**Cell Types Ontology**  
 connective tissue cell  
 pericyte cell  
 mesangial cell  
 extraglomerular mesangial cell  
 glomerular mesangial cell

3D Reference Object Library

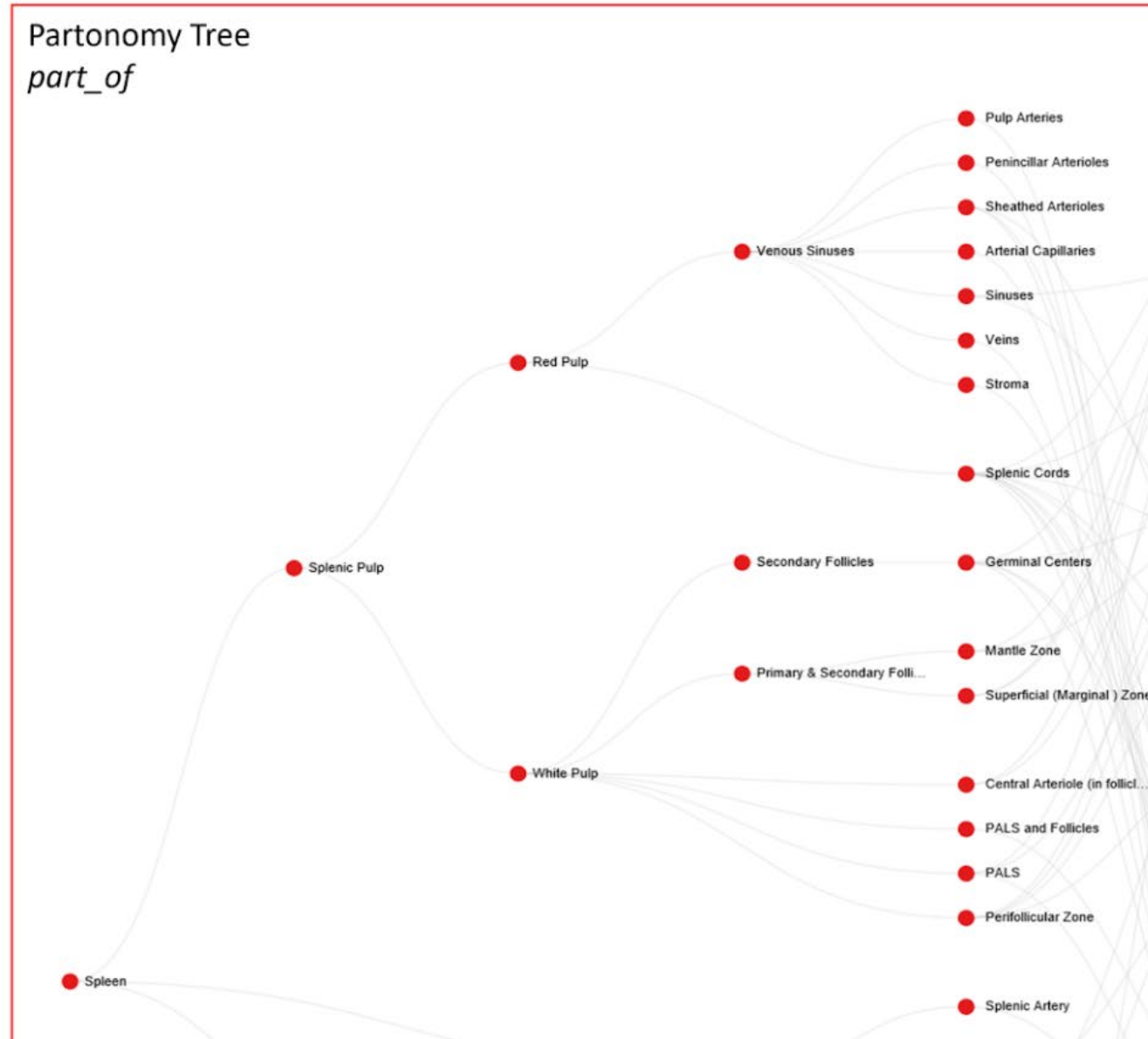


## Anatomical Structures (AS)

## Cell Types (CT)

## Biomarkers (B)

Partonomy Tree  
*part\_of*



Bimodal network describing which CT are located\_in what AS

Typology Tree  
*is\_a*

- adventitial stromal cell
- B cell
- Dendritic cell
- Endothelial
- Endothelial cell
- Erythrocytes
- fibroblast
- Fibroblastic reticular cell
- Follicular Dendritic cell
- Granulocytes
- Littoral cell
- Lymphatic endothelium
- macrophage
- Monocytes
- Myofibroblast
- neurons
- NK cell
- Plasma cell
- Plasmablasts
- Platelets

Bimodal network describing which B characterize what CT

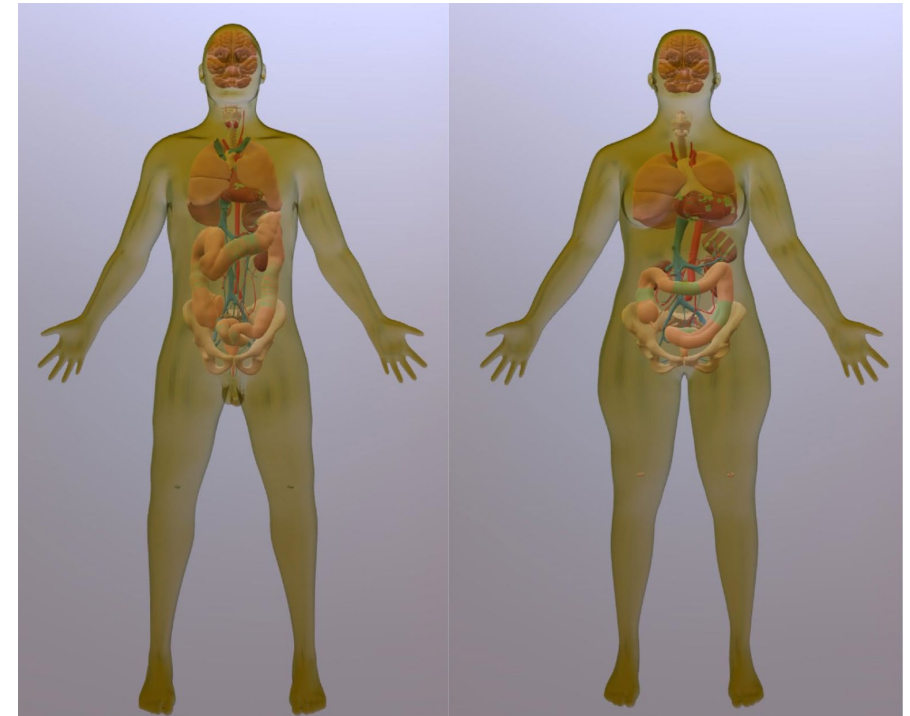
BG - Genes  
BP - Proteins

- CD10
- CD11b
- CD11c
- CD138
- CD14
- CD141
- CD15
- CD163
- CD19
- CD20
- CD21
- CD22
- CD23+
- CD235a
- CD27
- CD27-
- CD271
- CD271-
- CD3
- CD3-
- CD31
- CD34
- CD4
- CD4 (helper)
- CD41



Male

Female



Organ	3D Ref. Organ				ASCT+B Table			#AS-AS (part_of)	#CT-AS (located_in)	#B-CT (characterizes)
	#AS M-L	#AS F-L	#AS M-R	#AS F-R	#AS	#CT	#B			
BM & Blood / Pelvis	23	23			14	46	202	24	97	296
Brain	141	141			187	127	29	187	127	36
Heart	39	46			50	25	48	57	164	78
Intestine, Large	10	10			66	69	89	409	1410	192
Kidney	38	44	39	41	64	64	129	63	58	215
Lung	74	74			91	85	174	108	123	296
Lymph Nodes	7	7	7	7	40	49	161	60	117	342
Skin	1	1			16	42	70	17	19	105
Spleen	8	8			46	66	0	68	172	0
Thymus	2	2			18	46	55	20	103	64
Vasculature	84	85			869	2	1	868	606	2
<b>Totals</b>	<b>427</b>	<b>441</b>	<b>46</b>	<b>48</b>	<b>1461</b>	<b>621</b>	<b>958</b>	<b>1881</b>	<b>2996</b>	<b>1626</b>



<https://hubmapconsortium.github.io/ccf/pages/ccf-anatomical-structures.html>

<https://hubmapconsortium.github.io/ccf/pages/ccf-3d-reference-library.html> (NLM VH organs)

<https://community.brain-map.org/t/allen-human-reference-atlas-3d-2020-new/> (brain)

<https://www3.cs.stonybrook.edu/~ari/> (male colon)

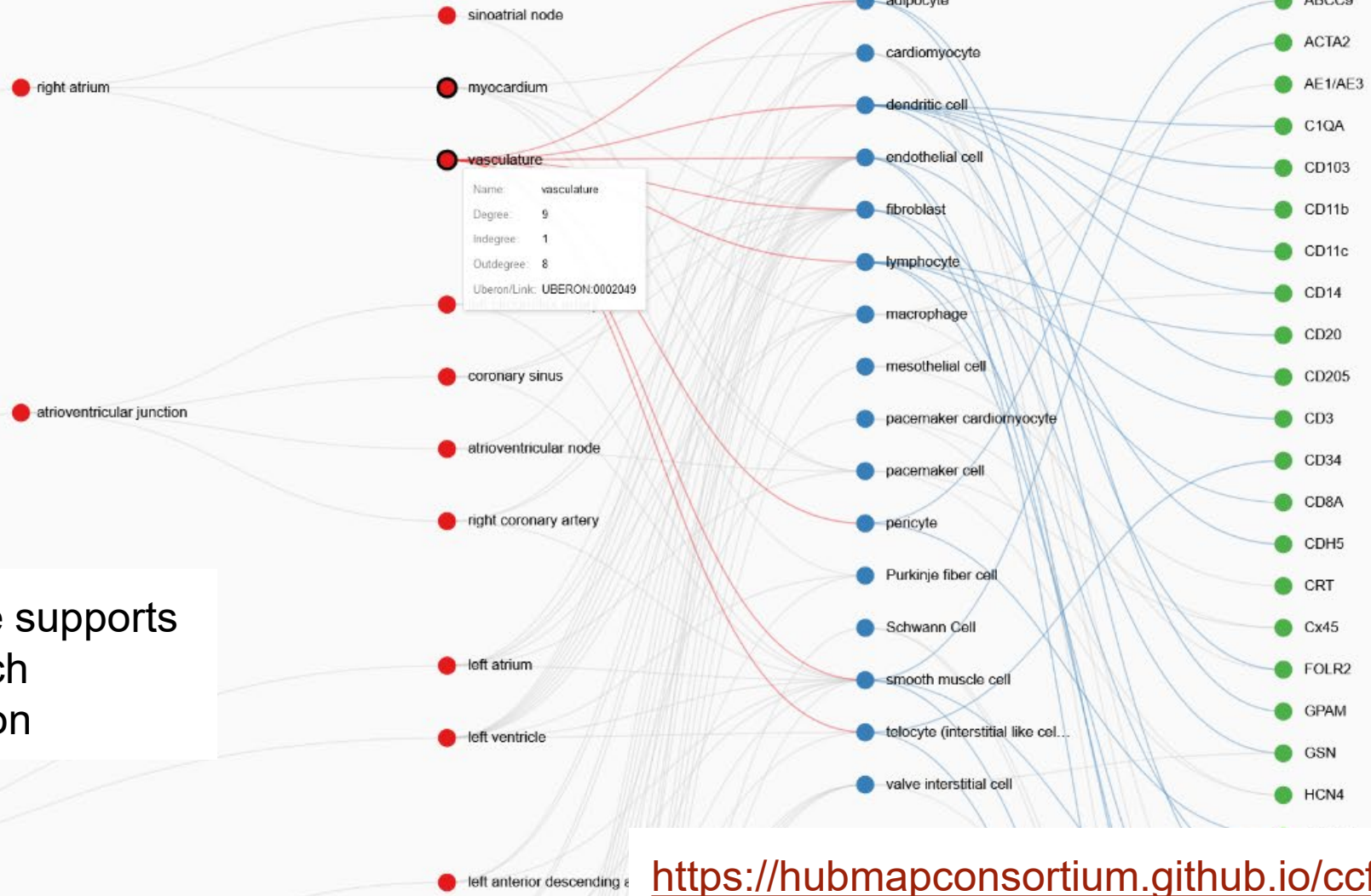
Anatomical Structures

Cell Types

Biomarkers

Legend

- Anatomical Structures
- Cell Types
- Biomarkers
- See Debug Log



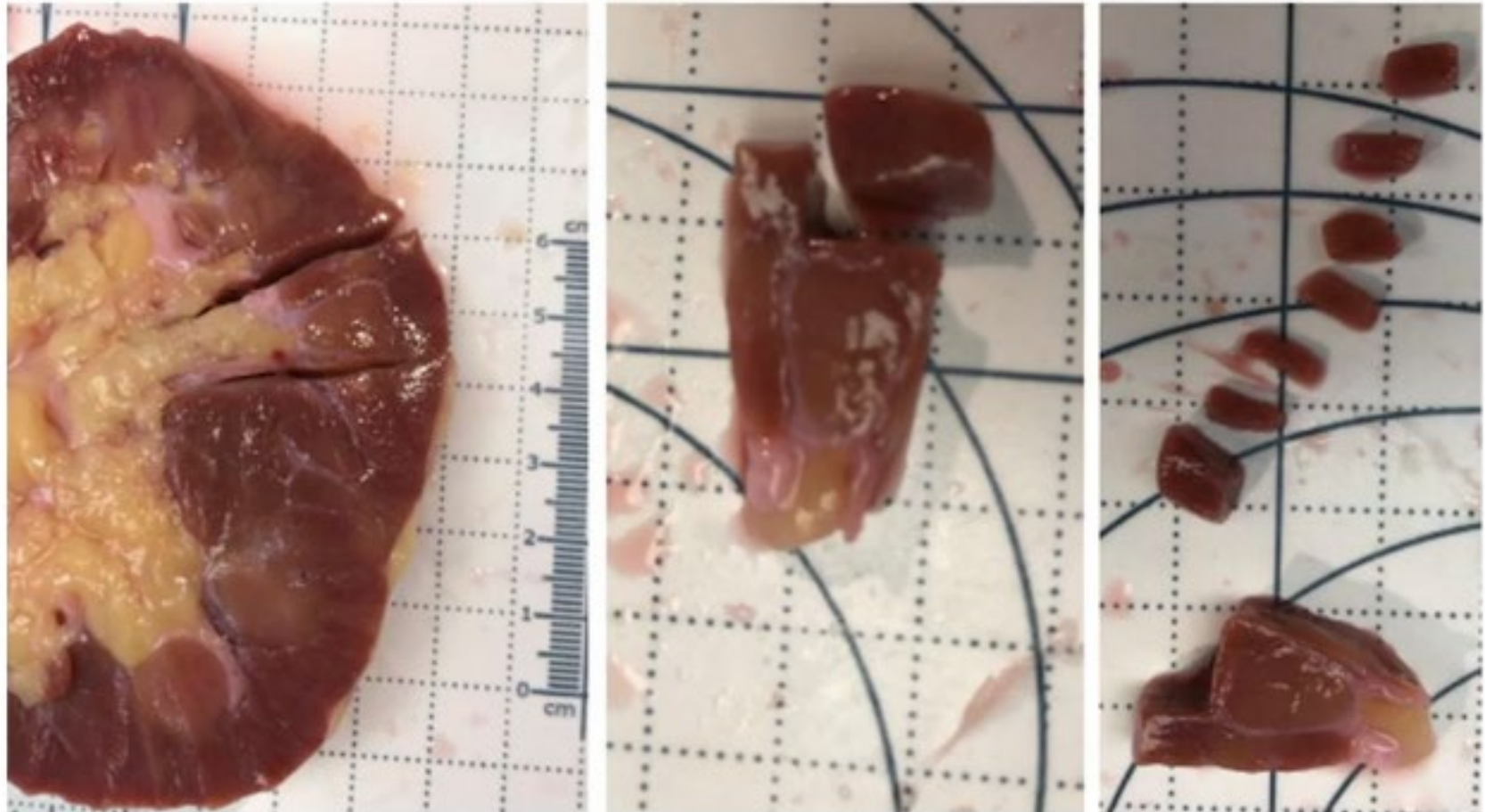
XMAS 2020 release supports

- AS, CT, B Search
- Table comparison

<https://hubmapconsortium.github.io/ccf-asct-reporter>



Document the tissue extraction site by registering tissue blocks within a 3D reference organ.

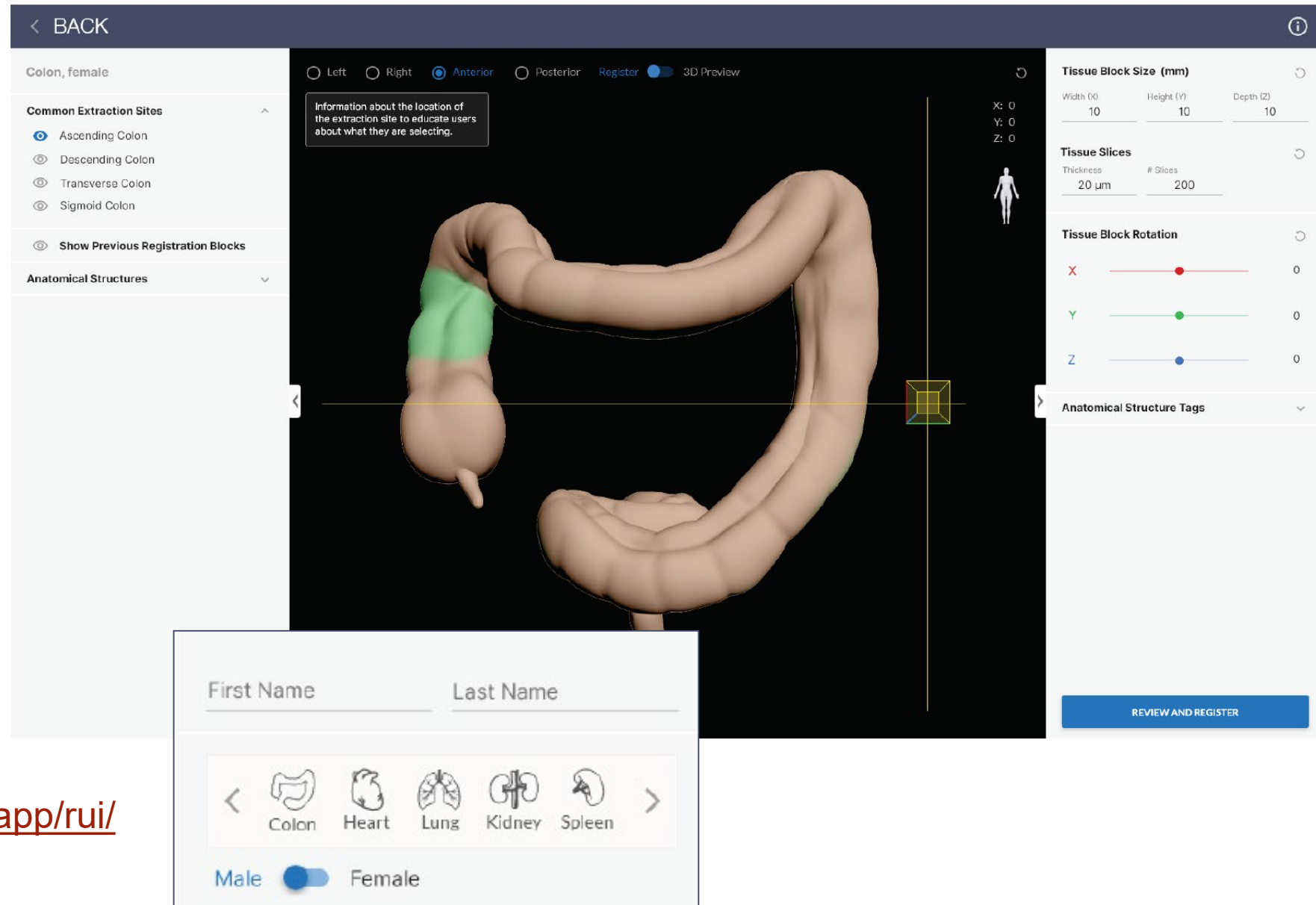


*Image provided by Sanjay Jain, TMC-UCSD*

# CCF Registration User Interface (RUI) v1.0.0

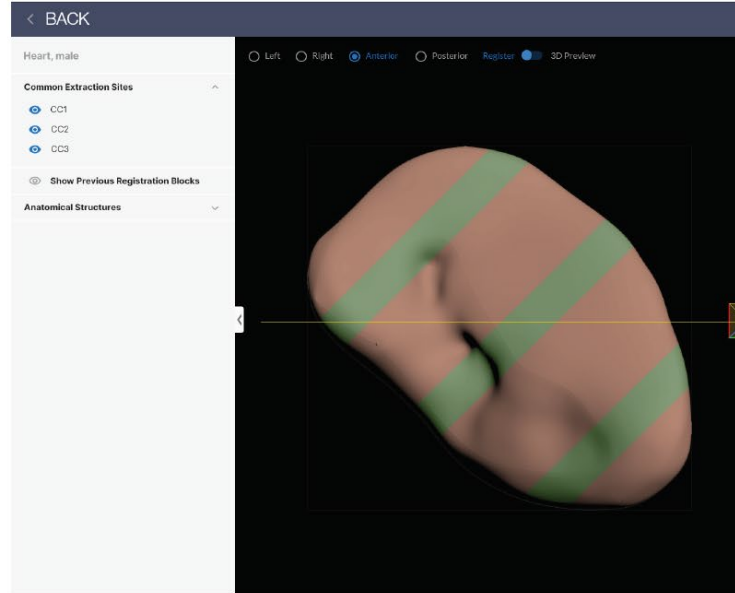
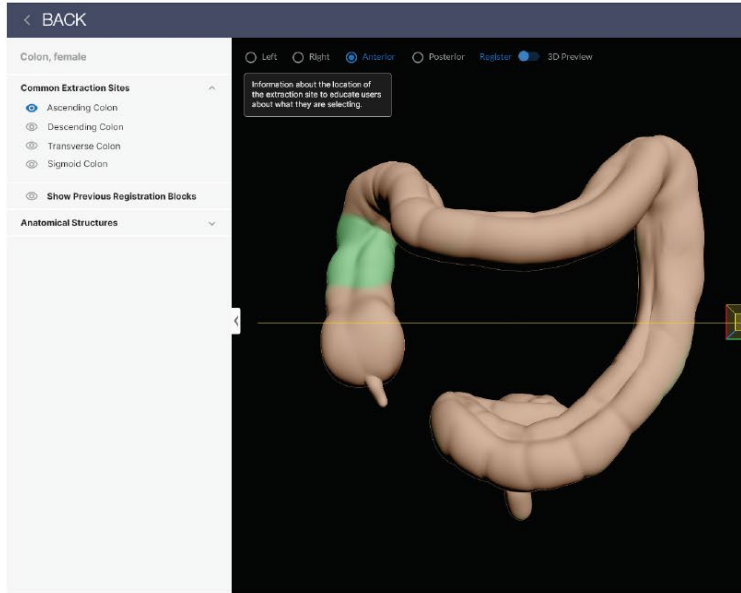
## New Features:

- Organ carousel with 4 reference organs
- Support for tissue extraction sites
- Expanded ontology
- Semantic annotation via collision detection & manual annotation
- Support for non-HuBMAP usage



<https://hubmap-ccf-ui.netlify.app/rui/>





### Kidney

- Bisection Line

### Spleen

- CC1
- CC2
- CC3

### Colon

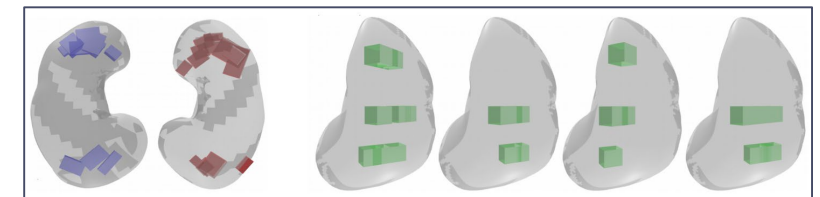
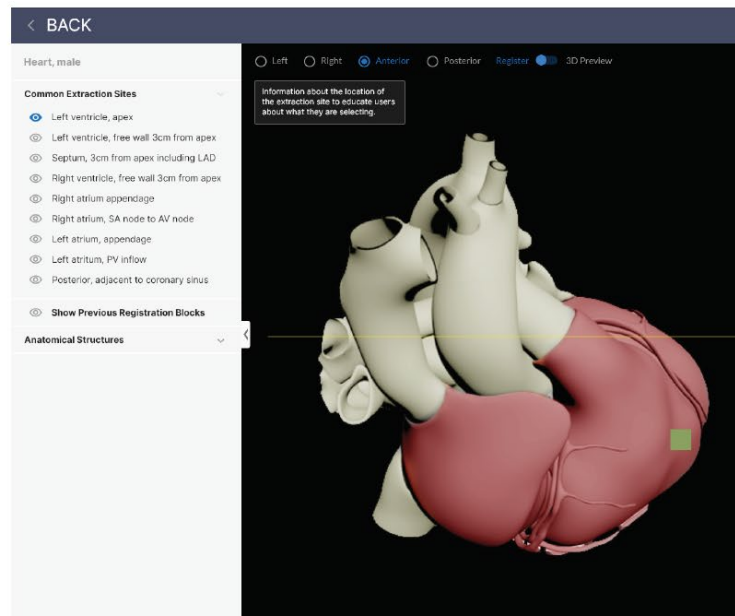
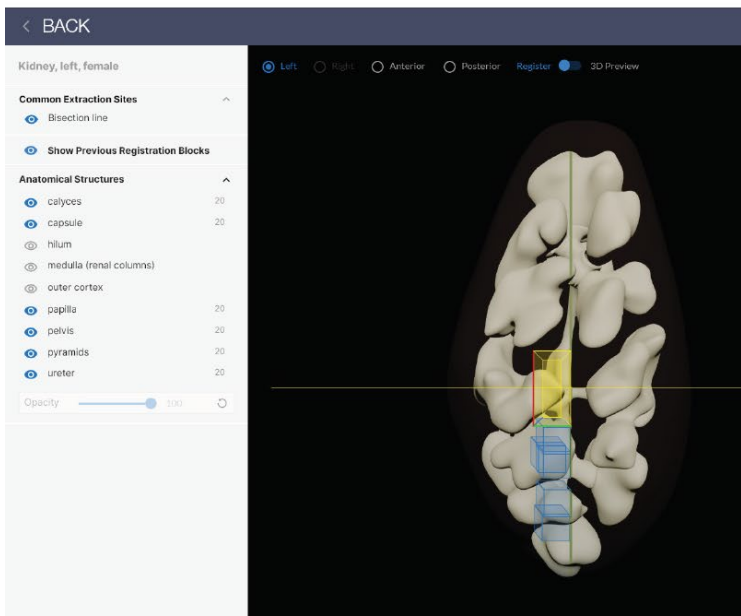
- Ascending Colon
- Descending Colon
- Transverse Colon
- Sigmoid Colon

### Heart

- Left atrium, appendage
- Left atrium, PV inflow
- Left ventricle, apex
- Left ventricle, free wall 3cm from apex
- Septum, 3cm from apex including LAD
- Posterior, adjacent to coronary sinus
- Right atrium appendage
- Right atrium, AV (atrioventricular) node
- Right atrium, SA (sinoatrial) node
- Right ventricle, free wall 3cm from apex

### Extraction Site Mapping

• Left atrium, appendage	7
• Left atrium, PV inflow	8
• Left ventricle, apex	1
• Left ventricle, free wall 3cm from apex	2
• Septum, 3cm from apex including LAD	3
• Posterior, adjacent to coronary sinus	9
• Right atrium appendage	5
• Right atrium, AV (atrioventricular) node	6a
• Right atrium, SA (sinoatrial) node	6b
• Right ventricle, free wall 3cm from apex	4



For the first HuBMAP portal release, 48 tissue blocks were registered.

# CCF Exploration User Interface (EUI)

**HuBMAP** Sex: Both Age: 1-110 BMI: 13-83 [Login](#)

Search ontology terms ...

- body
  - heart
  - lung
  - kidney
    - right kidney
    - left kidney
    - kidney capsule
    - cortex of kidney
    - renal medulla
    - renal column
    - renal pyramid
    - hilum of kidney
    - kidney interstitium
    - kidney calyx
    - renal pelvis
    - ureter
    - renal papilla
    - renal fat pad
    - nephron

**body**

- 2 Centers
- 27 Donors
- 41 Samples

	<b>Female, Age 14, BMI 14.7</b> HBM894.MPVN.828 TMC-Florida First case collected. Incomplete d...	
	<b>Male, Age 18, BMI 27.1</b> HBM436.GHWX.449 TMC-Florida section is 190um from block surface	
	<b>Male, Age 56, BMI 32.5</b> HBM696.XTVL.498 TMC-Vanderbilt Age 56, White Male	
	<b>Male, Age 53, BMI 26.5</b> HBM652.VRLD.292 TMC-Vanderbilt Age 53, Black Male	
	<b>Male, Age 58, BMI 22.0</b> HBM477.CJKM.888 TMC-Vanderbilt 107-111	
	<b>Male, Age 18, BMI 25.5</b> HBM473.VKCM.878 TMC-Florida section is 255um from block surface	
	<b>Male, Age 55, BMI 25.4</b> HBM824.BLXF.883 TMC-Vanderbilt 13-16	





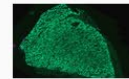





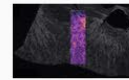

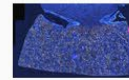

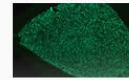



<https://portal.hubmapconsortium.org/ccf-eui>






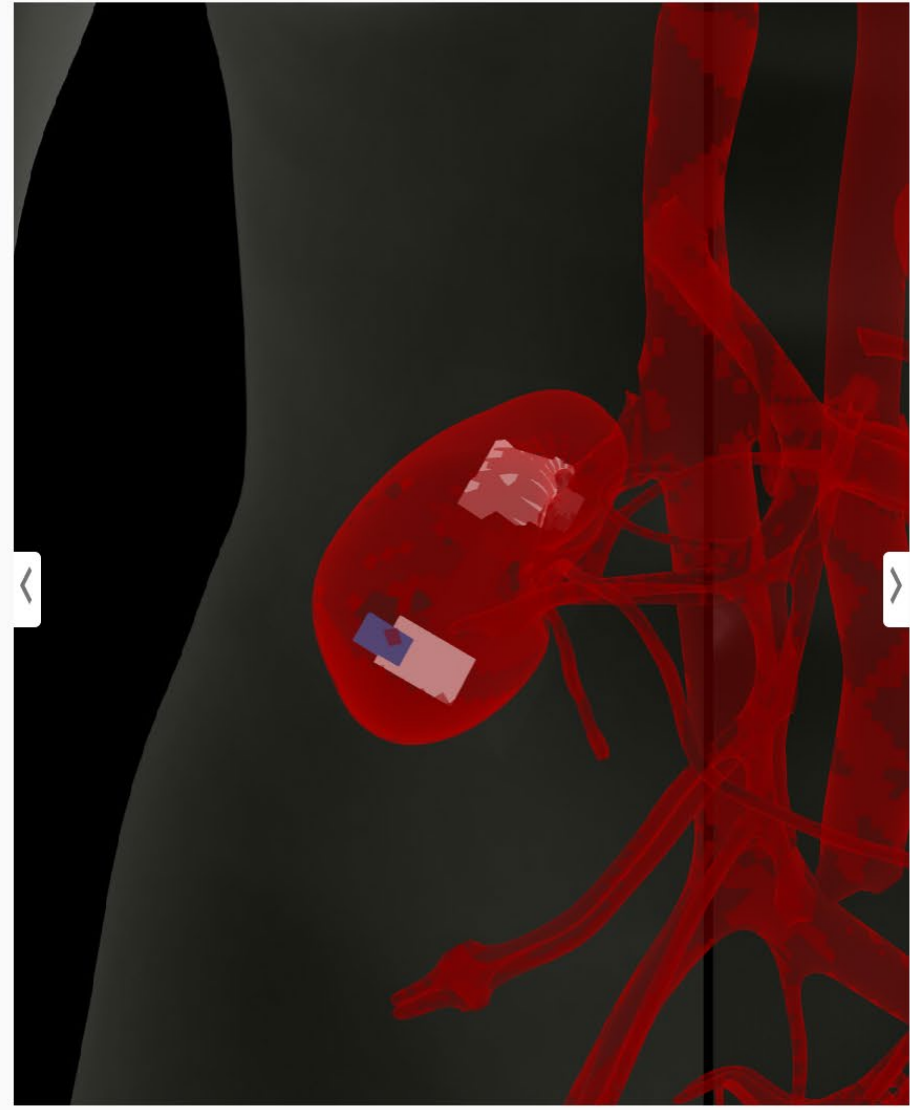
body

- 1 Centers
- 9 Donors
- 40 Samples

	<b>Male, Age 55, BMI 25.4</b> HBM695.RTLJ.484 TMC-Vanderbilt 13-16	
	<b>Male, Age 21, BMI 21.8</b> HBM634.MMGK.572 TMC-Vanderbilt Age 21 , White Male, Trauma Patient	
	<b>Female, Age 44, BMI 28.0</b> HBM457.NNQN.252 TMC-Vanderbilt Age 44, white female.	
	<b>Female, Age 44, BMI 28.0</b> HBM465.VKHL.532 TMC-Vanderbilt Age 44, white female.	
	<b>Male, Age 21, BMI 21.8</b> HBM693.HFFJ.752 TMC-Vanderbilt Age 21 , White Male, Trauma Patient	
	<b>Female, Age 58, BMI 23.0</b> HBM536.LDTZ.757 TMC-Vanderbilt Age 58, White Female	
	<b>Male, Age 48, BMI 35.3</b> HBM334.GCCX.874 TMC-Vanderbilt Age 48, White Male	
	<b>Male, Age 31, BMI 32.6</b> HBM776.PKJF.786 TMC-Vanderbilt Age 21, White Male	
	<b>Female, Age 66, BMI 31.3</b> HBM284.TRCV.726	

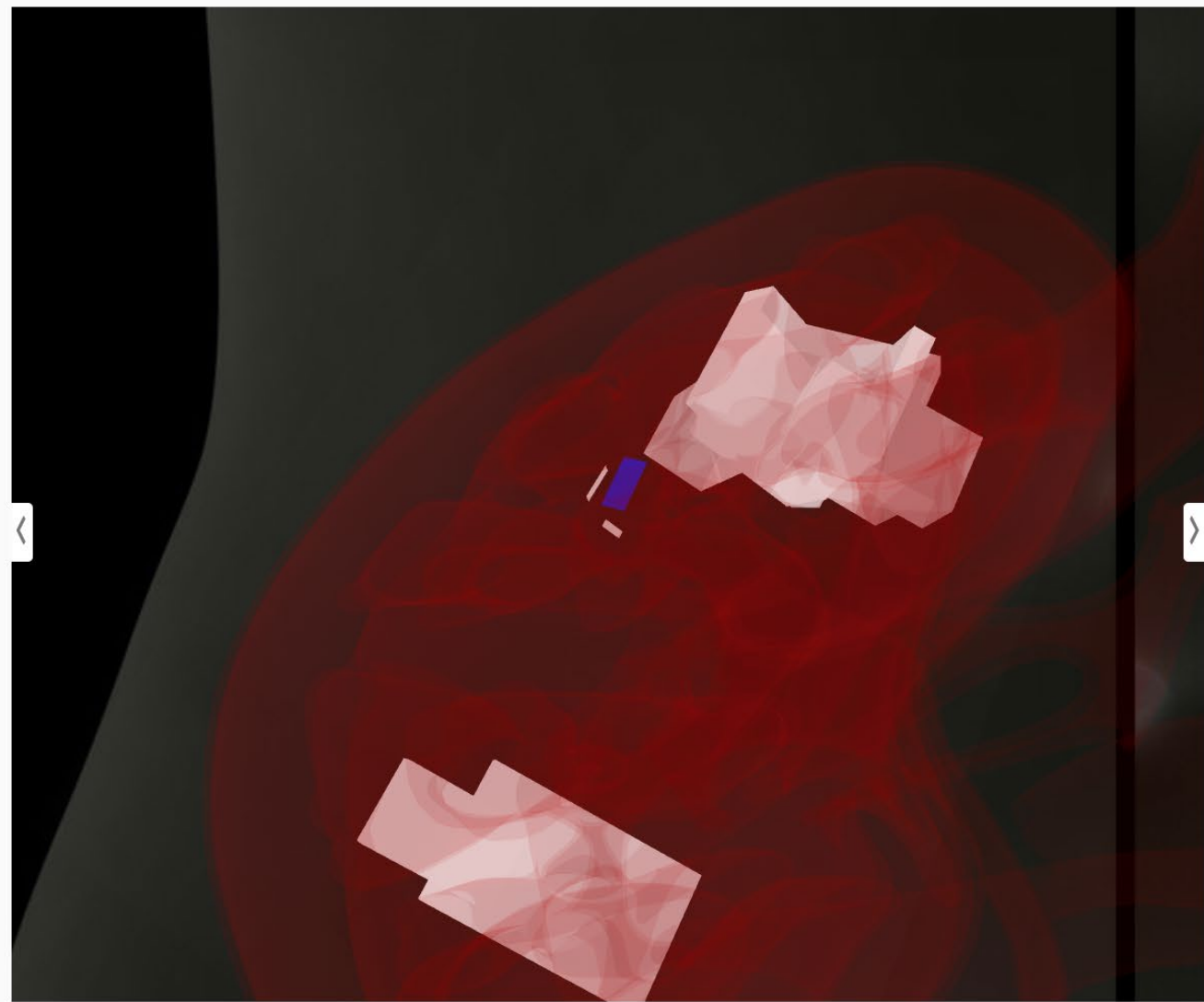
Search ontology terms ... 

- body
  - heart
  - lung
  - kidney
    - right kidney
    - left kidney
    - kidney capsule
    - cortex of kidney
    - renal medulla
    - renal column
    - renal pyramid
    - hilum of kidney
    - kidney interstitium
    - kidney calyx
      - major calyx
      - minor calyx
    - renal pelvis
    - ureter
    - renal papilla
    - renal fat pad
  - nephron
  - spleen
  - colon



Search ontology terms ... 🔍

- ▼ body
  - heart
  - lung
  - ▼ kidney
    - right kidney
    - left kidney
    - kidney capsule
    - ▼ cortex of kidney
      - outer cortex of kidney
    - ▼ renal medulla
      - outer medulla
      - inner medulla
    - renal column
    - renal pyramid
    - hilum of kidney
    - kidney interstitium
    - ▼ kidney calyx
      - major calyx
      - minor calyx
    - renal pelvis
    - ureter
    - renal papilla
    - renal fat pad
    - nephron
  - spleen
  - colon
  - small intestine



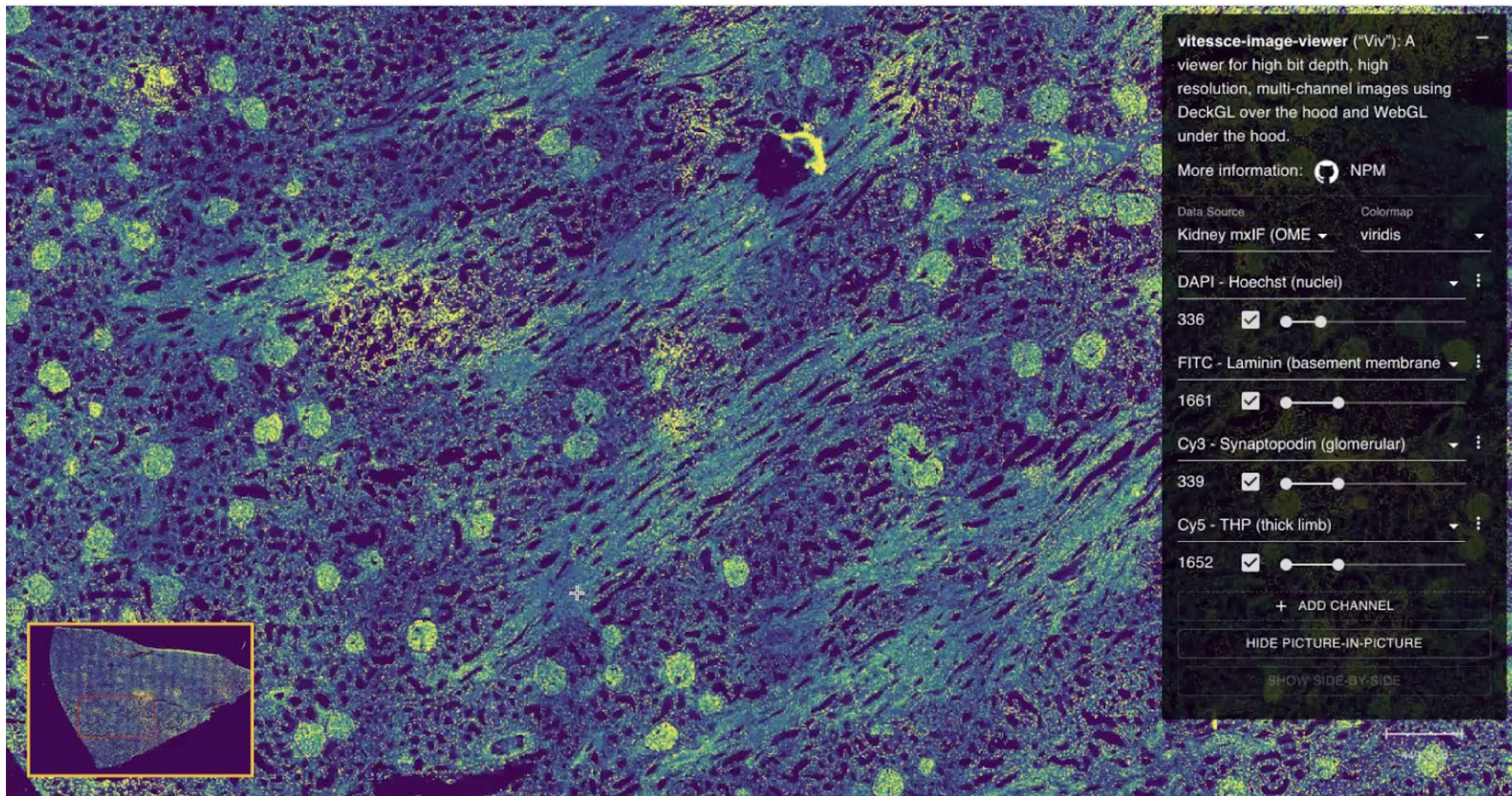
body

- 2 Centers
- 9 Donors
- 14 Samples

	<b>CoverNephrectomy</b> 10.1016/j.trsl.2017.07.006 KPMP-IU/OSU Isolated as a part of a kidney st...	☰
	<b>Patient B Cortical biopsy</b> 10.1681/ASN.2016091027 KPMP-IU/OSU Biopsy from Nephrology bioban...	☰
	<b>Patient A Cortical biopsy</b> 10.1681/ASN.2016091027 KPMP-IU/OSU Biopsy from Nephrology bioban...	☰
	<b>Male, Age 55, BMI 25.4</b> HBM824.BLXF.883 TMC-Vanderbilt 13-16	☰
	<b>Female, Age 66, BMI 31.3</b> HBM554.ZRCG.496 TMC-Vanderbilt 21-24	☰
	<b>Female, Age 58, BMI 23.0</b> HBM926.VBJV.597 TMC-Vanderbilt Age 58, White Female	☰
	<b>Male, Age 62, BMI 34.9</b> HBM947.VLDP.894 TMC-Vanderbilt Kidneys 153-156	☰
	<b>Female, Age 44, BMI 28.0</b> HBM457.NNQN.252 TMC-Vanderbilt Age 44, white female.	☰
	<b>Male, Age 21, BMI 21.8</b> HBM693.HFFJ.752 TMC-Vanderbilt Age 21, White Male, Trauma Pat...	☰
	<b>Female, Age 58, BMI 23.0</b> HBM536.LDTZ.757 TMC-Vanderbilt Age 58, White Female	☰
	<b>Male, Age 48, BMI 35.3</b>	☰

Register your data via <https://hubmap-ccf-ui.netlify.app/rui/> so it can be spatially/semantically explored in EUI.





<http://gehlenborglab.org/research/projects/vitessce/>



# VH Massive Open Online Course (VHMOOC)

## Goals

- Communicate tissue data acquisition and analysis,
- Demonstrate single-cell analysis and CCF mapping techniques, and
- Introduce major features of the HuBMAP portal.

## Learning modules come with

- Videos (incl. interviews, tool demos)
- Hands-on exercises
- Self-quizzes



**HuBMAP Visible Human MOOC (VHMOOC)**  
Started Aug 4, 2020  
[GO TO CANVAS COURSE](#)  
You are enrolled.



### Course Introduction

This 10h course introduces the HuBMAP project which aims to create an open, global reference atlas of the human body at the cellular level. Among others, the course describes the compilation and coverage of HuBMAP data, demonstrates new single-cell analysis and mapping techniques, and introduces major features of the HuBMAP portal.

Delivered entirely online, all coursework can be completed asynchronously to fit busy schedules. If you have questions or experience issues during registration, please email [cnsctr@indiana.edu](mailto:cnsctr@indiana.edu).

### Learning Outcomes

- Theoretical and practical understanding of different single-cell tissue analysis techniques.
- Expertise in single-cell data harmonization used to federate data from different individuals analyzed using different technologies in diverse labs.
- Hands-on skills in the design and usage of semantic ontologies that describe human anatomy, cell types, and biomarkers (e.g., marker genes or proteins).
- Knowledge on the design and usage of a semantically annotated three-dimensional reference system for the healthy human body.
- An understanding of how the HuBMAP reference atlas might be used to understand human health but also to diagnose and treat disease.

### Module Topics Include

- HuBMAP Overview: Project Goals, Setup, and Ambitions
- Tissue Data Acquisition and Analysis
- Biomolecular Data Harmonization
- Ontology, 3D Reference Objects, and User Interfaces
- HuBMAP Portal Design and Usage

### Meet the Instructors



**Katy Börner**, Victor H. Yingve Distinguished Professor of Engineering and Information Science. Founding Director of the [Cyberinfrastructure for Network Science Center](#) at Indiana University.



**Ellen M. Quardokus**, staff in the Chemistry Department and research scientist, [Cyberinfrastructure for Network Science Center](#), SICE with expertise in molecular biology, microscopy, anatomy, and interdisciplinary communication.

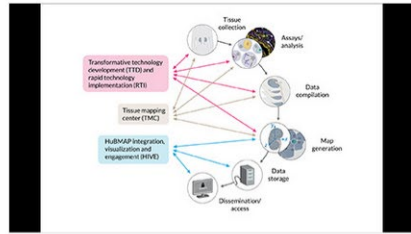


**Andreas Bueckle**, PhD Candidate in Information Science, performing research on information visualization, specifically virtual and augmented reality.

- Length:** 10 hours
- Department:** Cyberinfrastructure Network Science
- Credit:** None
- Audience:** Biomedical students and professionals interested in single-cell tissue analysis and visualization

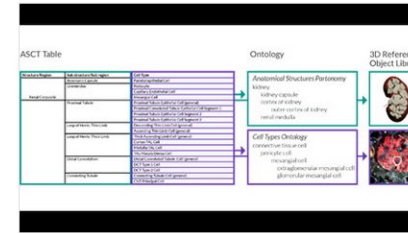
<https://expand.iu.edu/browse/sice/cns/courses/hubmap-visible-human-mooc>





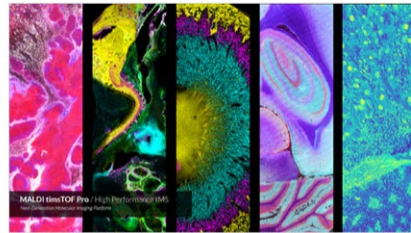
## HuBMAP Overview

- Project Goals, Setup, and Ambitions



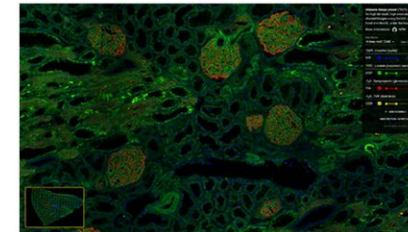
## CCF Ontology, 3D Reference Objects, and User Interfaces

- Creating an Atlas of the Human Body



## Tissue Data Acquisition and Analysis

- Behind the Scenes at Vanderbilt University



## Portal Design and Usage

- Datasets and Software in the 1st HuBMAP Portal Release



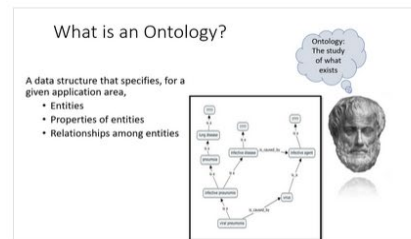
## Biomolecular Data Harmonization

- An Introduction to Seurat



## Open Consent Your Data

- In Support of Research



## Ontologies 101

- A gentle introduction on how to use ontologies the world.



## Anatomical Structures, Cell Types, and Biomarkers (ASCT+B) Tables

- What are ASCT+B tables and how they are used.

# Acknowledgements

HuBMAP Consortium (<https://hubmapconsortium.org>)



Thanks go to all the **patients** that agreed to volunteer healthy tissue and open use of their data.



## TMCs



**Jeffrey Spraggins**  
TMC-Vanderbilt  
Vanderbilt University



**Sanjay Jain**  
TMC-UCSD  
Washington University,  
St. Louis



**Clive Wasserfall**  
TMC-UFL  
University of Florida



**Marda Jorgensen**  
TMC-UFL  
University of Florida



**Kristen Browne**  
Medical Imaging and  
3D Modeling Specialist  
NIAID

## 3D Models

## MC-IU HIVE Team



**Katy Börner**  
MC-IU PI  
CNS Director



**Griffin Weber**  
Assoc. Professor of Medicine  
Harvard Medical School



**Lisel Record**  
MC-IU PM  
CNS Associate Director



**Bruce Herr II**  
Sr. Systems Architect/PM



**Ellen Quardokus**  
Sr. Research Analyst



**Yingnan Ju**  
PhD Candidate



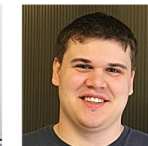
**Andreas Bueckle**  
PhD Candidate



**Leonard Cross**  
Sr. UX/UI Designer



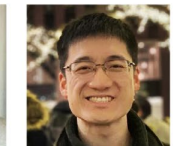
**Matthew Martindale**  
Center Assistant



**Daniel Bolin**  
Software Developer



**Adam Phillips**  
Software Developer



**Edward Lu**  
Software Developer



**Paul Hrishikesh**  
Research Assistant



**Leah Scherschel**  
Research Assistant



**Avinash Boppana**  
Research Consultant



# Data Visualization Literacy Framework

Börner, Katy, Andreas Bueckle, and Michael Ginda. 2019. Data visualization literacy: Definitions, conceptual frameworks, exercises, and assessments. *PNAS*, 116 (6) 1857-1864.



# Data Visualization Literacy (DVL)

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

- literacy (ability to read and write text in titles, axis labels, legends, etc.),
- visual literacy (ability to find, interpret, evaluate, use, and create images and visual media), and
- mathematical literacy (ability to formulate, employ, and interpret math in a variety of contexts).

Being able to “read and write” data visualizations is becoming as important as being able to read and write text. Understanding, measuring, and improving data and visualization literacy is important to strategically approach local and global issues.



# DVL Framework: Desirable Properties

- Most existing frameworks focus on **READING**. We believe that much expertise is gained from also **CONSTRUCTING** data visualizations.
- Reading and constructing data visualizations needs to take human perception and cognition into account.
- Frameworks should build on and consolidate prior work in cartography, psychology, cognitive science, statistics, scientific visualization, data visualization, learning sciences, etc. in support of a de facto standard.
- Theoretically grounded + practically useful + easy to learn/use.
- Highly modular and extendable.

# DVL Framework: Development Process

- The initial DVL-FW was developed via an extensive literature review.
- The resulting DVL-FW typology, process model, exercises, and assessments were then tested in the *Information Visualization* course taught for more than 17 years at Indiana University. More than 8,500 students enrolled in the IVMOOC version (<http://ivmooc.cns.iu.edu>) over the last six years.
- The FW was further refined using feedback gained from constructing and interpreting data visualizations for 100+ real-world client projects.
- Data on student engagement, performance, and feedback guided the continuous improvement of the DVL-FW typology, process model, and exercises for defining, teaching, and assessing DVL.
- The DVL-FW used in this course supports the systematic construction and interpretation of data visualizations.

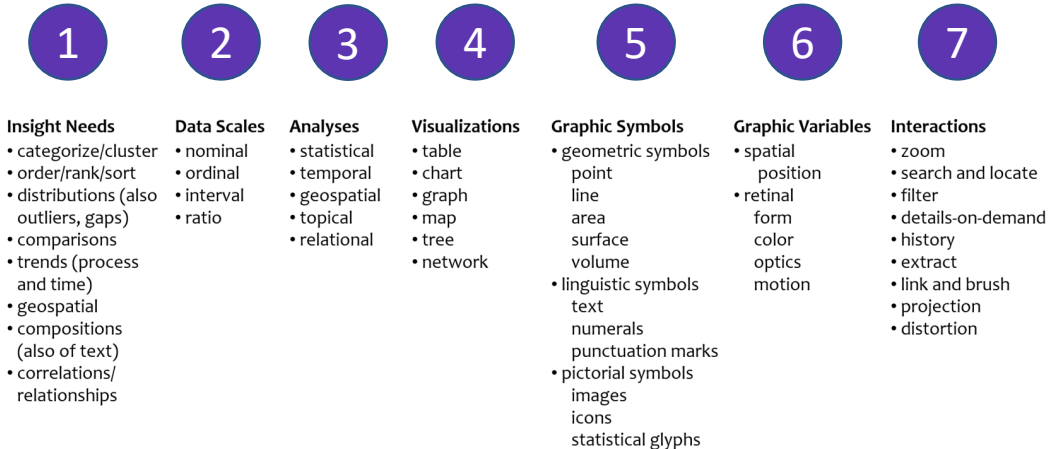


# Data Visualization Literacy Framework (DVL-FW)

Consists of two parts:

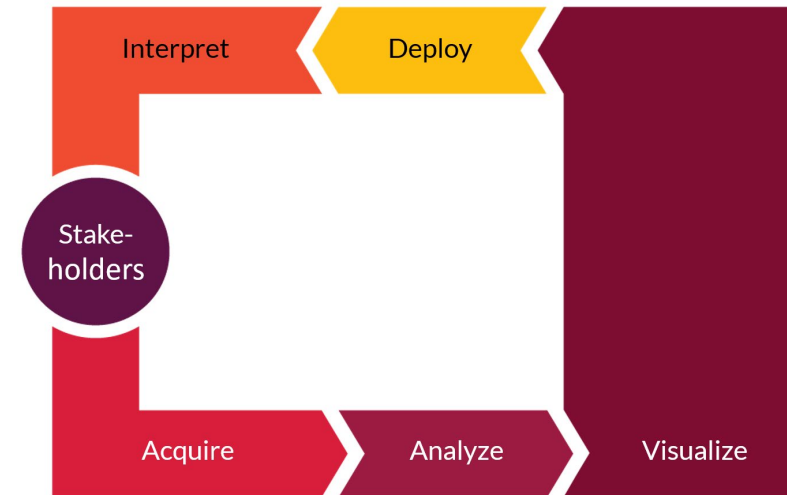
## DVL Typology

Defines 7 types with 4-17 members each.



## DVL Workflow Process

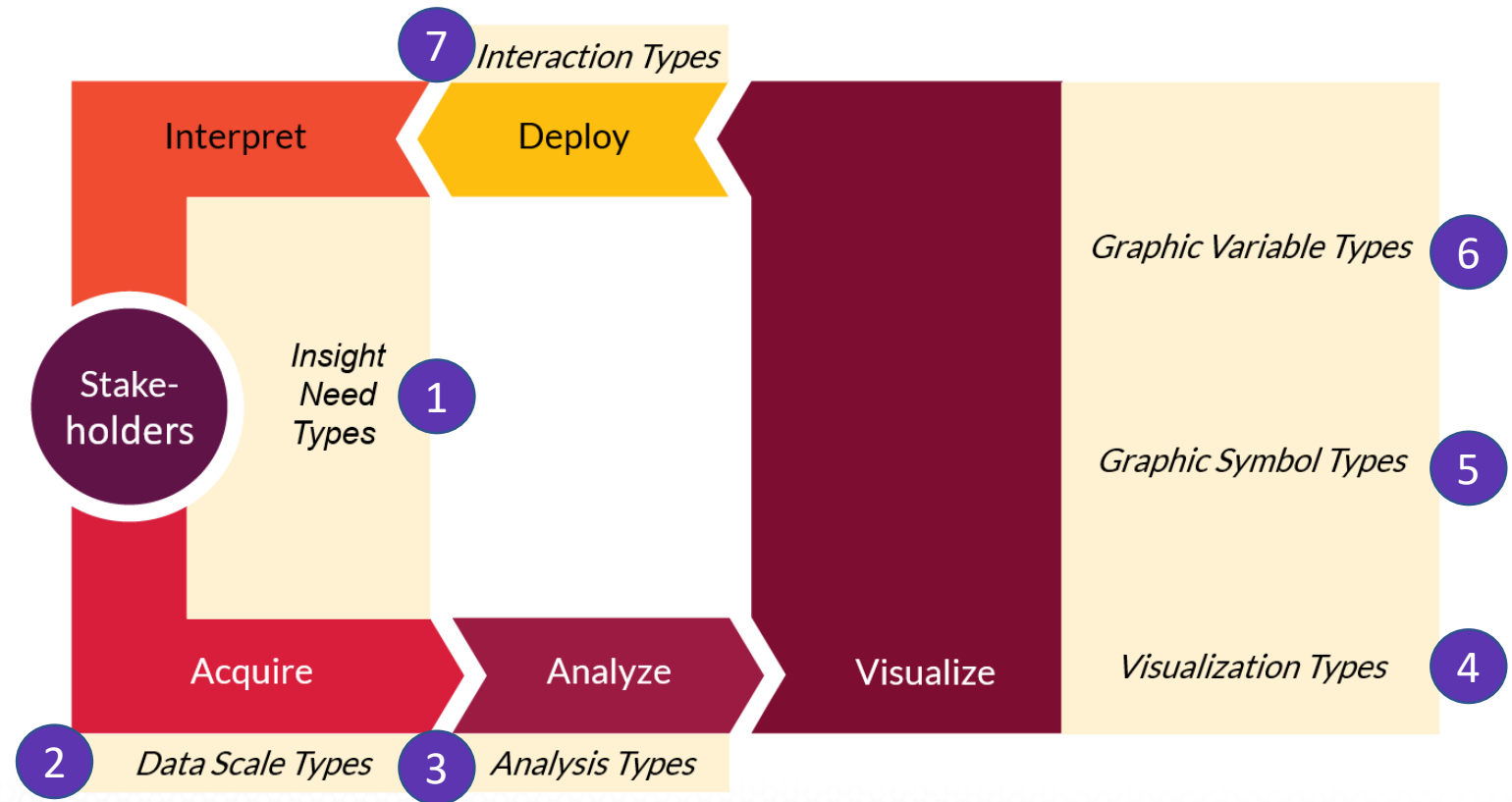
Defines 5 steps required to render data into insights.



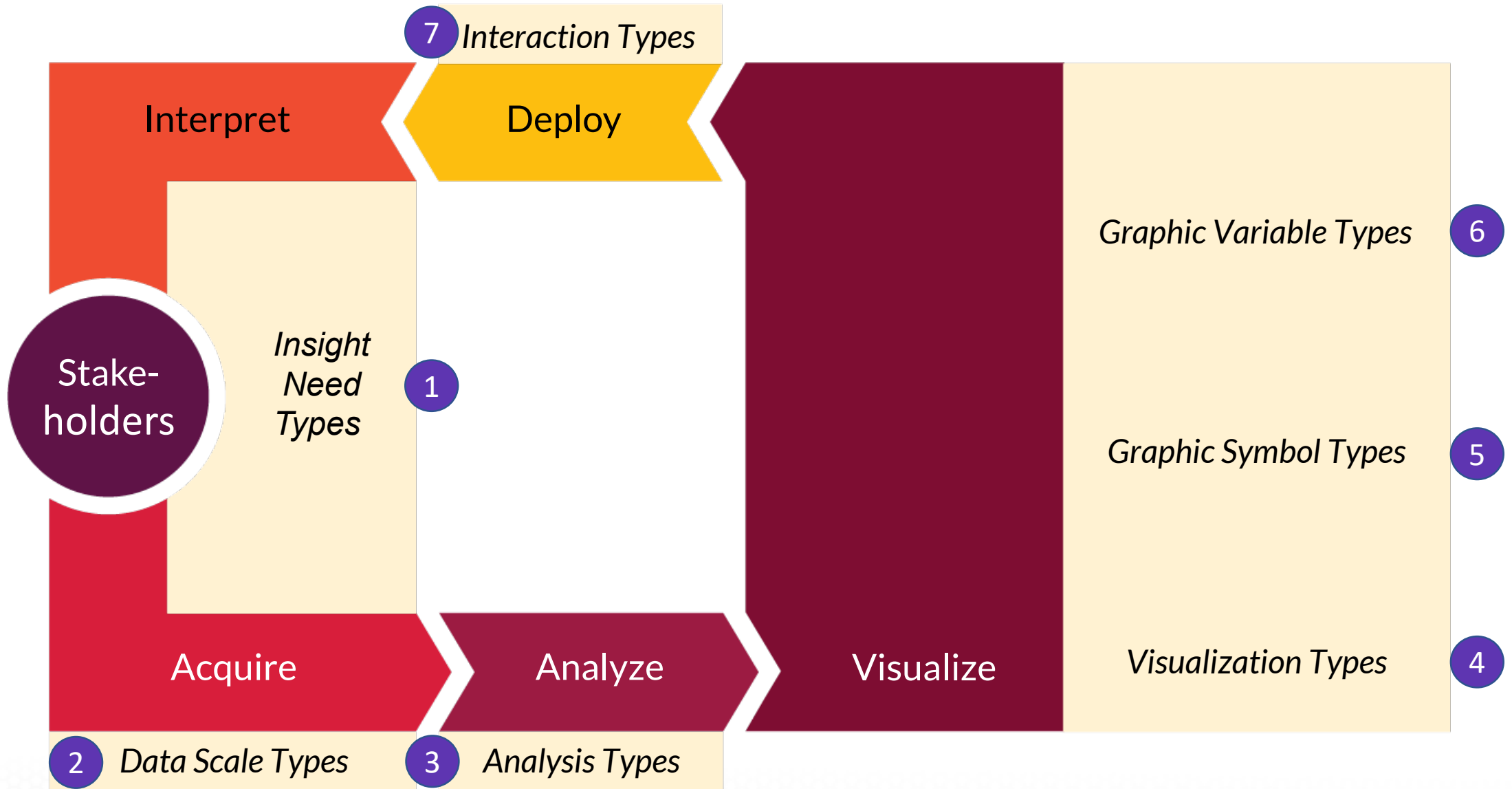
# Data Visualization Literacy Framework (DVL-FW)

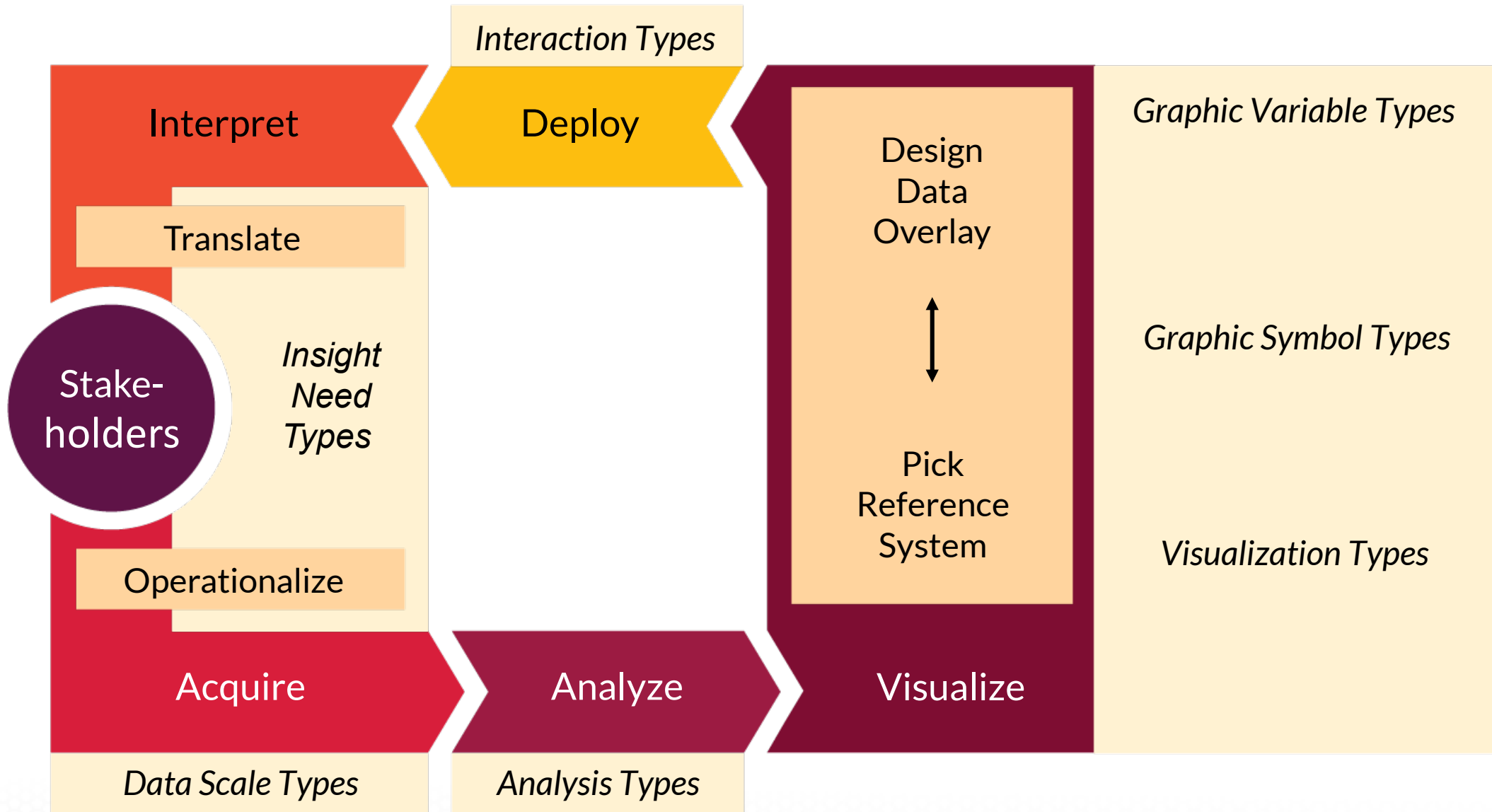
Consists of two parts *that are interlinked*:

**DVL Typology +  
DVL Workflow Process**

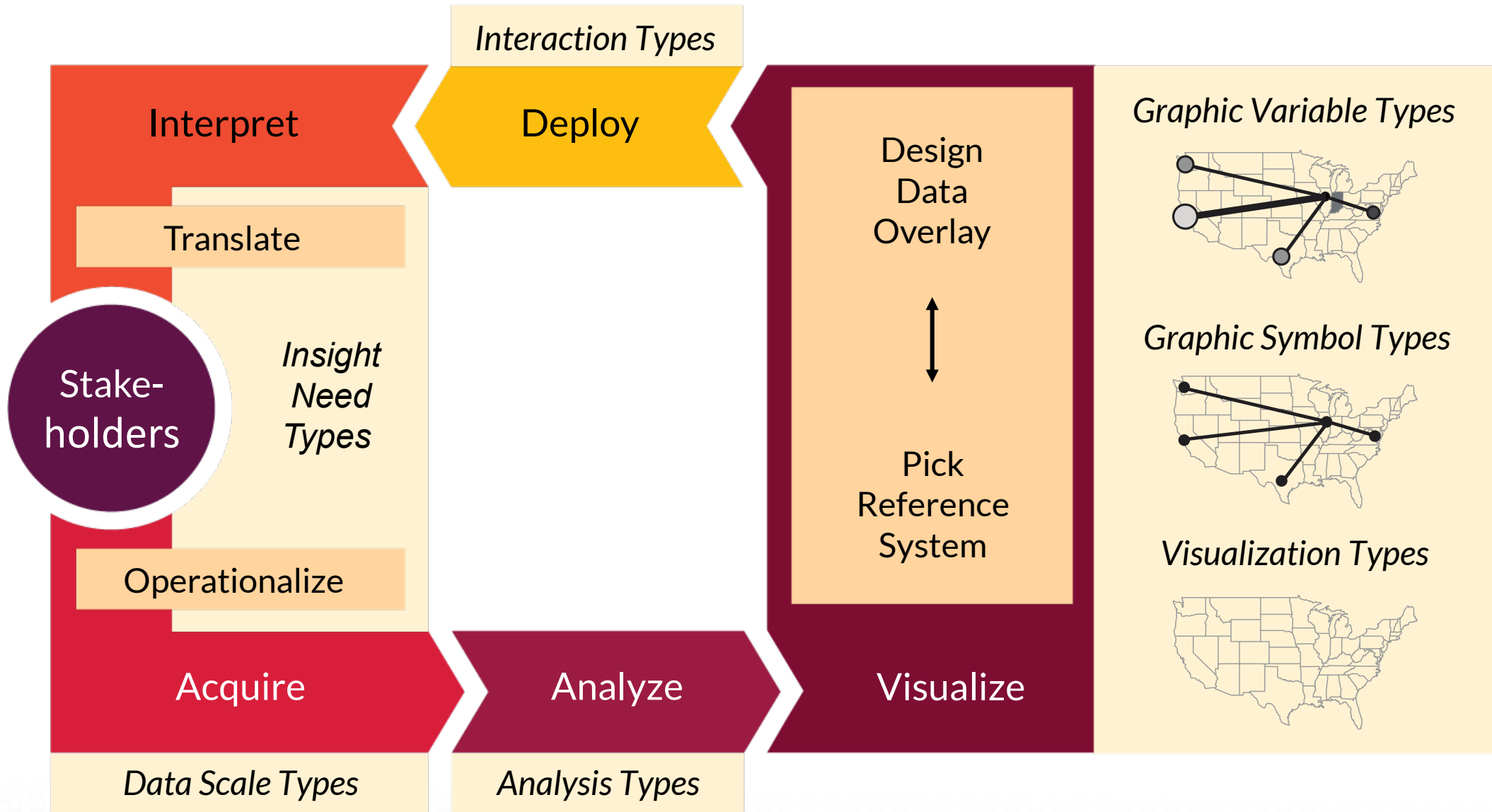












# Data Visualization Literacy Framework (DVL-FW)

Implemented in Make-A-Vis (MAV) to support learning via horizontal transfer, scaffolding, hands-on learning, etc.

☰ Make-A-Vis
i

### Data

ISI Publications: (CSV) Preprocessed-wos

Title	Authors	Journal	Year	#Cites
Total Records: 562				

Journals: (from ISI Publications)

Name	#Papers	#Cites	First Year	Last Year
BMC EVOL BIOL	1	7	2006	2006
FEBS J	2	0	2005	2005
NAT PHYS	3	18	2005	2006

Total Records: 562

### Make Visualization

Select Visualization Type

Scatter Graph

Temporal Bar Graph

Geomap

Scimap

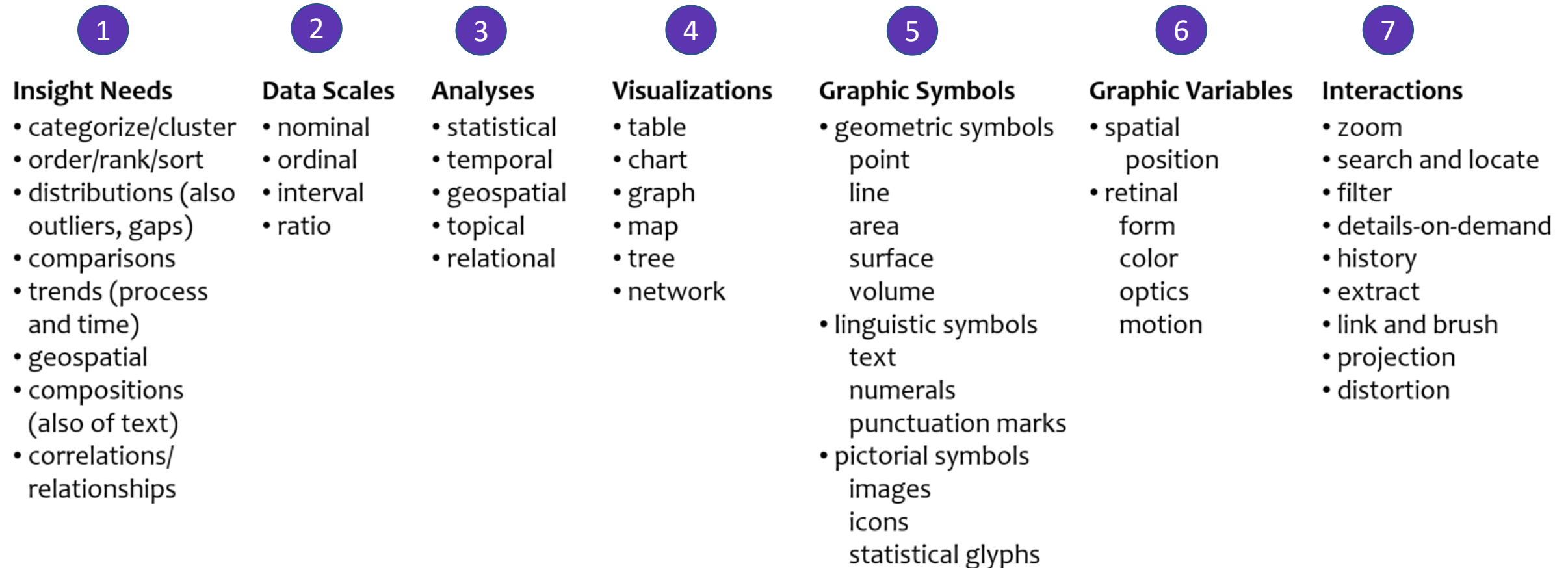
Done

Select Graphic Symbol Type(s)

Select Graphic Variable Types

Temporal Bar Graph

# Typology of the Data Visualization Literacy Framework



Börner, Katy. 2015. [Atlas of Knowledge: Anyone Can Map](#). Cambridge, MA: The MIT Press. 25.



# Typology of the Data Visualization Literacy Framework

1

## Insight Needs

- categorize/cluster
- order/rank/sort
- distributions (also outliers, gaps)
- comparisons
- trends (process and time)
- geospatial
- compositions (also of text)
- correlations/relationships

## Data Scales

- nominal
- ordinal
- interval
- ratio

## Analyses

- statistical
- temporal
- geospatial
- topical
- relational

## Visualizations

- table
- chart
- graph
- map
- tree
- network

## Graphic Symbols

- geometric symbols
  - point
  - line
  - area
  - surface
  - volume
- linguistic symbols
  - text
  - numerals
  - punctuation marks
- pictorial symbols
  - images
  - icons
  - statistical glyphs

## Graphic Variables

- spatial
  - position
- retinal
  - form
  - color
  - optics
  - motion

## Interactions

- zoom
- search and locate
- filter
- details-on-demand
- history
- extract
- link and brush
- projection
- distortion

Börner, Katy. 2015. [Atlas of Knowledge: Anyone Can Map](#). Cambridge, MA: The MIT Press. 26-27.

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

# Typology of the Data Visualization Literacy Framework

4

## Insight Needs

- categorize/cluster
- order/rank/sort
- distributions (also outliers, gaps)
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  - color
  - optics
  - motion

## Interactions

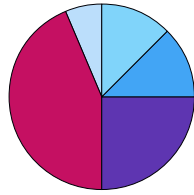
- zoom
- search and locate
- filter
- details-on-demand
- history
- extract
- link and brush
- projection
- distortion

Börner, Katy. 2015. [Atlas of Knowledge: Anyone Can Map](#). Cambridge, MA: The MIT Press. 30-31.

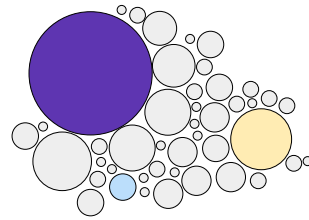


# Visualization Types

Chart

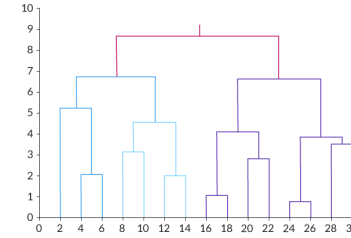


*Pie Chart*

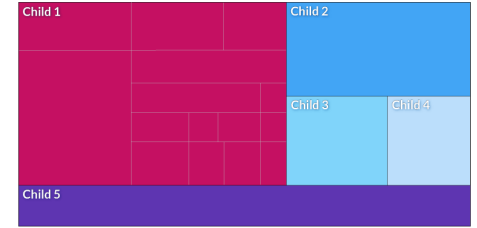


*Bubble Chart*

Tree

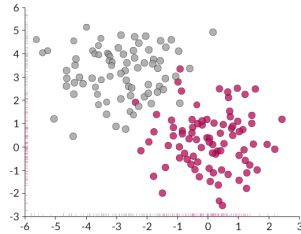


*Dendrogram*

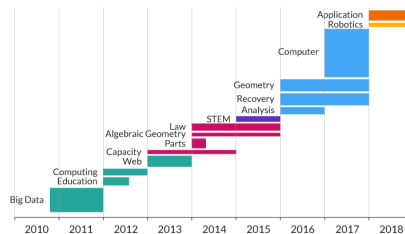


*Tree Map*

Graph

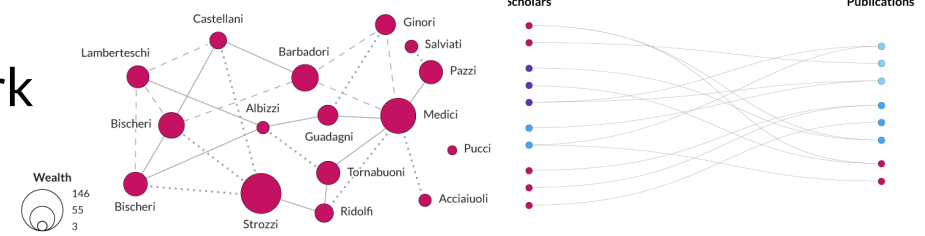


*Scatter Graph*



*Temporal Bar Graph*

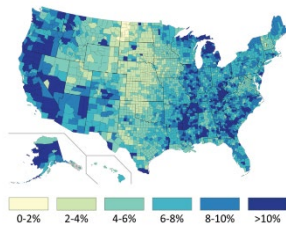
Network



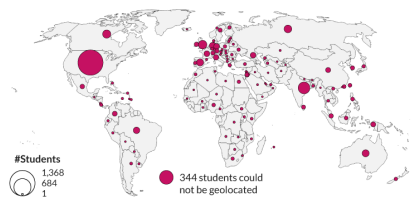
*Force-Directed Network Layout*

*Bimodal Network Layout*

Map



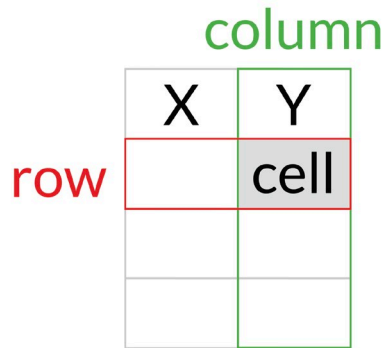
*Choropleth Map*



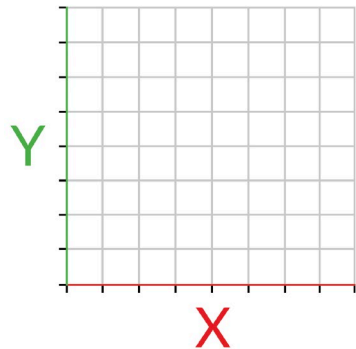
*Proportional Symbol Map*

# Visualize: Reference Systems

**Table**  
columns by rows



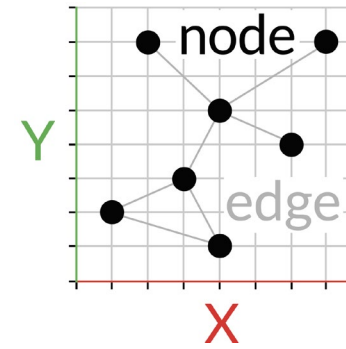
**Graph**  
x-y coordinates



**Map**  
latitude/  
longitude



**Network**  
local similarity

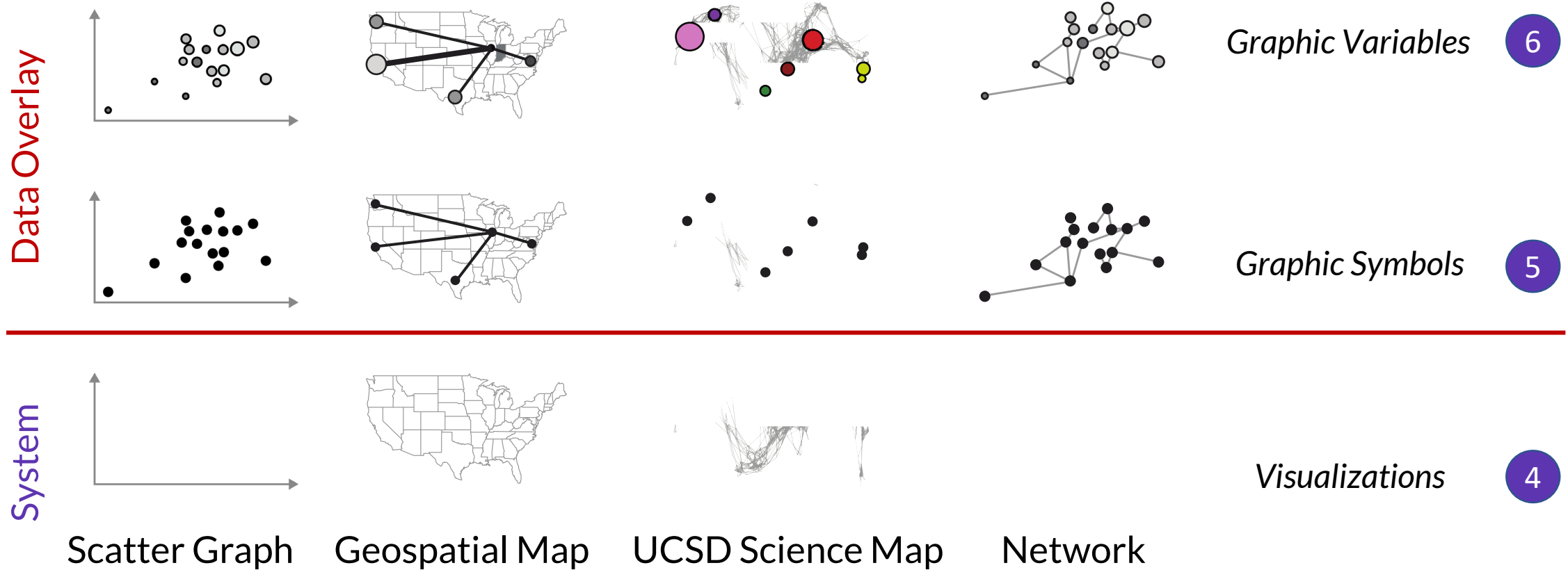


4

**Visualization Types**

- table
- chart
- graph
- map
- network layout

# Visualize: Reference Systems, Graphic Symbols and Variables





# Typology of the Data Visualization Literacy Framework

5

## Insight Needs

- categorize/cluster
- order/rank/sort
- distributions (also outliers, gaps)
- comparisons
- trends (process and time)
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- compositions (also of text)
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- linguistic symbols
  - text
  - numerals
  - punctuation marks
- pictorial symbols
  - images
  - icons
  - statistical glyphs

## Graphic Variables

- spatial
  - position
- retinal
  - form
  - color
  - optics
  - motion

## Interactions

- zoom
- search and locate
- filter
- details-on-demand
- history
- extract
- link and brush
- projection
- distortion

Börner, Katy. 2015. [Atlas of Knowledge: Anyone Can Map](#). Cambridge, MA: The MIT Press. 32-33.

# Typology of the Data Visualization Literacy Framework

6

## Insight Needs

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- link and brush
- projection
- distortion

Börner, Katy. 2015. [Atlas of Knowledge: Anyone Can Map](#). Cambridge, MA: The MIT Press. 34-35.

# Graphic Variable Types

**Position:** x, y; possibly z

**Form:**

- Size
- Shape
- Rotation (Orientation)

**Color:**

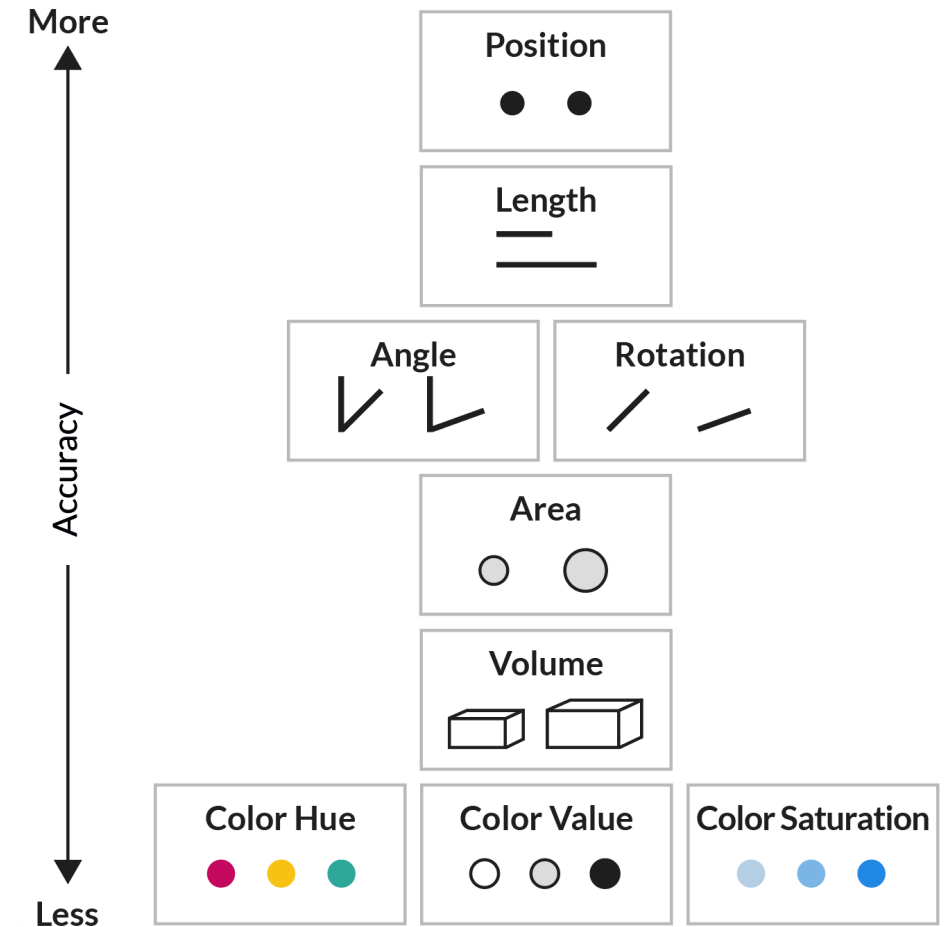
- Value (Lightness)
- Hue (Tint)
- Saturation (Intensity)



**Optics:** Blur, Transparency, Shading, Stereoscopic Depth

**Texture:** Spacing, Granularity, Pattern, Orientation, Gradient

**Motion:** Speed, Velocity, Rhythm





# Graphic Symbol Types

			Geometric Symbols		Linguistic Symbols	Pictorial Symbols
			Point	Line		
Spatial	Position	X Y				
		Retinal	Form	Size		
Shape					Text Text Text	
Color	Value				Text Text Text	
	Hue				Text Text Text	
	Saturation				Text Text Text	
Texture	Granularity					
	Pattern					
Motion Optics	Blur				Text Text Text	
	Speed					

Graphic Variable Types

See *Atlas of Knowledge* pages 36-39 for complete table.

 **Qualitative**

Also called:  
Categorical Attributes  
Identity Channels

**Quantitative**

Also called:  
Ordered Attributes  
Magnitude Channels

# Graphic Variable Types Versus Graphic Symbol Types

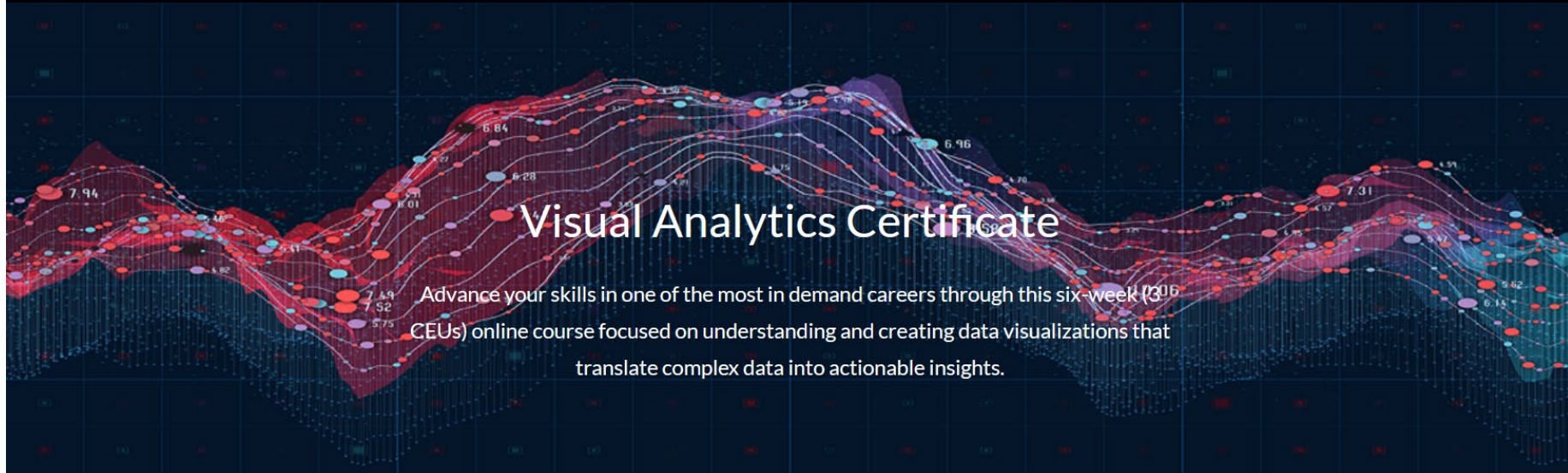
			Geometric Symbols					Linguistic Symbols Text, Numerals, Punctuation Marks					Pictorial Symbols Images, Icons, Statistical Glyphs					
			Point	Line	Area	Surface	Volume											
Spatial	x	quantitative																
	y	quantitative																
	z	quantitative																
Retinal	Form	Size	quantitative	NA (Not Applicable)														
		Shape	qualitative	NA														
		Rotation	quantitative	NA														
		Curvature	quantitative	NA														
	Angle	quantitative	NA															
	Closure	quantitative	NA															
	Value	quantitative																
	Color	Hue	qualitative															
Saturation	quantitative																	
Retinal	Texture	Spacing	quantitative															
		Granularity	quantitative															
		Pattern	qualitative															
		Orientation	quantitative	NA														
		Gradient	quantitative															
	Optics	Blur	quantitative															
		Transparency	quantitative															
		Shading	quantitative															
	Motion	Stereoscopic Depth	quantitative	Point in foreground .. background	Line in foreground .. background	Area in foreground .. background	Surface in foreground .. background	Volume in foreground .. background	Text in foreground .. background					Icons in foreground .. background				
		Speed	quantitative															
Velocity		quantitative																
Rhythm	quantitative	Blinking point slow .. fast	Blinking line slow .. fast	Blinking area slow .. fast	Blinking surface slow .. fast	Blinking volume slow .. fast	Blinking text slow .. fast					Blinking icons slow .. fast						

See Atlas of Knowledge pages 36-39 for complete table.

# Empower Yourself and Others! Data Visualization Literacy

Börner, Katy, Andreas Bueckle, and Michael Ginda. 2019. Data visualization literacy: Definitions, conceptual frameworks, exercises, and assessments. *PNAS*, 116 (6) 1857-1864.





# Visual Analytics Certificate

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FAQS



## Learn from Experts

Connect with industry professionals and leading researchers.



## Evolve Yourself

Gain forever knowledge and skill-up in powerful data visualization tools.



## Make a Difference

Embrace data-driven decision-making in your personal and professional life.

<https://visanalytics.cns.iu.edu>

US Employers which have sent students include  
**The Boeing Company, Eli Lilly, DOE, CDC, NSWC Crane.**

# 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](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>



# Q&A

