

(Data) Visualization Literacy

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



Visualization Literacy for General Audiences - Can We Make A Difference? Panel at IEEE Vis2021 Conference
<http://ieevis.org/year/2021/info/panels>

October 27, 2021



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.

Börner, Katy (2015) *Atlas of Knowledge: Anyone Can Map*. The MIT Press.

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).

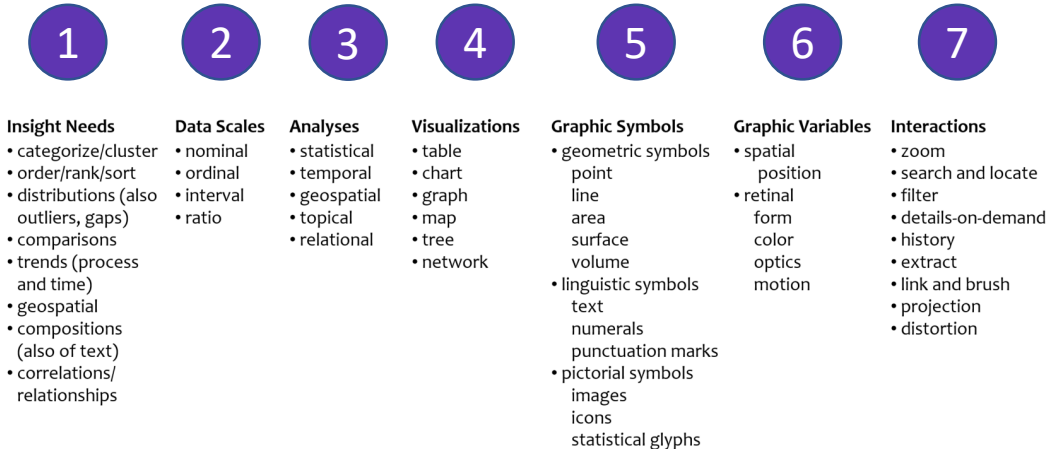
Being able to “read and write” data visualizations is becoming as important as being able to read and write text. Understanding, measuring, and improving data and visualization literacy is important to strategically approach local and global issues.

Data Visualization Literacy Framework (DVL-FW)

Consists of two parts:

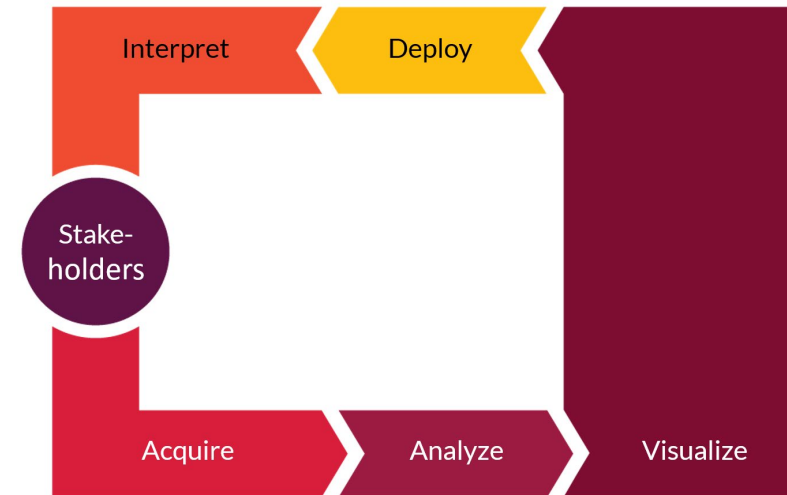
DVL Typology

Defines 7 types with 4-17 members each.



DVL Workflow Process

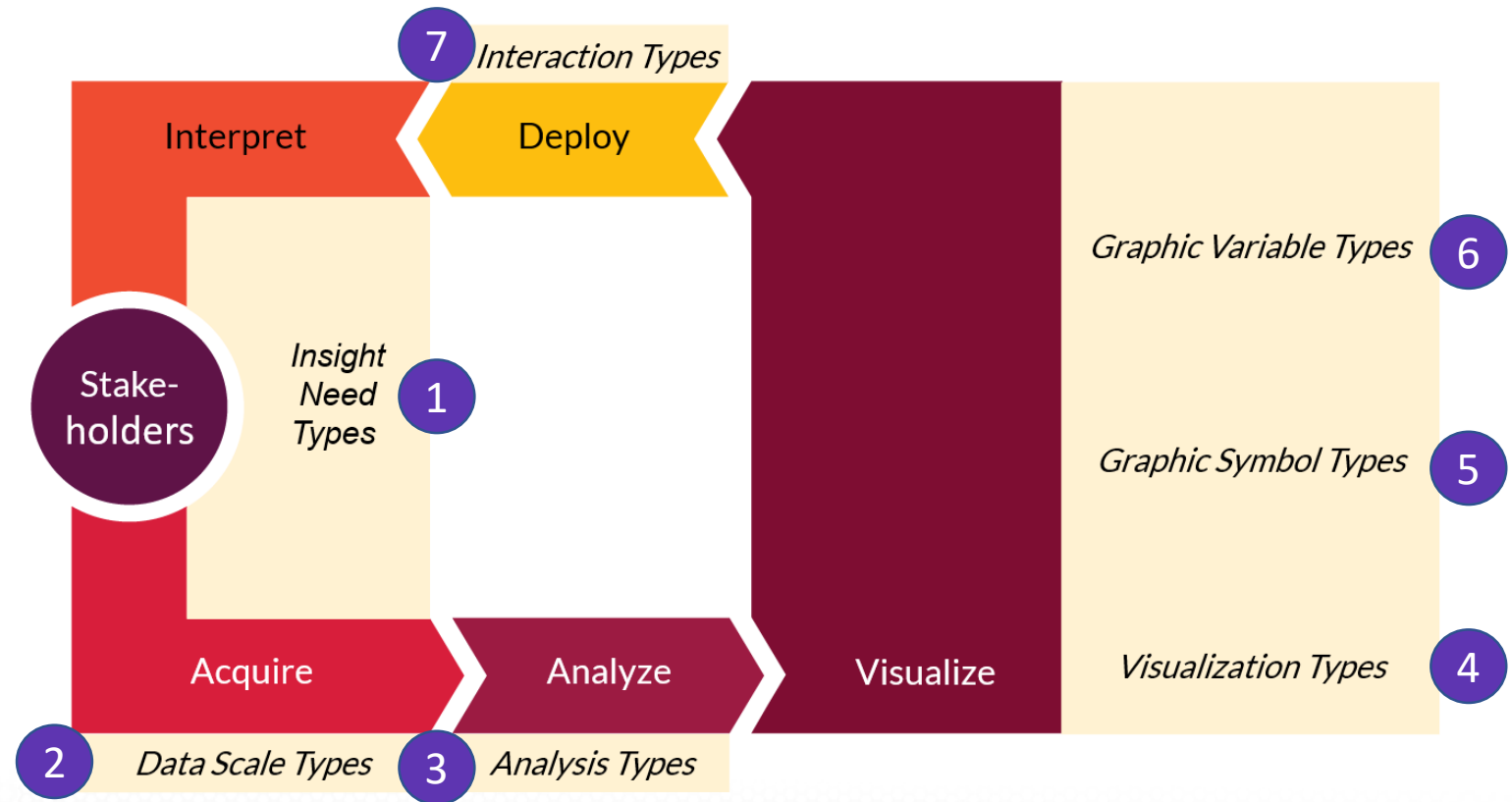
Defines 5 steps required to render data into insights.



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.

The screenshot shows the Make-A-Vis interface with three main sections: Data, Make Visualization, and a visualization preview.

Data Section:

- ISI Publications: (CSV) Preprocessed-wos**

Title	Authors	Journal	Year	#Cites
[Progress bar]				

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 Section:

- Select Visualization Type:** Scatter Graph, Geomap, Scimap, Temporal Bar Graph (selected).
- Select Graphic Symbol Type(s):** (Dropdown menu)
- Select Graphic Variable Types:** (Dropdown menu)

Temporal Bar Graph Preview:

The visualization shows a horizontal bar chart from 1998 to 2017. The bars represent different categories over time:

- Machine (1998-2002)
- Big Data (2000-2002)
- Education (2000-2001)
- Building (2000-2001)
- Making (2001-2002)
- Computing (2001-2002)
- Web (2002-2003)
- Form (2002-2003)
- Smart (2003-2004)
- Capacity (2004-2005)
- Algebraic Geometry (2005-2006)
- Parts (2006-2007)
- Law (2007-2008)
- Stem (2008-2009)
- Analysis (2009-2011)
- Recovery (2011-2012)
- Geometry (2012-2016)
- Computer (2013-2014)
- Application (2014-2015)
- Robotics (2015-2017)

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

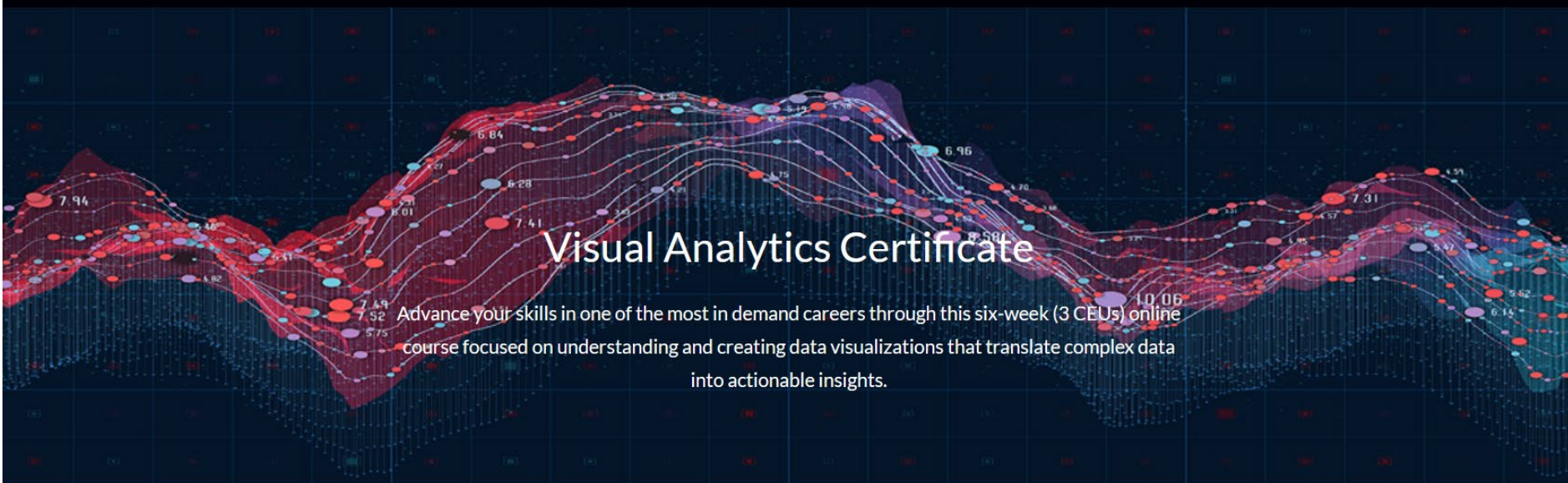
Quantitative

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.



Visual Analytics Certificate

Advance your skills in one of the most in demand careers through this six-week (3 CEUs) online course focused on understanding and creating data visualizations that translate complex data into actionable insights.

FLYER

REGISTER FOR JAN 10–FEB 20, 2022

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>

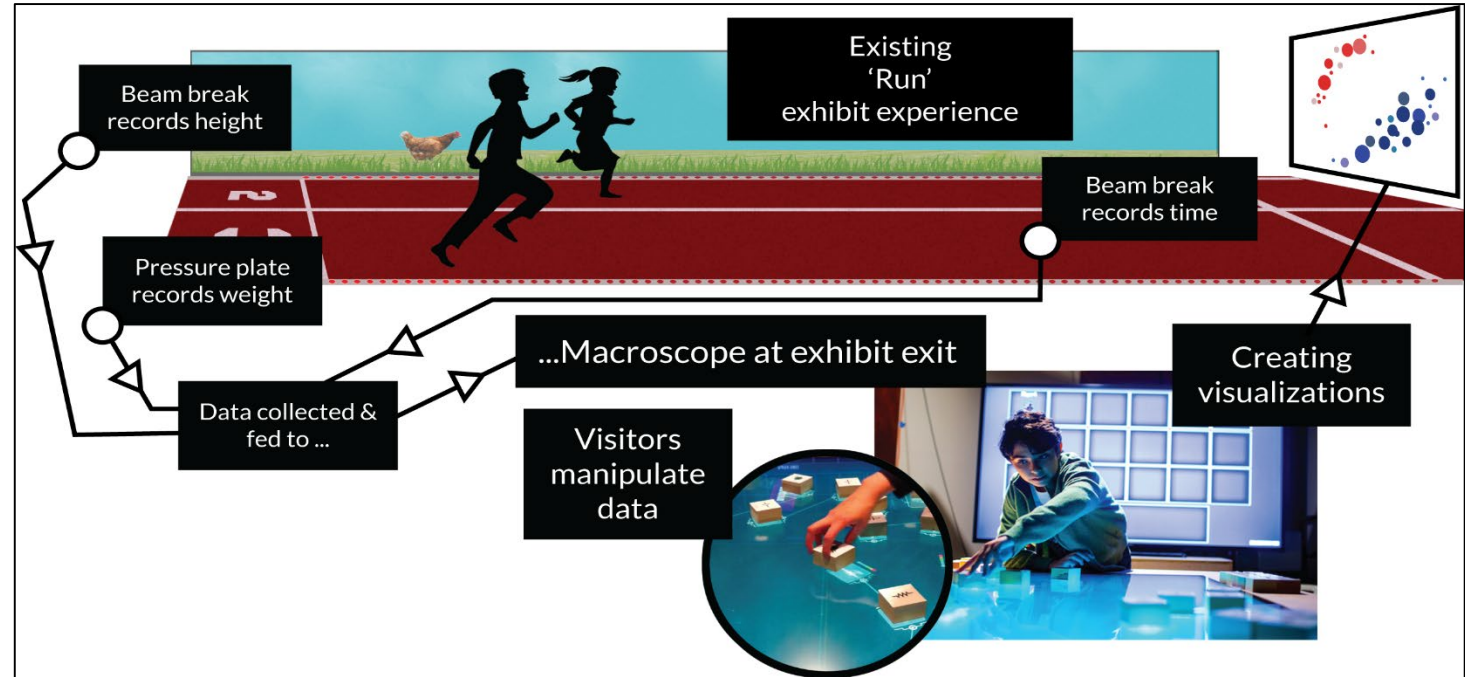
US Employers which have sent students include
The Boeing Company, Eli Lilly, DOE, CDC, NSWC Crane.

Teaching Data Visualization Literacy

in Science Museums

















xMacrosopes in Science Museums















Data Visualization Literacy: Research and Tools that Advance Public Understanding of Scientific Data.
NSF AISL #1713567

DATA TABLE

	Age	Height (inches)	Time (seconds)	Opponent	Shoes	Zip
	2	32	.5	Visitor	Paws	47401
	7	6	136	Visitor	Paws	47402
	1					
	32					
	64					
	7					
	12					
	15					
	72					
	80					
	7					
	11					
	9					
	0					

Find and Select Your Data Record

-  1
-  32
-  64
-  7
-  12
-  15
-  72
-  80
-  7
-  11
-  9
-  0

Sort by

Age

SCATTER GRAPH

Y Axis

Age

Height

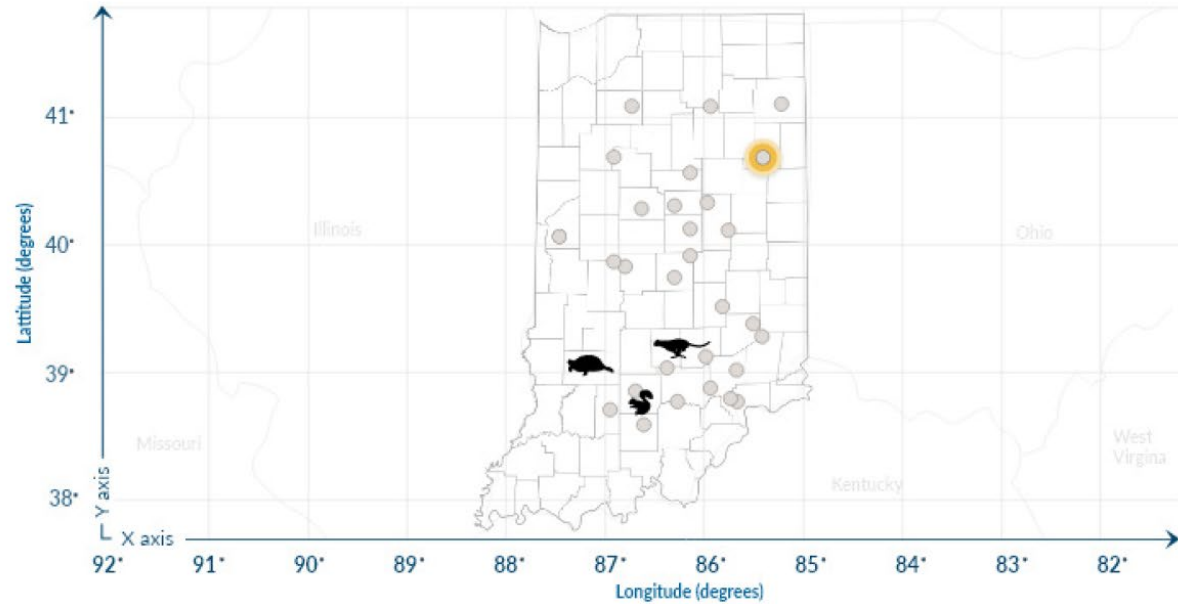
Time



MAP

United States

Indiana



Color

Opponent

Shoes

Shape

Opponent

Shoes

Size

Age

Height

Time



HOME

Data Table



HOME



HOME

Data Table

Scatter Graph

Map

Color: Opponent

- Turtle
- Squirrel
- Cheetah

Shape: Shoes

- Paws
- Wheels
- Sneakers
- Casual
- + Dress
- * Heels

Size: Age

- 5
- 23
- 78



Data Visualization Literacy
NSF AISL #1713567

Investigating Aspects of Data Visualization Literacy Using 20 Information Visualizations and 273 Science Museum Visitors

<https://cns.iu.edu/2015-VisLit.html>

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

Abstract:

In the information age, a person's ability to read and make data visualizations is nearly as important as being able to read and write text. This paper reports the results of a multi-phase study conducted in informal learning environments in three U.S. science museums. The goal of the study was to determine the familiarity of youth and adult museum visitors with different visualization types. To address this, a total of 273 visitors were shown five out of 20 different visualizations that included two charts, five maps, eight graphs, and five network layouts. They were asked to judge the familiarity of the visualization, provide information on how to read it, and to provide a name, identify typical locations where they would encounter the data display and possible data sources that might be visualized in this way.

Results show that while most participants have a strong interest in science, math and art, many have a hard time naming and interpreting visualizations. Participants in this study commonly encounter visualizations in school, in books, at work, on the Internet, and in the news. Overall they were more familiar with basic charts, maps and graphs, but very few are familiar with network layouts and most have no ability in reading network visualizations. When asked how they would interpret the visualizations, most participants pointed to superficial features such as color, lines, or text as important to developing understanding. Overall, we found that participants were interested in the visualizations we presented to them, but had significant limitations in identifying and understanding them.

The results substantiate intuitions shared by many regarding the rather low level of data visualization literacy of general audiences. We hope they will help catalyze novel research on the development of easy-to-use yet effective visualizations with standardized names and guaranteed properties that can be readily used by those interested to understand and solve real world problems. Results also have implications for how information visualizations are taught and used in formal and informal education, the media, or in different professions.

Links:

- [Data Collection Basics](#)
- [Instructions for Completing the Interview](#)
- [Data Collection Form](#)
- [20 Visual Stimuli \(see Figure 1\)](#)
- [Refusal Log](#)
- [Blank Data Entry Spreadsheet](#)



Data Visualization Literacy: Research and Tools that Advance Public Understanding of Scientific Data.
NSF AISL #1713567

Visualizations of the Scalable Precision Medicine Knowledge Engine (SPOKE)

<https://spoke.ucsf.edu>

Scalable Precision
Medicine Knowledge
Engine

Search... 

Data & Tools

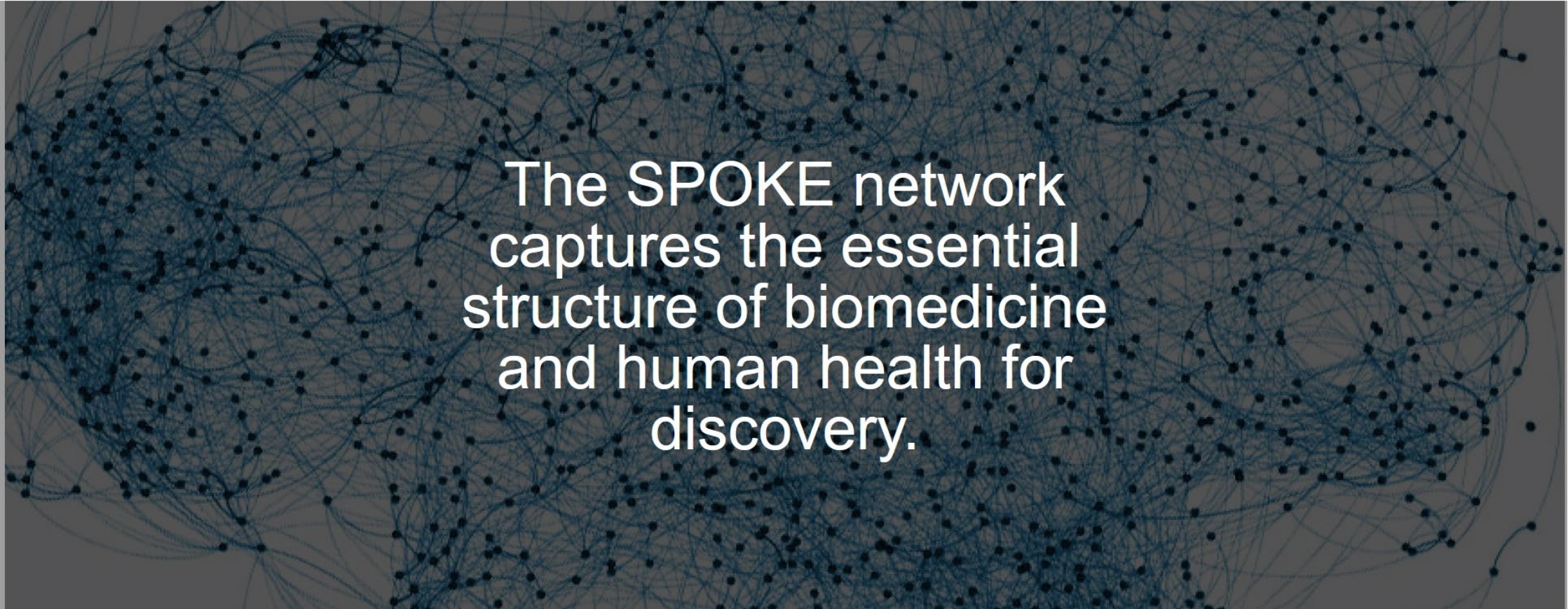
Neighborhood Explorer

Funding

Applications

People

Publications



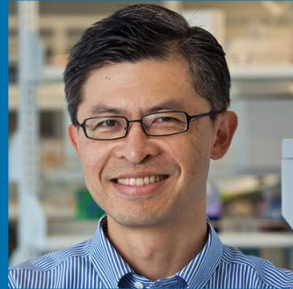
The SPOKE network
captures the essential
structure of biomedicine
and human health for
discovery.

<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

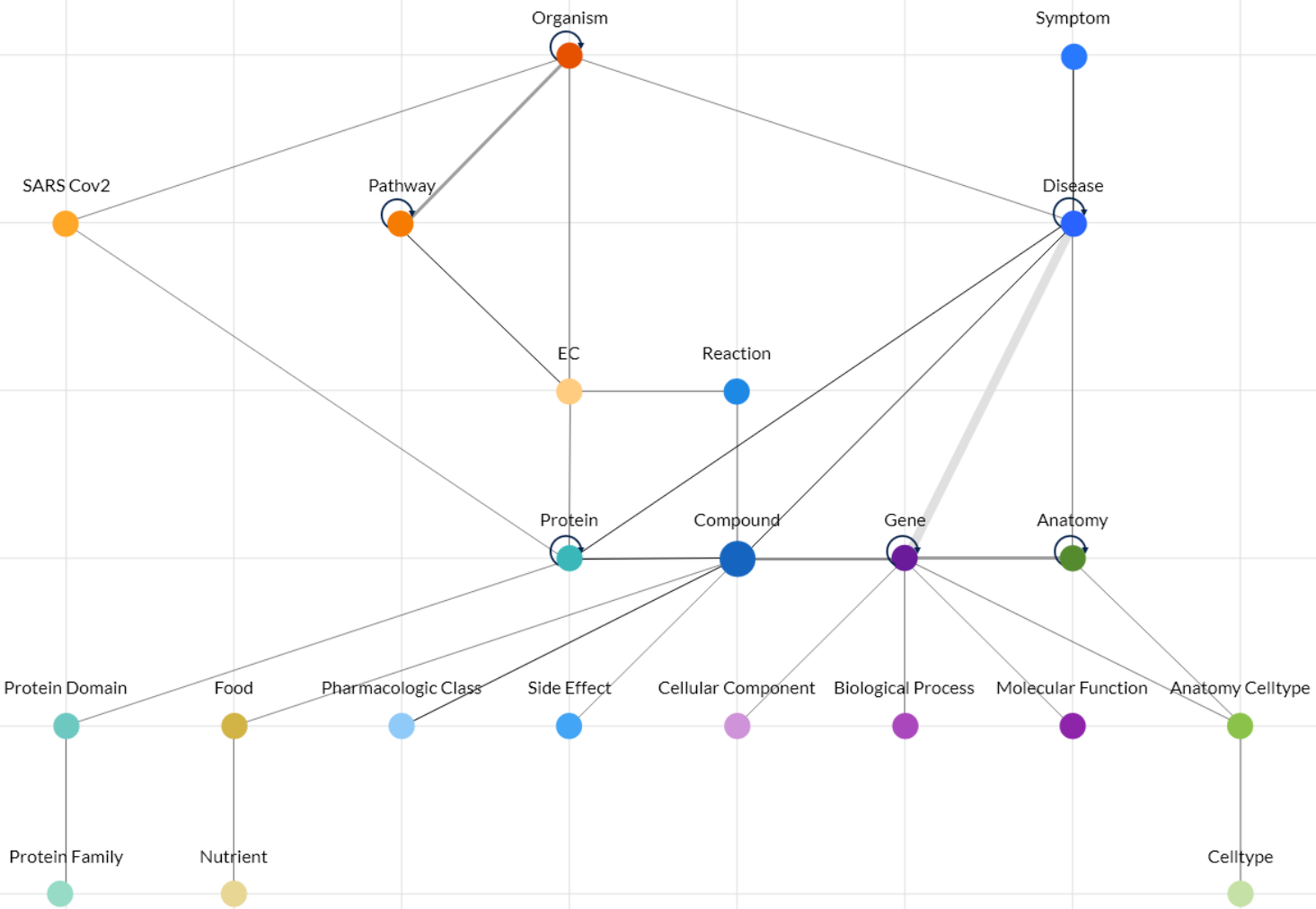
D D

E E

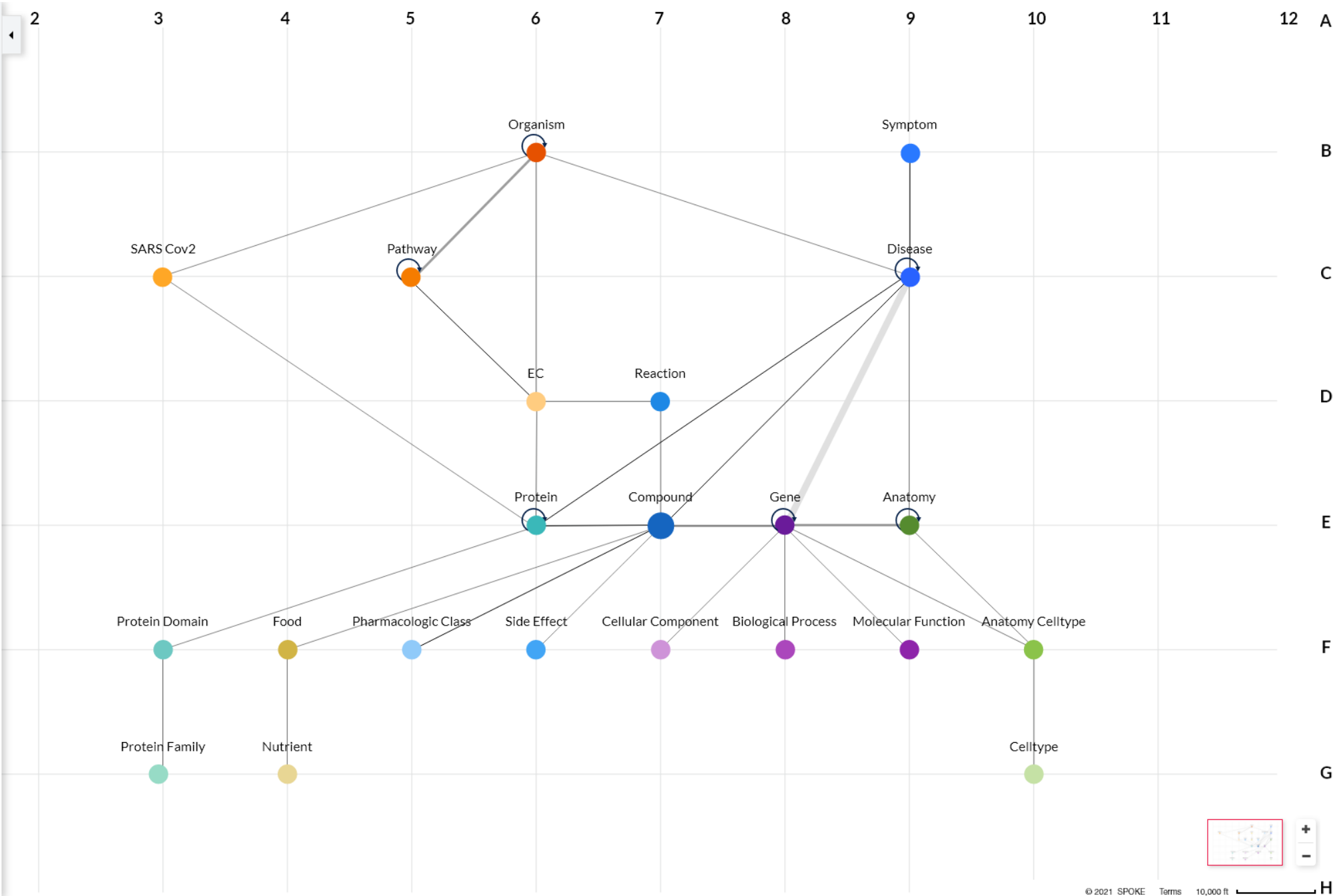
F F

G G

H H



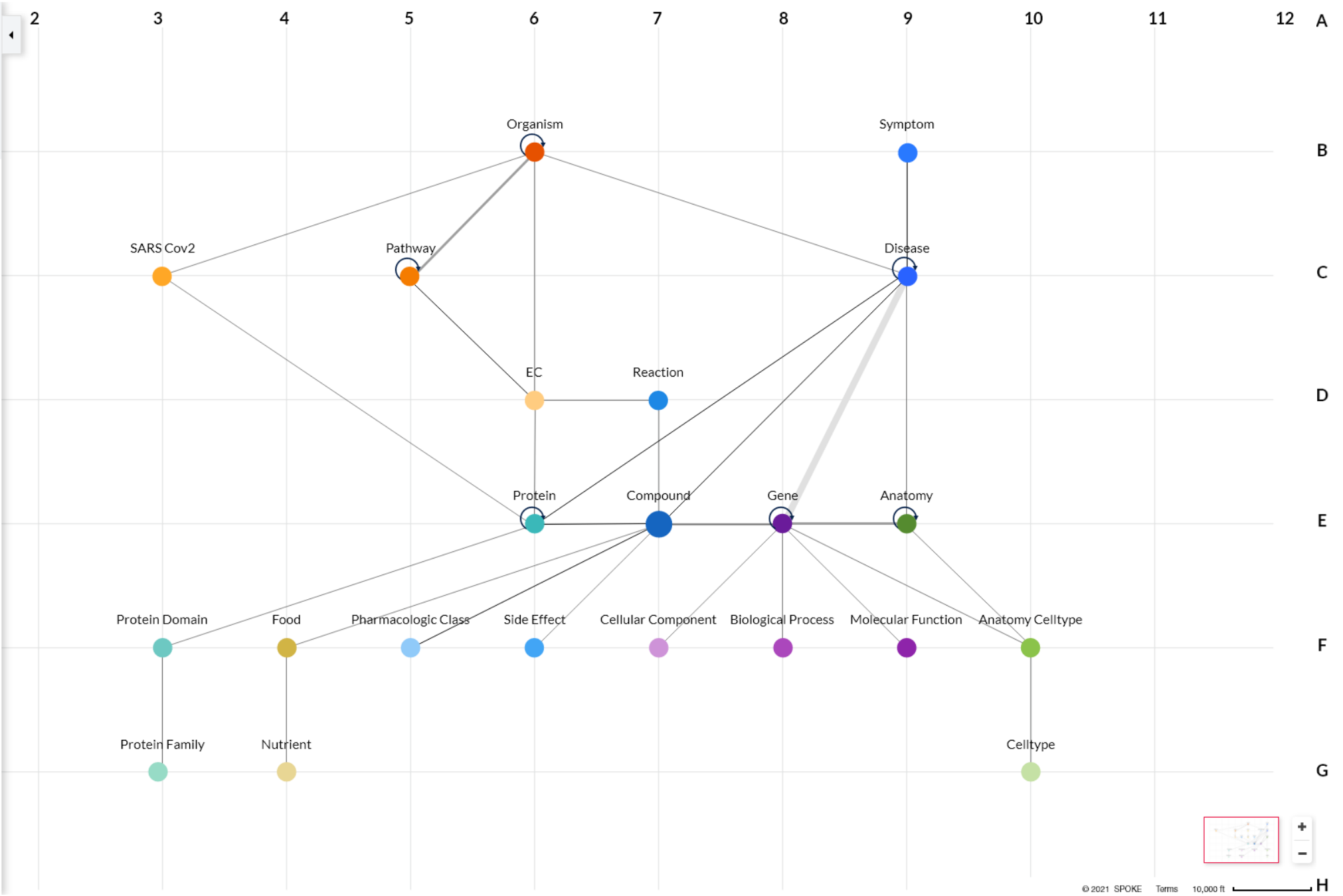
- potato
- heart
- alcoholic cardiomyopathy
- anterolateral myocardial infarction
- atrial fibrillation
- beta thalassemia
- brugada syndrome
- cardiac arrest
- cardiac tuberculosis
- coronary artery disease
- diabetic neuropathy
- diastolic heart failure
- ...



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 ✕

Favorite Similar Send to phone Share

Show Details



Send feedback



potato ✕

coronary artery disease ✕

SEARCH

Favorite Similar Send to phone Share

Show Details



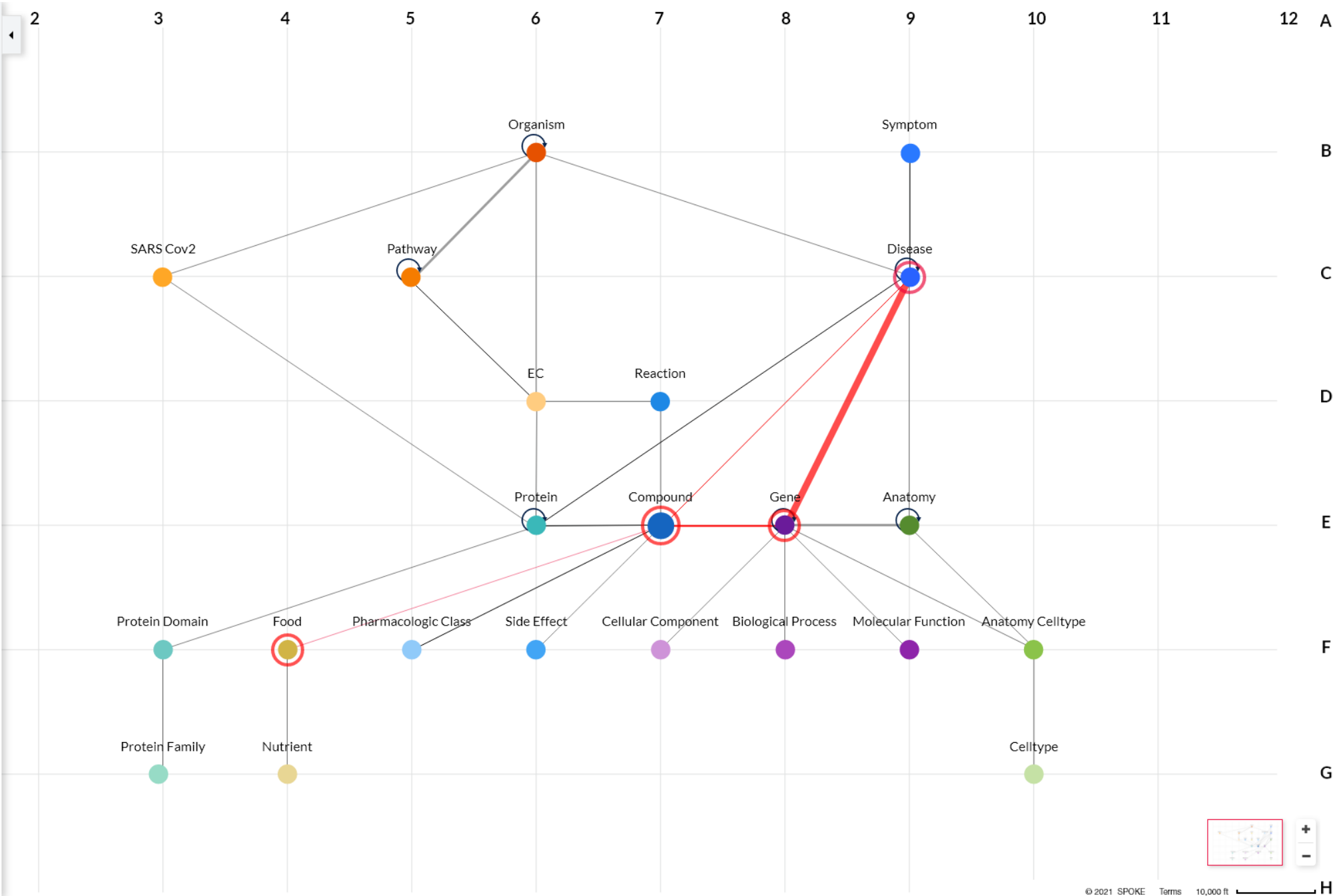
Send feedback

potato ✕

coronary artery disease ✕

SEARCH

Favorite Similar Send to phone Share



Send feedback



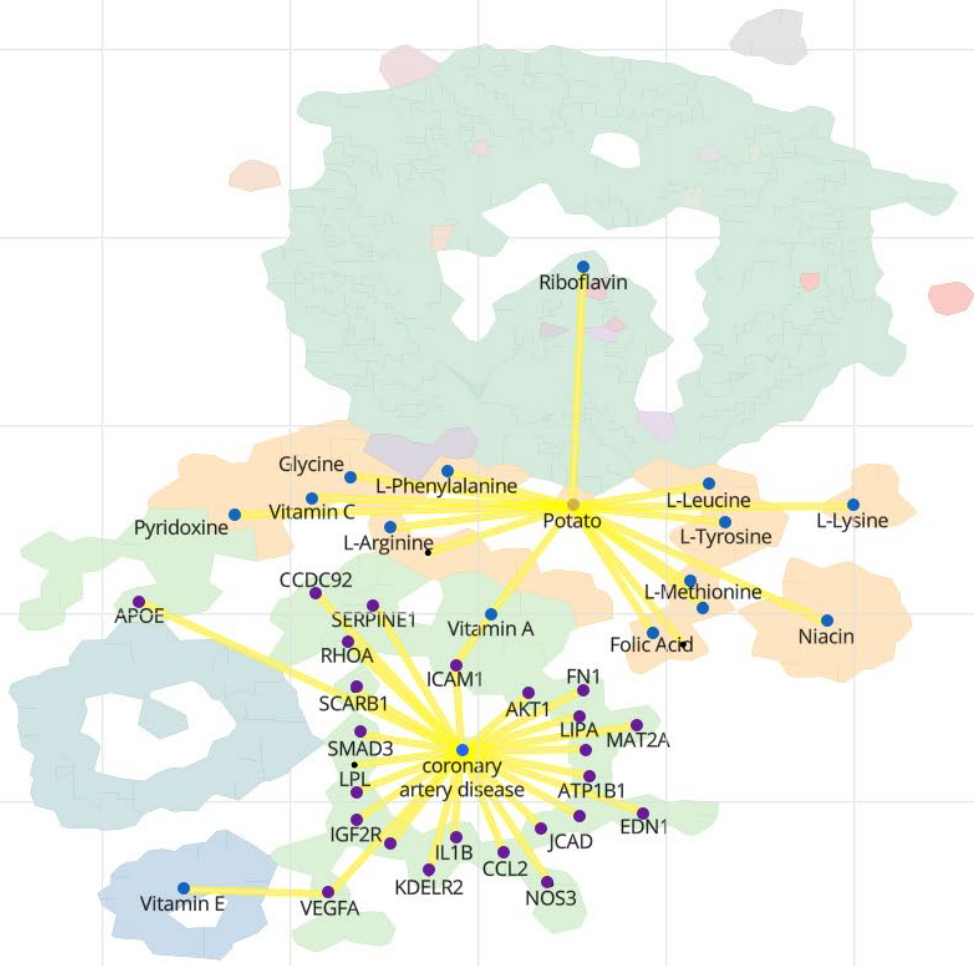
potato ✕

coronary artery disease ✕

SEARCH

Favorite Similar Send to phone Share

Show Overview



● Disease ● Compound ● Food ● Gene

Send feedback



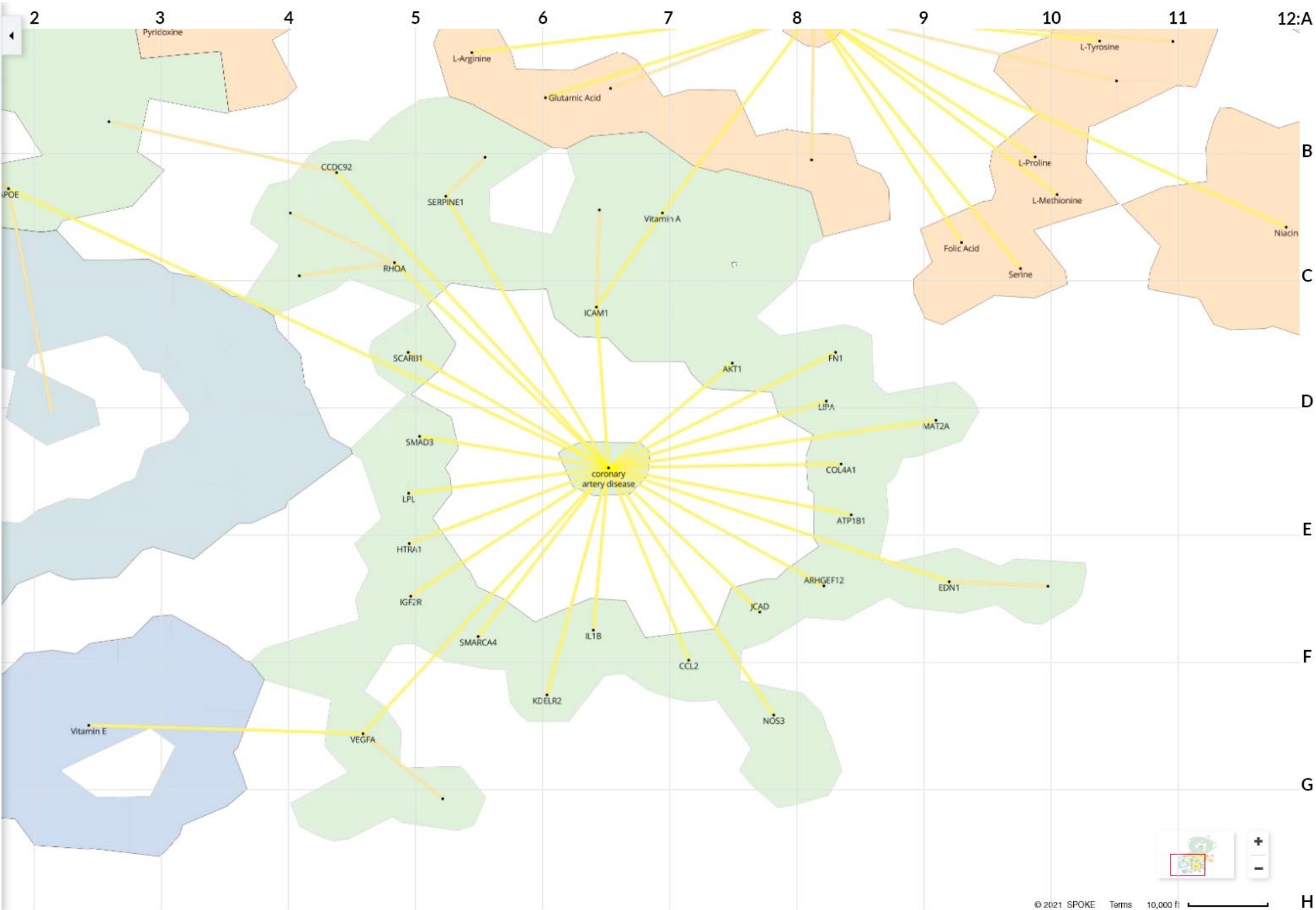
potato ×

coronary artery disease ×

SEARCH

Favorite Similar Send to phone Share

Show Overview



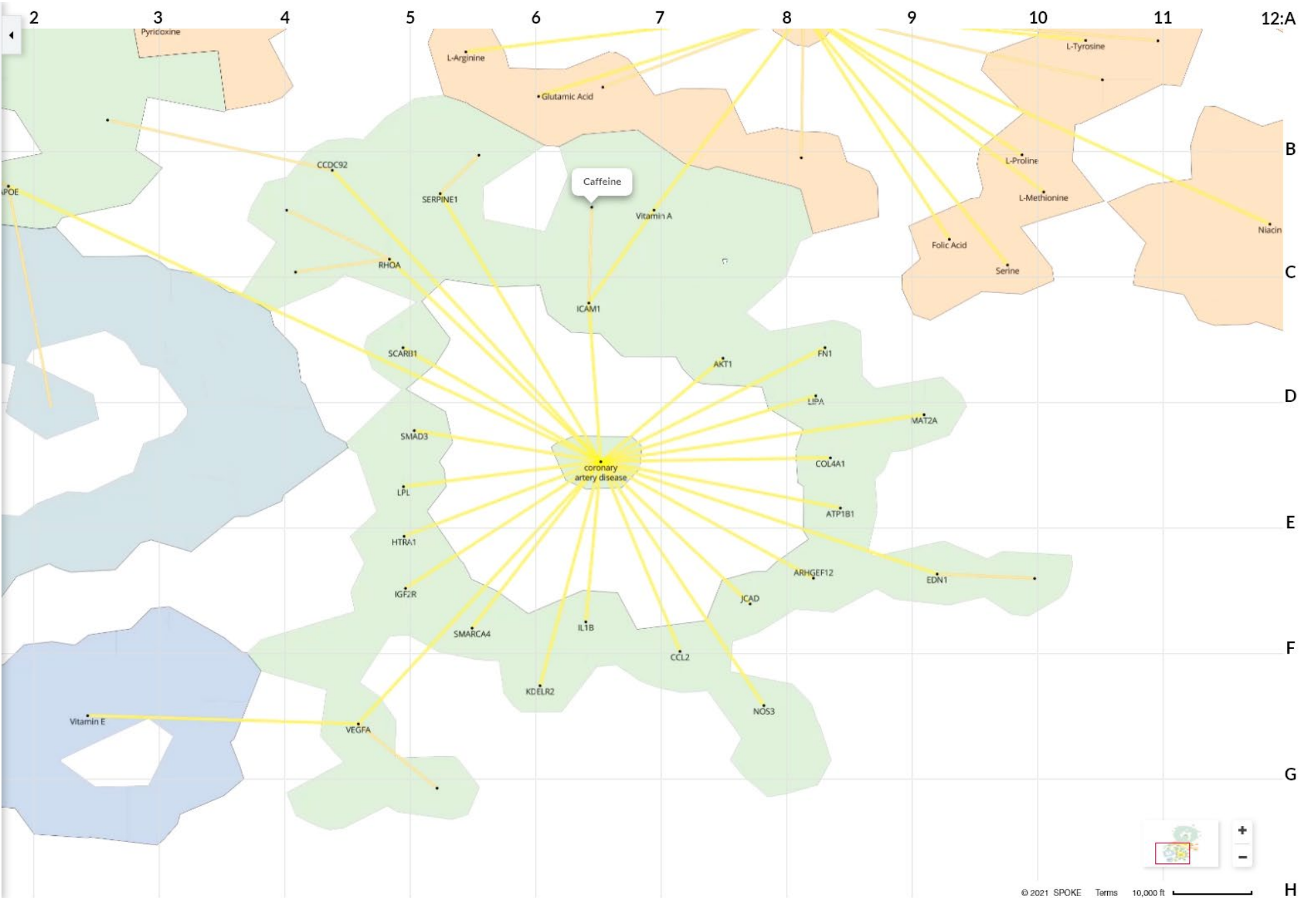
potato ×

coronary artery disease ×

SEARCH

Favorite Similar Send to phone Share

Show Overview



Send feedback

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>

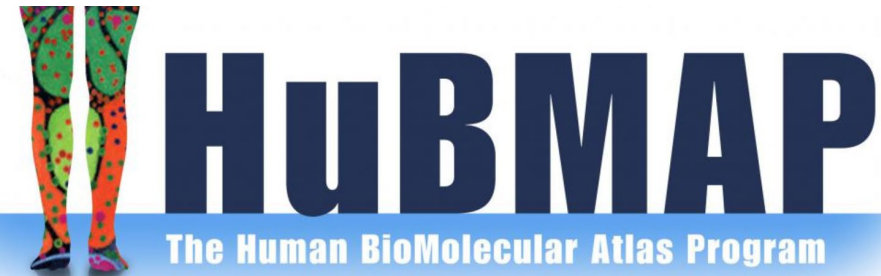
HuBMAP

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.

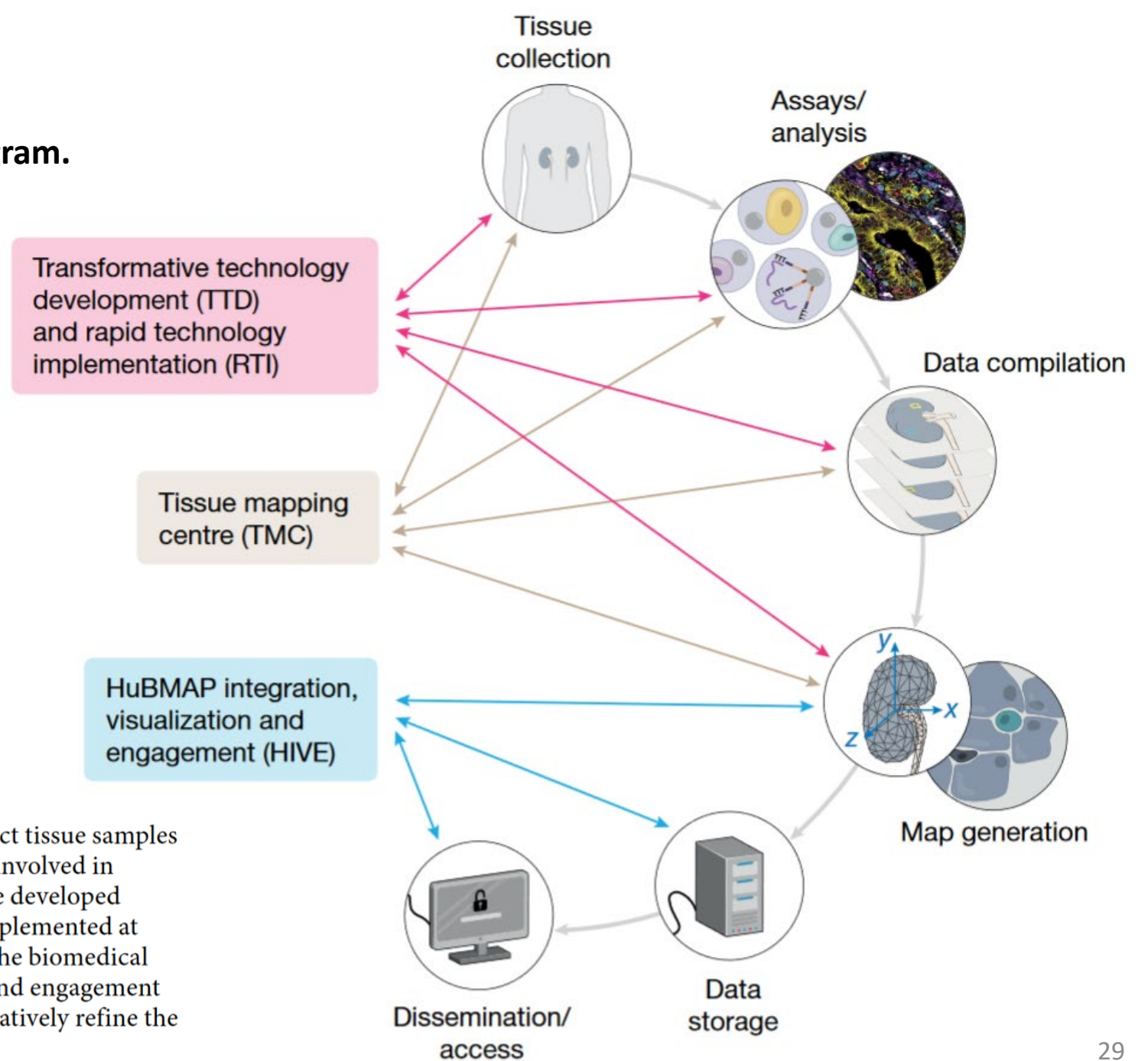


Fig. 1 | The HubMAP consortium. The TMCs will collect tissue samples and generate spatially resolved, single-cell data. Groups involved in TTD and RTI initiatives will develop emerging and more developed technologies, respectively; in later years, these will be implemented at scale. Data from all groups 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.

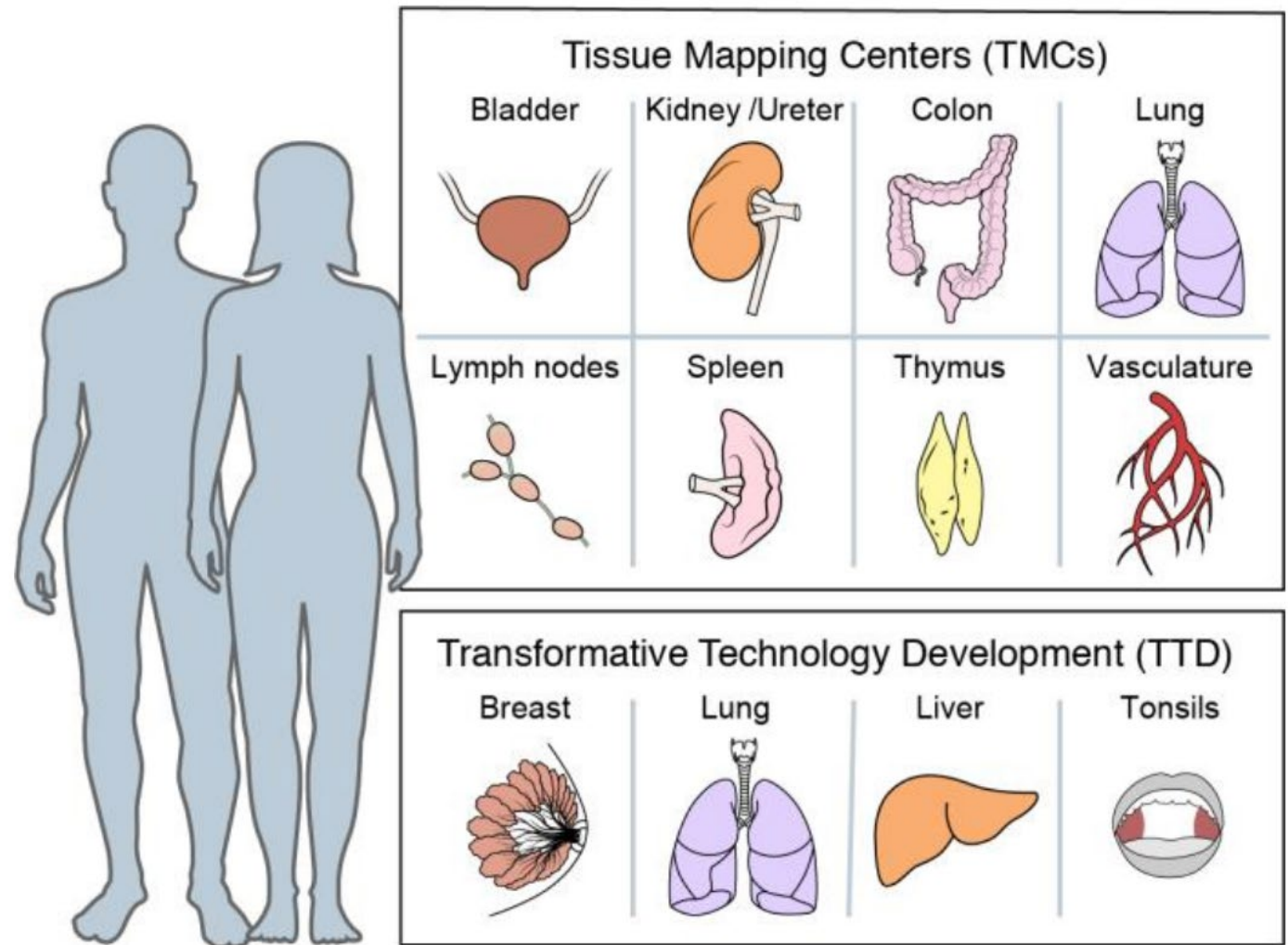


Fig. 2 | Key tissues and organs initially analysed by the consortium.

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.

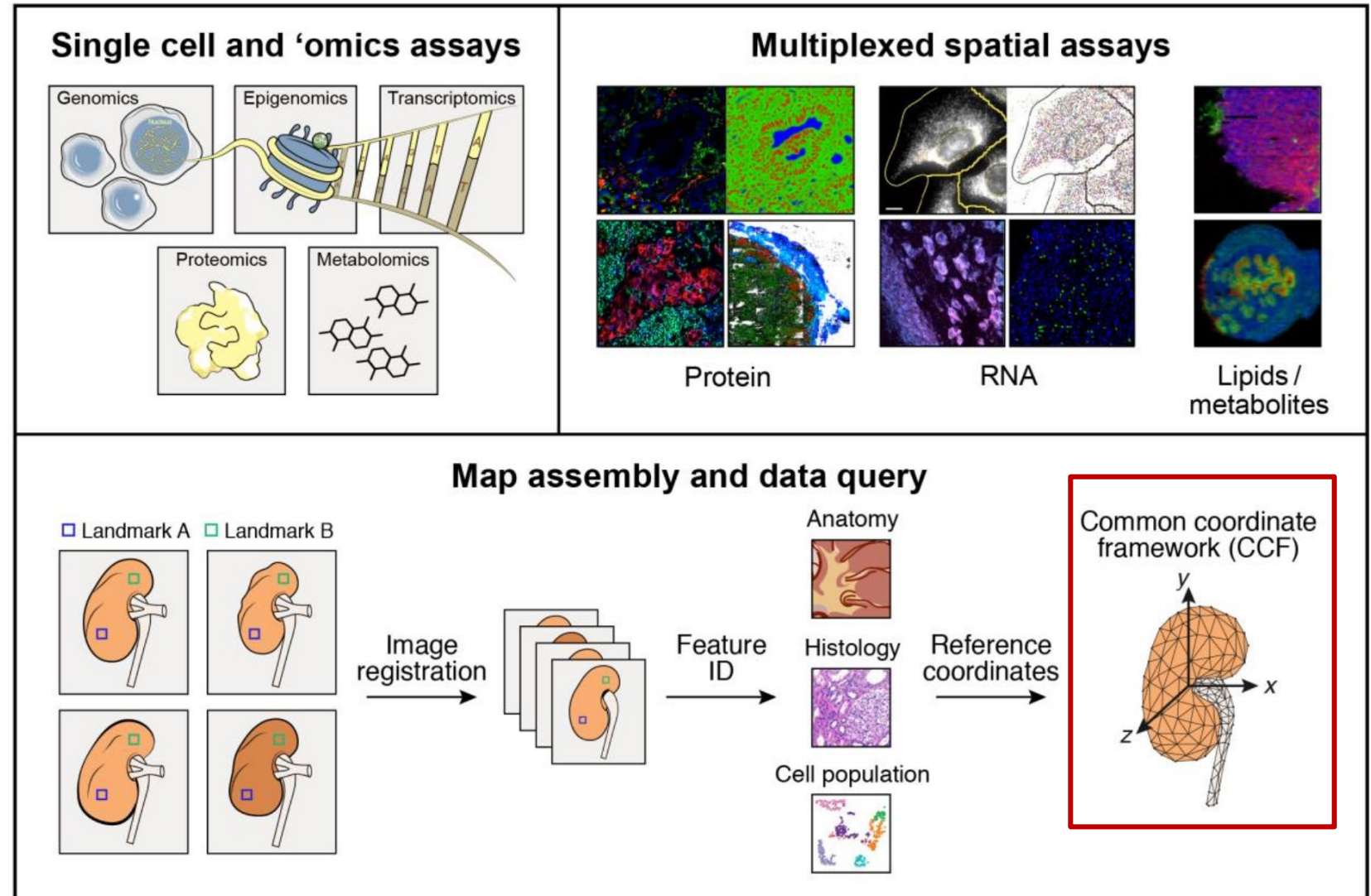


Fig. 3 | Map generation and assembly across cellular and spatial scales. HuBMAP aims to produce an atlas in which users can refer to a histological slide from a specific part of an organ and, in any given cell, understand its contents on multiple 'omic levels—genomic, epigenomic, transcriptomic, proteomic, 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.

What is a Human Reference Atlas?

The Human Reference Atlas (HRA)

1. defines the 3D space and shape of anatomical structures and cell types that are of biomedical relevance plus the biomarkers used to characterize them. Anatomical structures, cell types and biomarkers are validated and represented in/added to ontologies (Uberon/FMA, CL, HGNC).
2. defines how new datasets can be mapped to the HRA, e.g., spatially using the Visible Human CCF and/or Vasculature CCF, via ASCT+B ontology terms/IDs such as gene or protein biomarkers, or via gene expression data as in Azimuth.
3. it is
 - authoritative (there exists expert agreement and it was validated by data),
 - computable (supports API queries, UIs, linkages; see slides #8 and #9),
 - published as LOD (connecting to disease and other ontologies and data),
 - open (anyone can use the HRA data and code), and
 - continuously evolving (e.g., as new technologies become available).

Börner, Katy, Sarah Teichmann, Ellen M. Quardokus, ..., Sanjay Jain, Griffin M. Weber. 2021. "Human Anatomical Structures, Cell Types, and Biomarkers of the Human Reference Atlas". *Nature Cell Biology - Accepted*

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

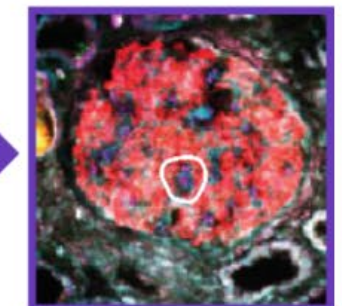
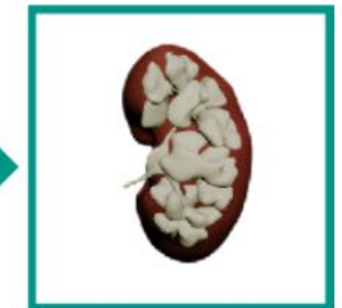
Structure/Region	Sub structure/Sub region	Cell Type
Renal Corpuscle	Bowman's (glomerular) Capsule/parietal layer	Parietal epithelial Cell
	Bowman's (glomerular) Capsule/visceral layer	Podocyte
	Glomerular Tuft	Capillary Endothelial Cell Mesangial Cell
Tubules	Proximal Tubule	Proximal Tubule Epithelial 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)		
DCT Type 1 Cell		
Connecting Tubule	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

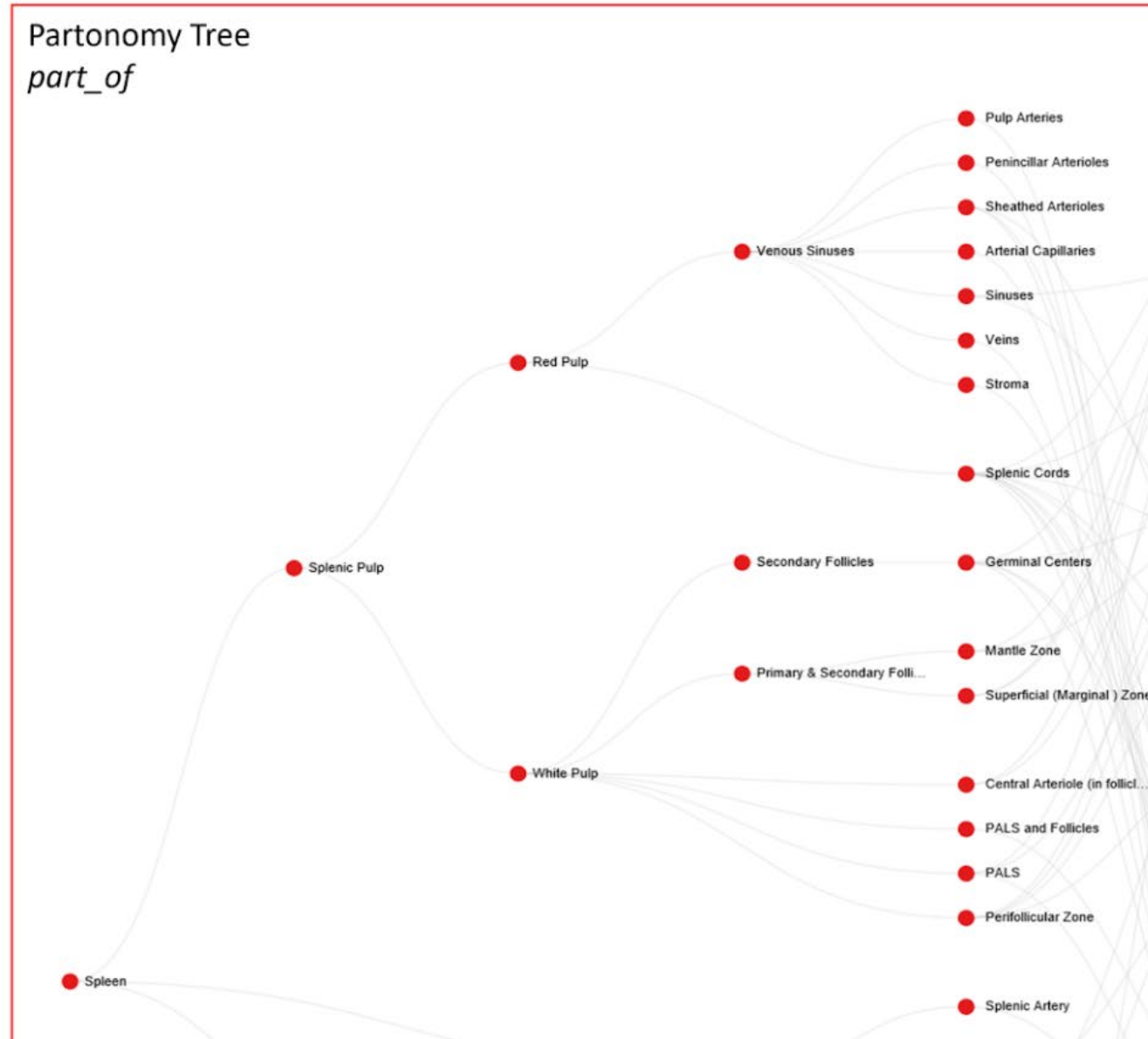


Anatomical Structures (AS)

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

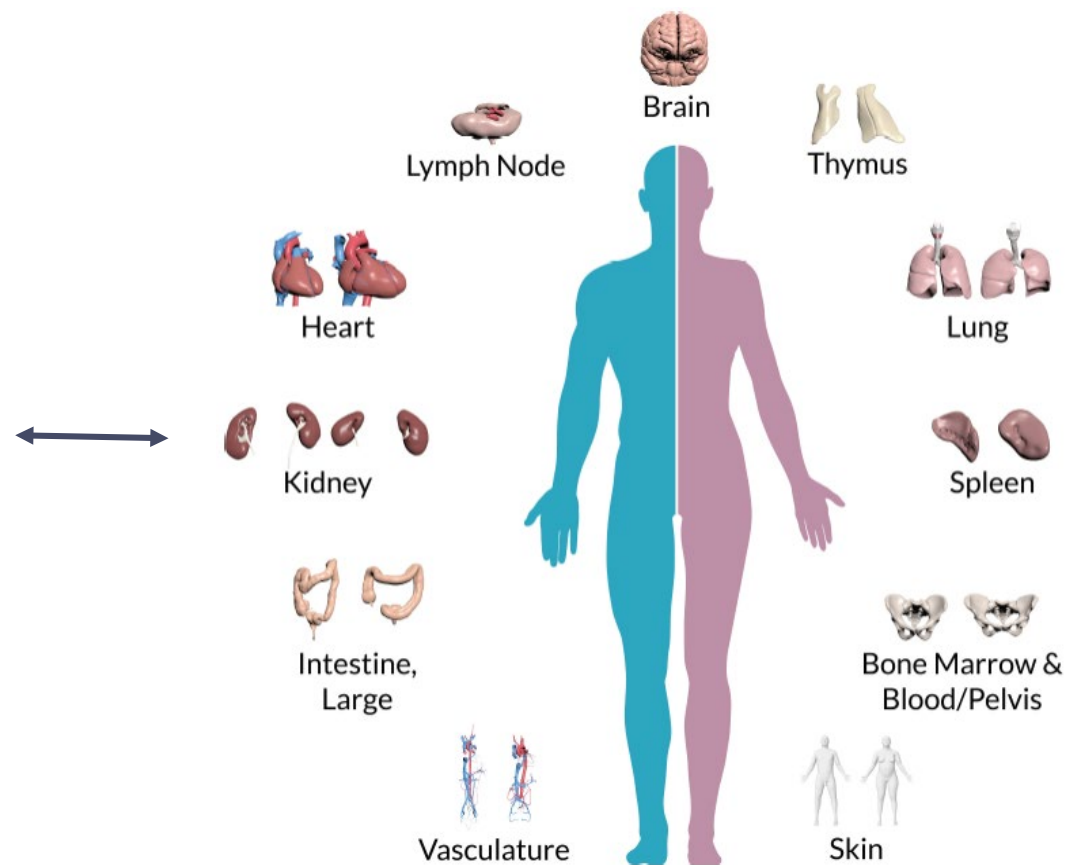
Bimodal network describing which B characterize what CT

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

An Atlas describes & names 2/3D entities

Organ	#AS	#CT	#B Total	#BG	#BP	#AS-AS	#AS-CT	#CT-B
Bone Marrow & Blood/Pelvis	3	46	327	201	126	2	70	710
Brain	187	127	254	254	0	187	127	330
Heart	52	25	48	48	0	61	164	78
Intestine, Large	65	69	94	88	6	389	1,361	197
Kidney	68	63	152	152	0	67	59	257
Lung	161	92	176	172	4	1,633	12,094	286
Lymph Node	41	49	266	108	158	62	135	544
Skin	16	42	70	0	70	17	19	105
Spleen	46	66	255	80	145	68	172	414
Thymus	25	41	511	388	123	38	180	657
Vasculature	870	2	1	1	0	869	606	2
Totals:	1,534	622	2,154	1,492	632	3,393	14,987	3,580



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

<https://hubmapconsortium.github.io/ccf/pages/ccf-3d-reference-library.html>

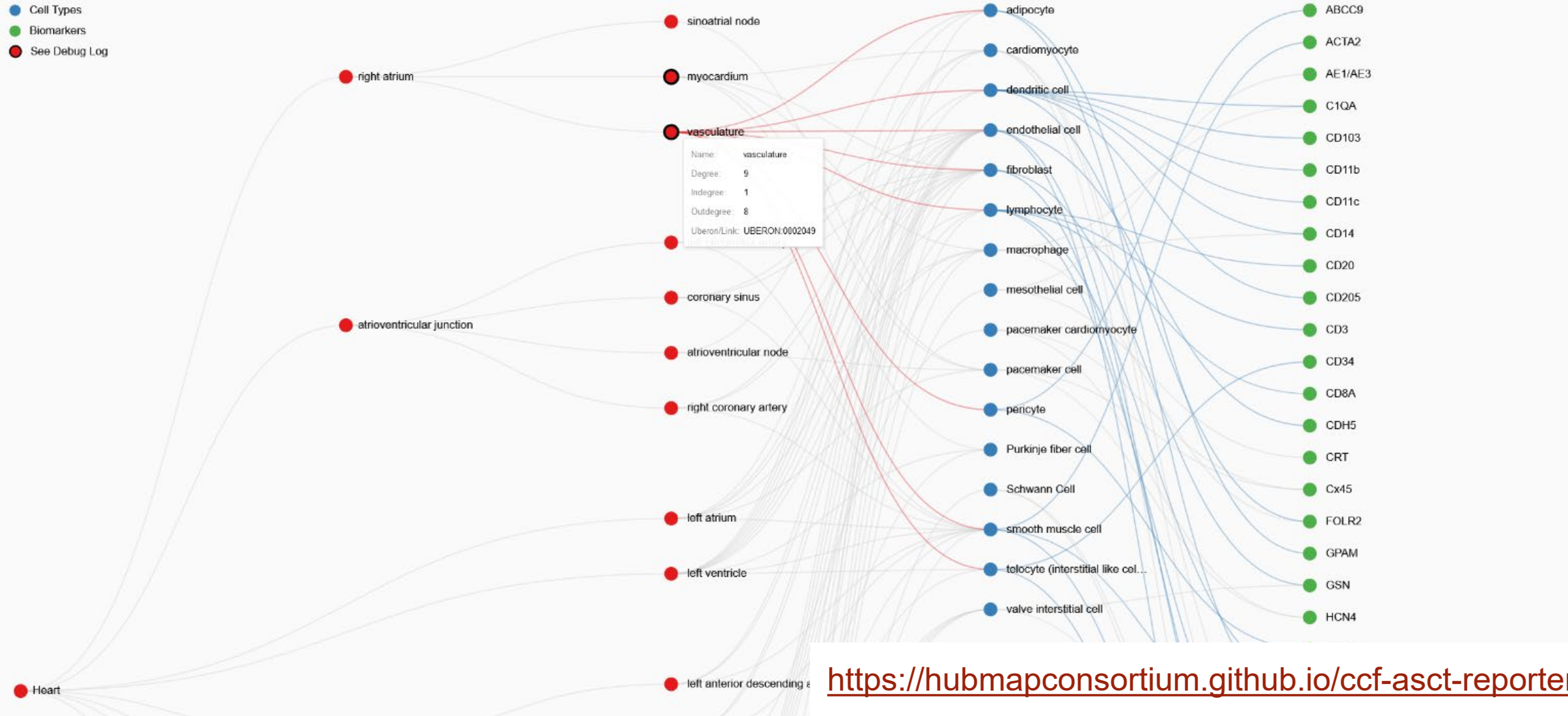
Anatomical Structures

Cell Types

Biomarkers

Legend

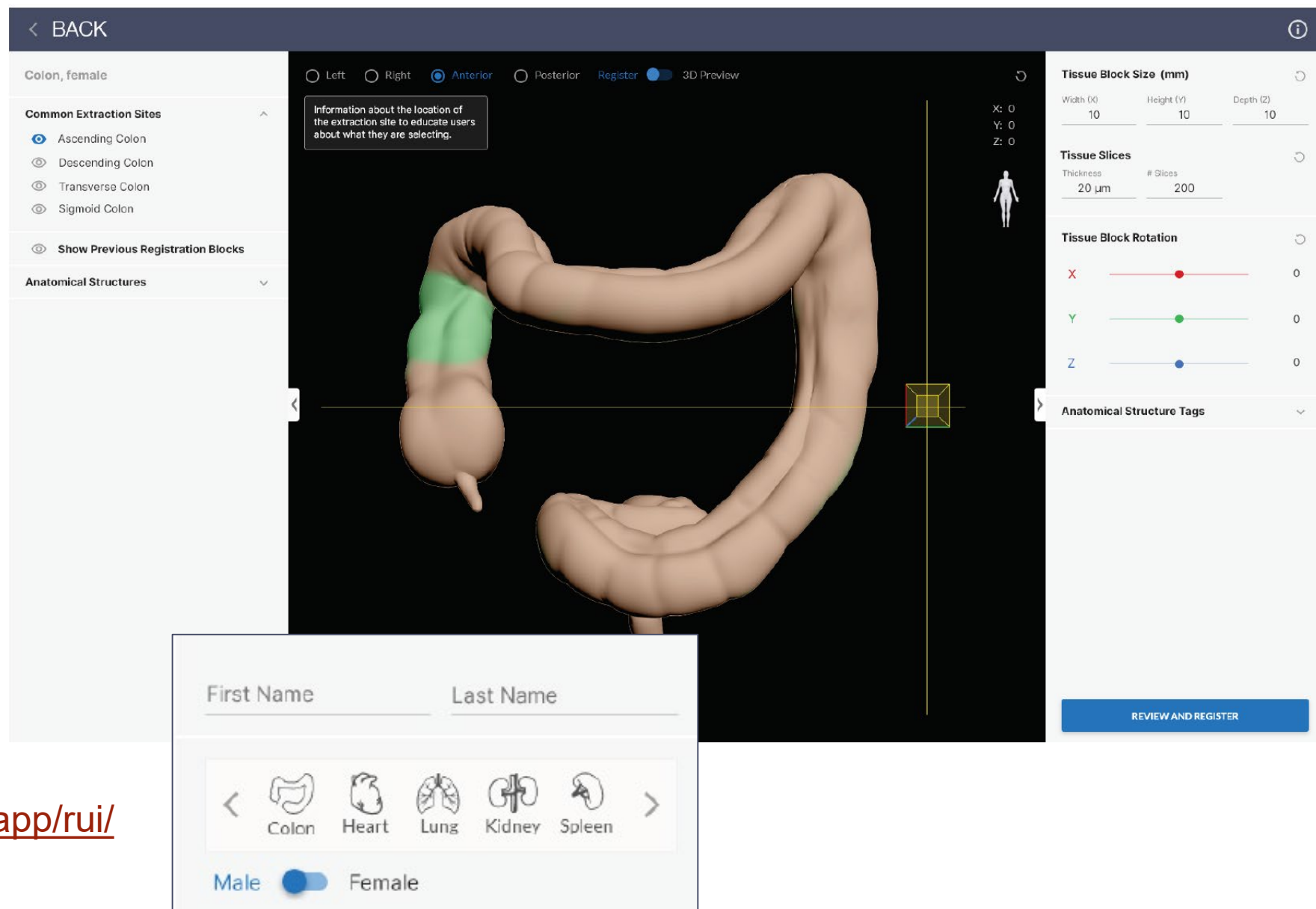
- Anatomical Structures
- Cell Types
- Biomarkers
- See Debug Log



CCF Registration User Interface (RUI)

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/>

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

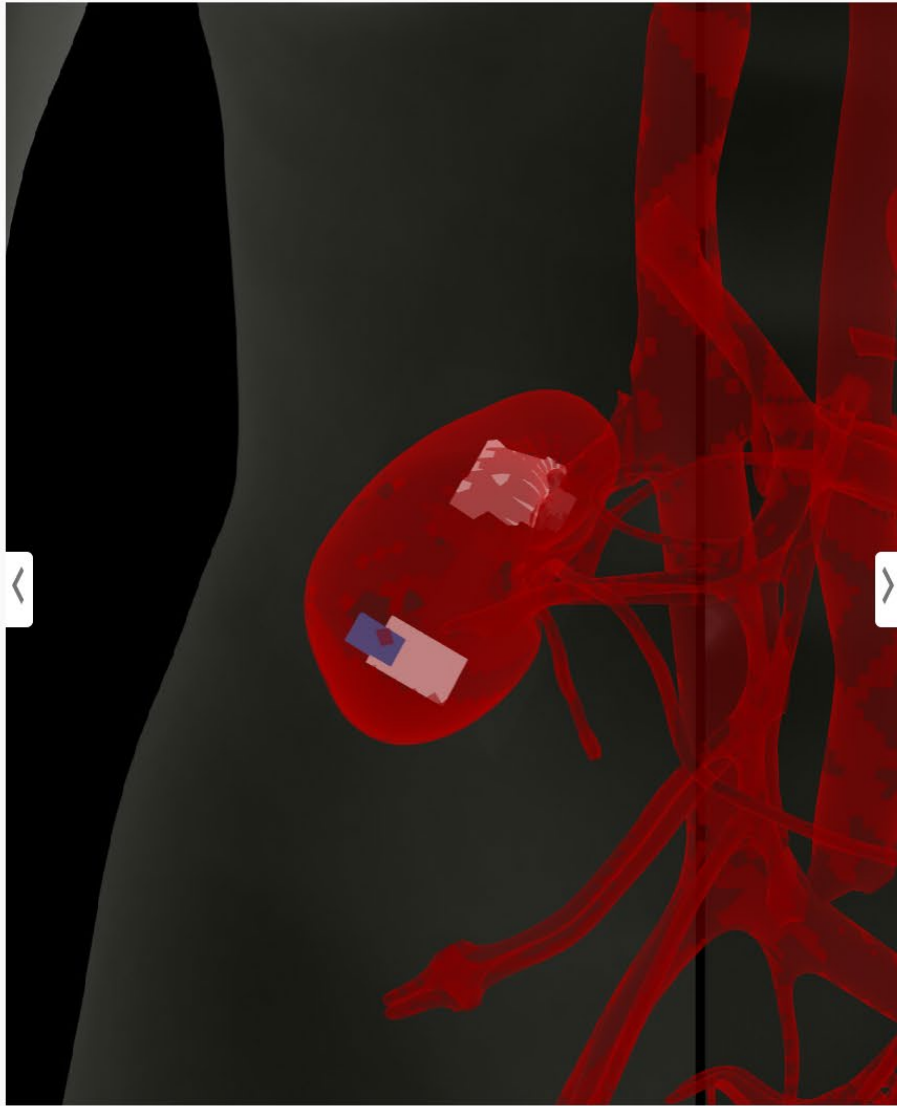
	Female, Age 14, BMI 14.7 HBM894.MPVN.828 TMC-Florida First case collected. Incomplete d...	
	Male, Age 18, BMI 27.1 HBM436.GHWX.449 TMC-Florida section is 190um from block surface	
	Male, Age 56, BMI 32.5 HBM696.XTVL.498 TMC-Vanderbilt Age 56, White Male	
	Male, Age 53, BMI 26.5 HBM652.VRLD.292 TMC-Vanderbilt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



body

- 1 Centers
- 9 Donors
- 40 Samples

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

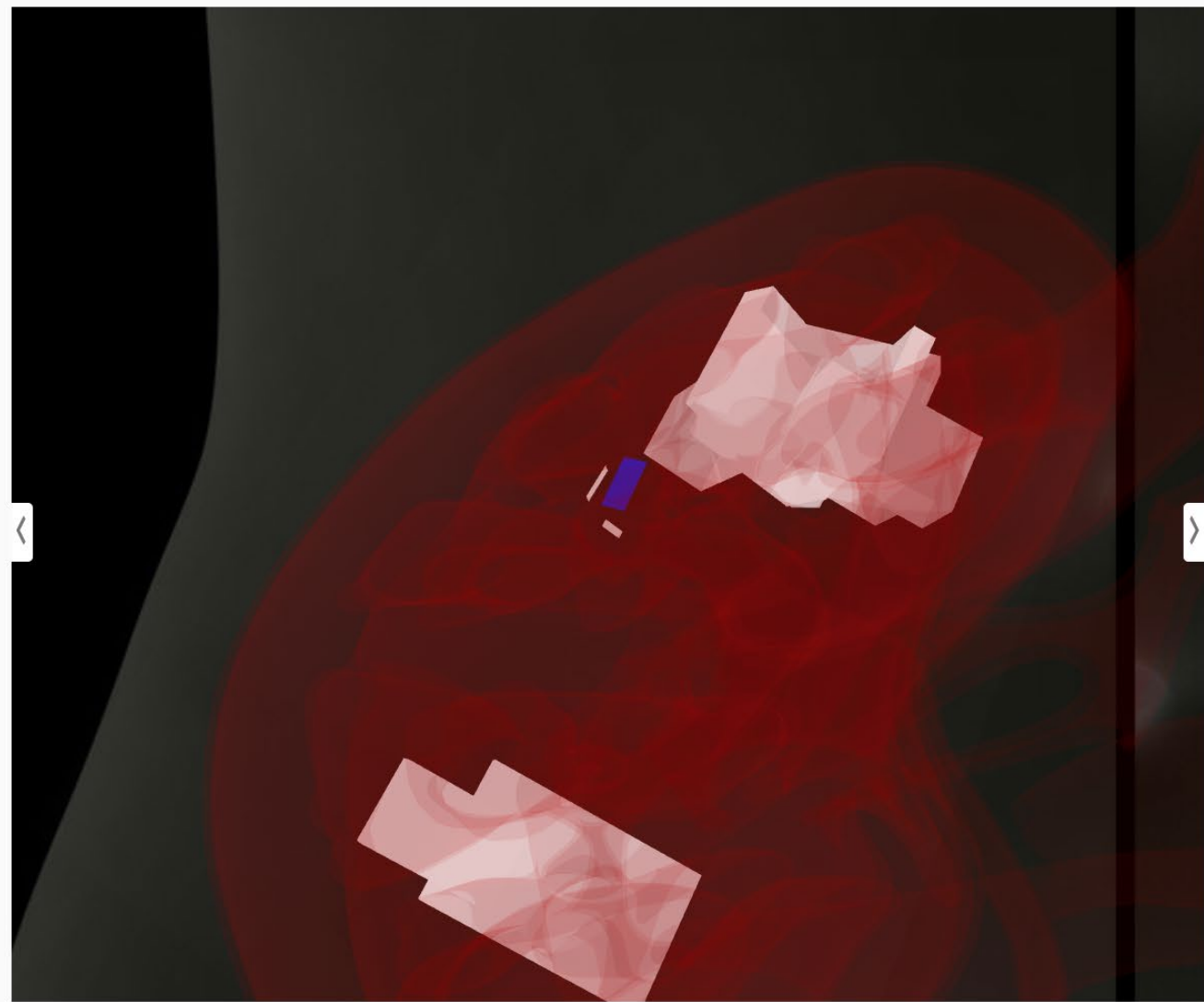


Search ontology terms ...

- 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















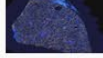







Search ontology terms ... 

- ▼ 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

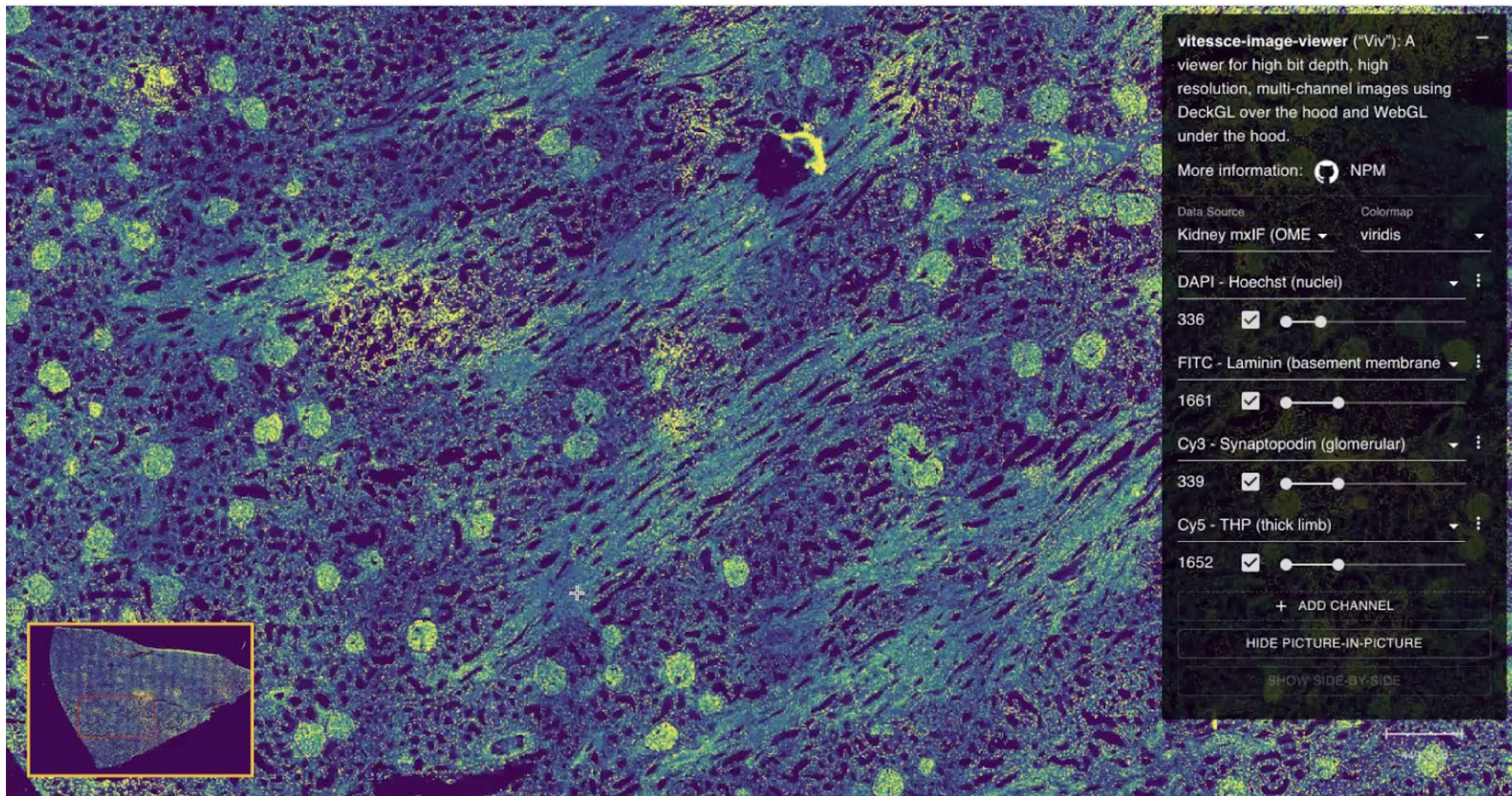


body

- 2 Centers
- 9 Donors
- 14 Samples

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

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.

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>

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



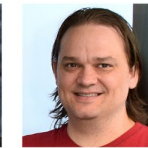
Katy Börner
MC-IU PI
CNS Director



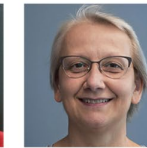
Griffin Weber
Assoc. Professor of Medicine
Harvard Medical School



Lisel Record
MC-IU PM
CNS Associate Director



Bruce Herr II
Sr. Systems Architect/PM



Ellen Quardokus
Sr. Research Analyst



Yingnan Ju
PhD Candidate



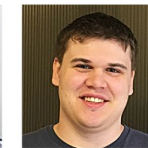
Andreas Bueckle
PhD Candidate



Leonard Cross
Sr. UX/UI Designer



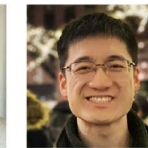
Matthew Martindale
Center Assistant



Daniel Bolin
Software Developer



Adam Phillips
Software Developer



Edward Lu
Software Developer



Paul Hrishikesh
Research Assistant



Leah Scherschel
Research Assistant



Avinash Boppana
Research Consultant

Upcoming Events & Books



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

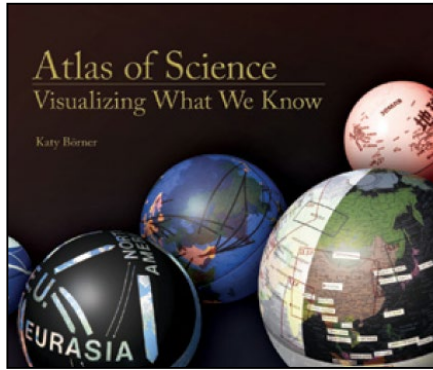
Final version due: 25 May 2022

Publication: July/August 2022

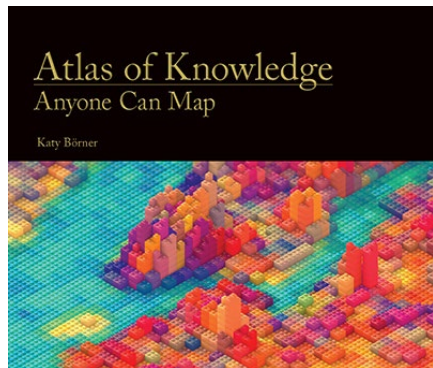


Indiana University Bloomington will host the
**International Society of Scientometrics & Informetrics
Conference (*ISSI*)**
in Summer 2023

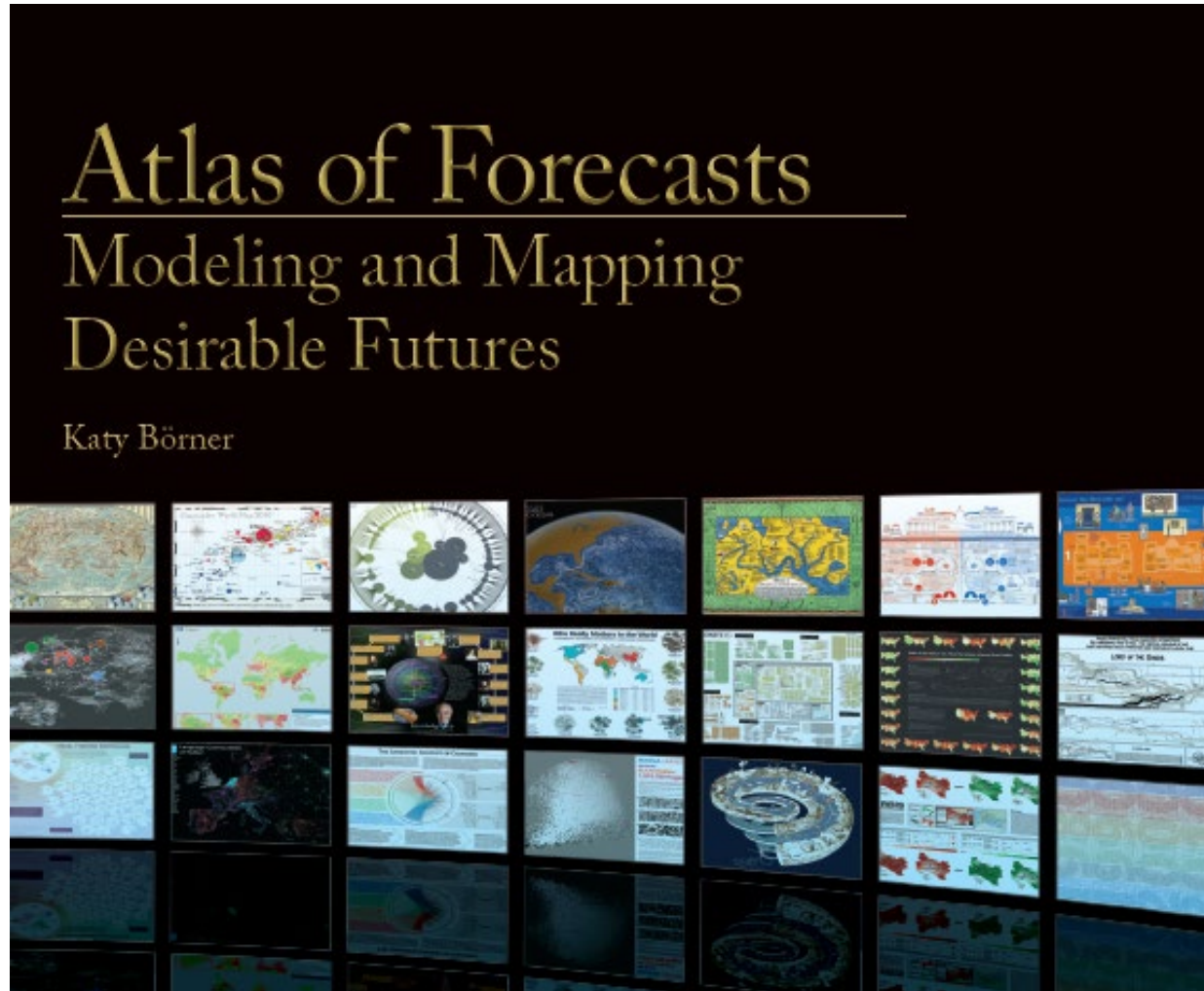
Atlas Trilogy



2010

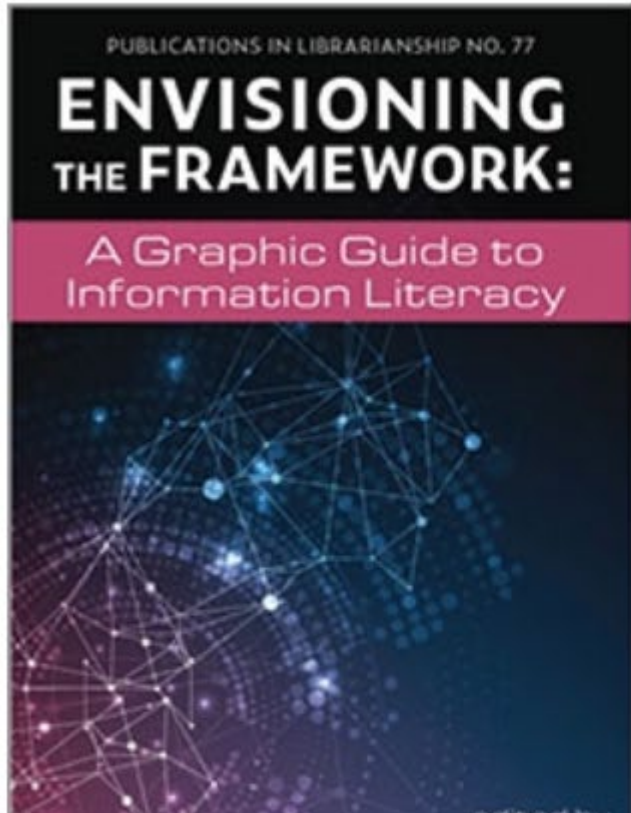


2015



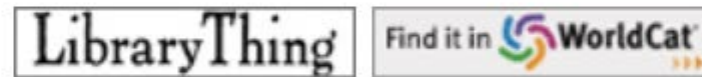
2021

<https://mitpress.mit.edu/books/atlas-forecasts>



Envisioning the Framework: A Graphic Guide to Information Literacy

Jannette L. Finch



Jannette L. Finch

Jannette L. Finch, MLIS, is a librarian in the College of Charleston Libraries system. Her research interests include information design and the effect of technology on student learning, online learning and teaching, effective teaching through experiential learning activities, constructivist techniques in the teaching and learning environment, visualizing data, library service models, the library role in the scholarly community, assessment, and planning. Contact her at finchj@cofc.edu.

Q&A

