

# Data Visualization Literacy

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*Data Visualization Session at the European Meeting on Applied Evolutionary Economics (EMAE)*

Brighton, UK | June 4, 2019



# Overview

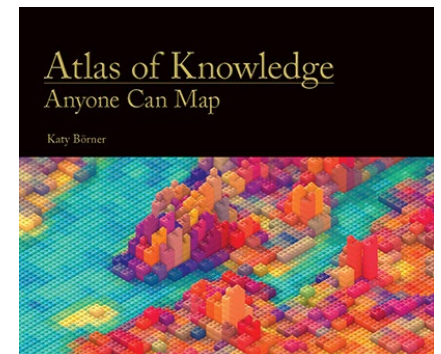
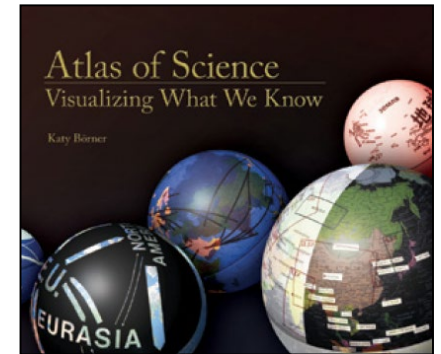
This talk will introduce a theoretical data visualization framework (DVL) meant to empower anyone to systematically render data into insights using temporal, geospatial, topical, and network analyses and visualizations.

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

Exemplarily, the DVL is applied to map science and technology, see interactive data visualizations from the *Places & Spaces: Mapping Science* exhibit (<http://scimaps.org>) and recent *PNAS* special issue on *Modelling and Visualizing Science and Technology Developments* (<https://www.pnas.org/modeling>)

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

Börner, Katy. 2010. [Atlas of Science: Visualizing What We Know](#). Cambridge, MA: The MIT Press.



Atlas of Forecasts

# Places & Spaces: Mapping Science Exhibit

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

## Maps

<p><b>Iteration I (2005)</b> The Power of Maps</p>	<p><b>Iteration II (2006)</b> The Power of Reference Systems</p>
<p><b>Iteration III (2007)</b> The Power of Forecasts</p>	<p><b>Iteration IV (2008)</b> Science Maps for Economic Decision Makers</p>
<p><b>Iteration V (2009)</b> Science Maps for Science Policy Makers</p>	<p><b>Iteration VI (2010)</b> Science Maps for Scholars</p>
<p><b>Iteration VII (2011)</b> Science Maps as Visual Interfaces to Digital Libraries</p>	<p><b>Iteration VIII (2012)</b> Science Maps for Kids</p>
<p><b>Iteration IX (2013)</b> Science Maps Showing Trends and Dynamics</p>	<p><b>Iteration X (2014)</b> The Future of Science Mapping</p>

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

## Macroscopes

<p><b>Iteration XI (2015)</b> Macroscopes for Interacting with Science</p>
<p><b>Iteration XIII (2017)</b> Macroscopes for Playing with Scale</p>
<p><b>Iteration XII (2016)</b> Macroscopes for Making Sense of Science</p>
<p><b>Iteration XIV (2018)</b> Macroscopes for Ensuring our Well-being</p>

3<sup>rd</sup> Decade (2015-2034)

<p>101st Annual Meeting of the Association of American Geographers, Denver, CO. April 5th - 9th, 2005 (First showing of Places &amp; Spaces)</p>	<p>University of Miami, Miami, FL. September 4 - December 11, 2014.</p>
<p>Duke University, Durham, NC. January 12 - April 10, 2015</p>	<p>The David J. Sencer CDC Museum, Atlanta, GA. January 25 - June 17, 2016.</p>

<http://scimaps.org>

# 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:

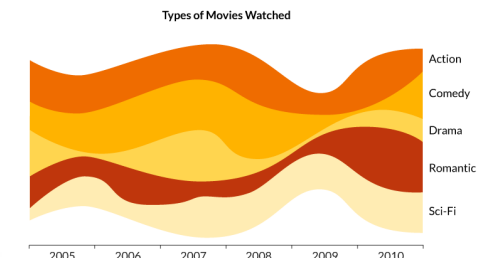
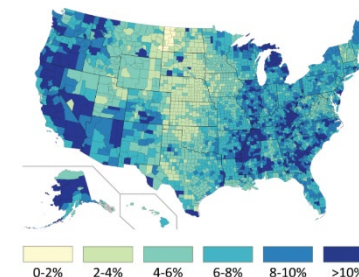
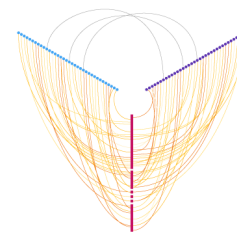
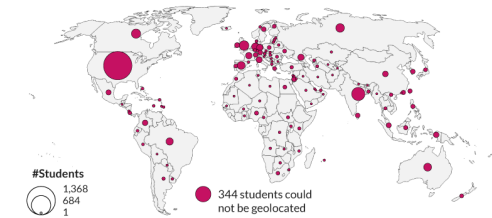
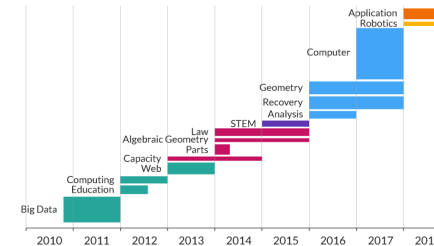
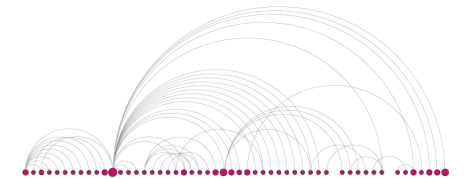
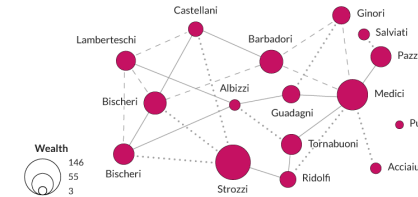
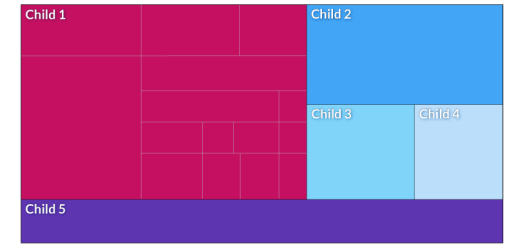
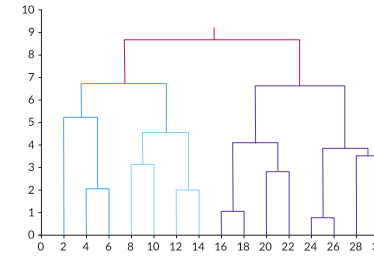
- literacy (ability to read and write text in titles, axis labels, legends, etc.),
- visual literacy (ability to find, interpret, evaluate, use, and create images and visual media), and
- mathematical literacy (ability to formulate, employ, and interpret math in a variety of contexts).

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.

# Visualization Frameworks

MANY frameworks and taxonomies have been proposed to

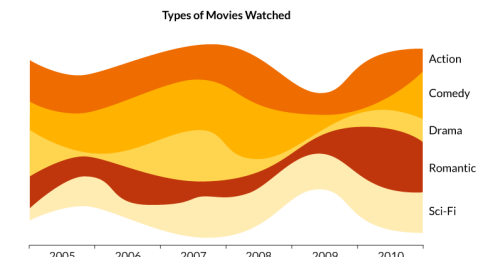
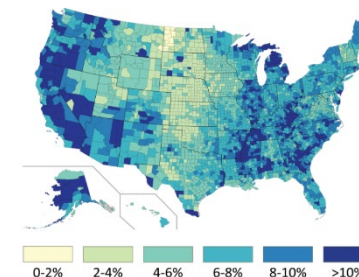
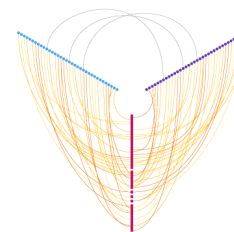
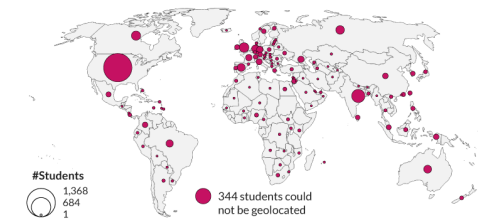
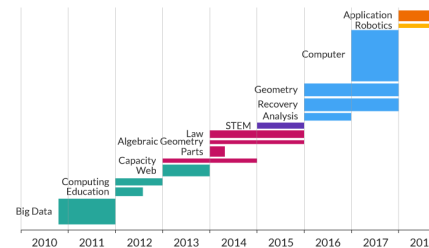
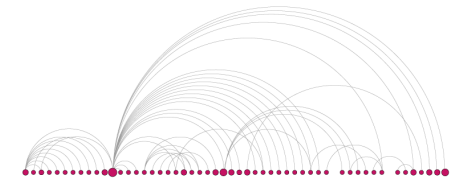
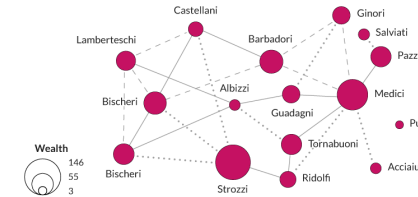
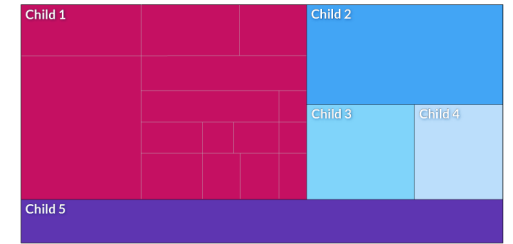
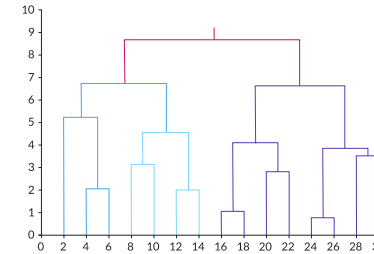
- help organize and manage the evolving zoo of 500+ different data visualization types,
- provide guidance when designing data visualizations, and
- facilitate teaching.



# Existing Visualization Frameworks

Organize data visualizations by

- User insight needs
- User task types
- Data to be visualized
- Data transformations
- Visualization technique
- Visual mapping transformations
- Interaction techniques
- Deployment options
- and other features ...



# 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

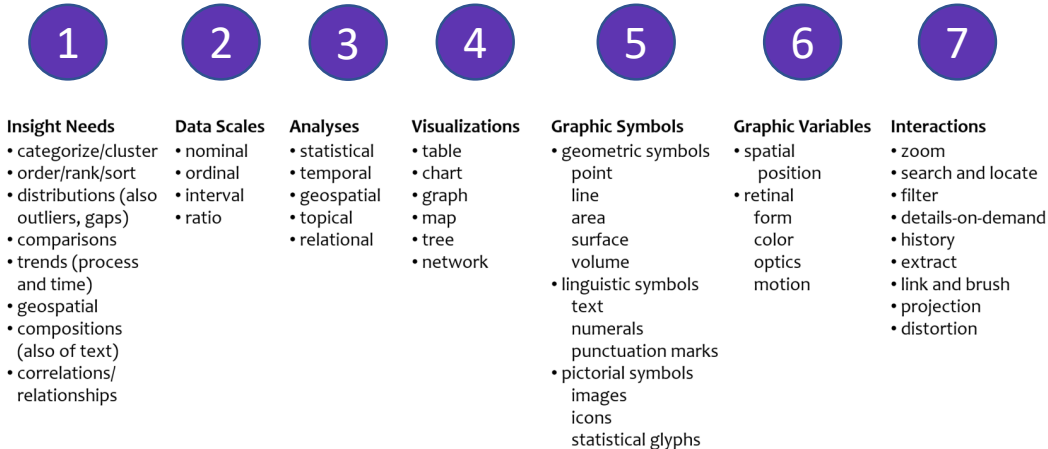
- The initial DVL-FW was developed via an extensive literature review.
- The resulting DVL-FW typology, process model, exercises, and assessments were then tested in the *Information Visualization* course taught for more than 15 years at Indiana University. More than 8,500 students enrolled in the IVMOOC 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:

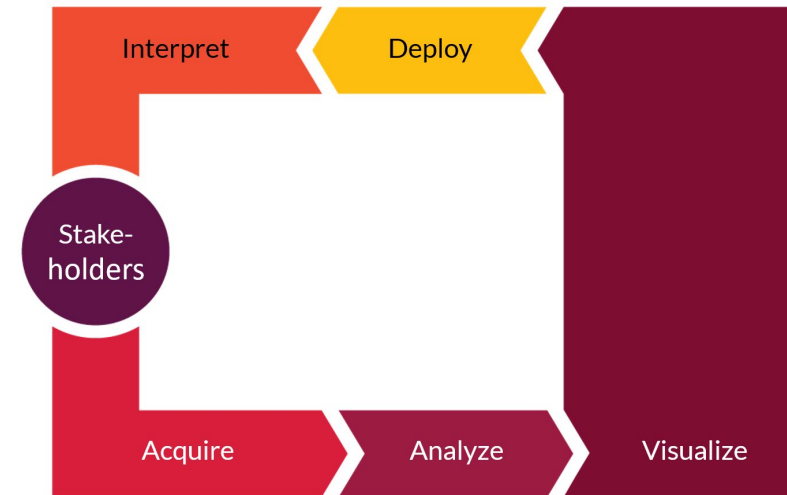
## DVL Typology

Defines 7 types with 4-17 members each.



## DVL Workflow Process

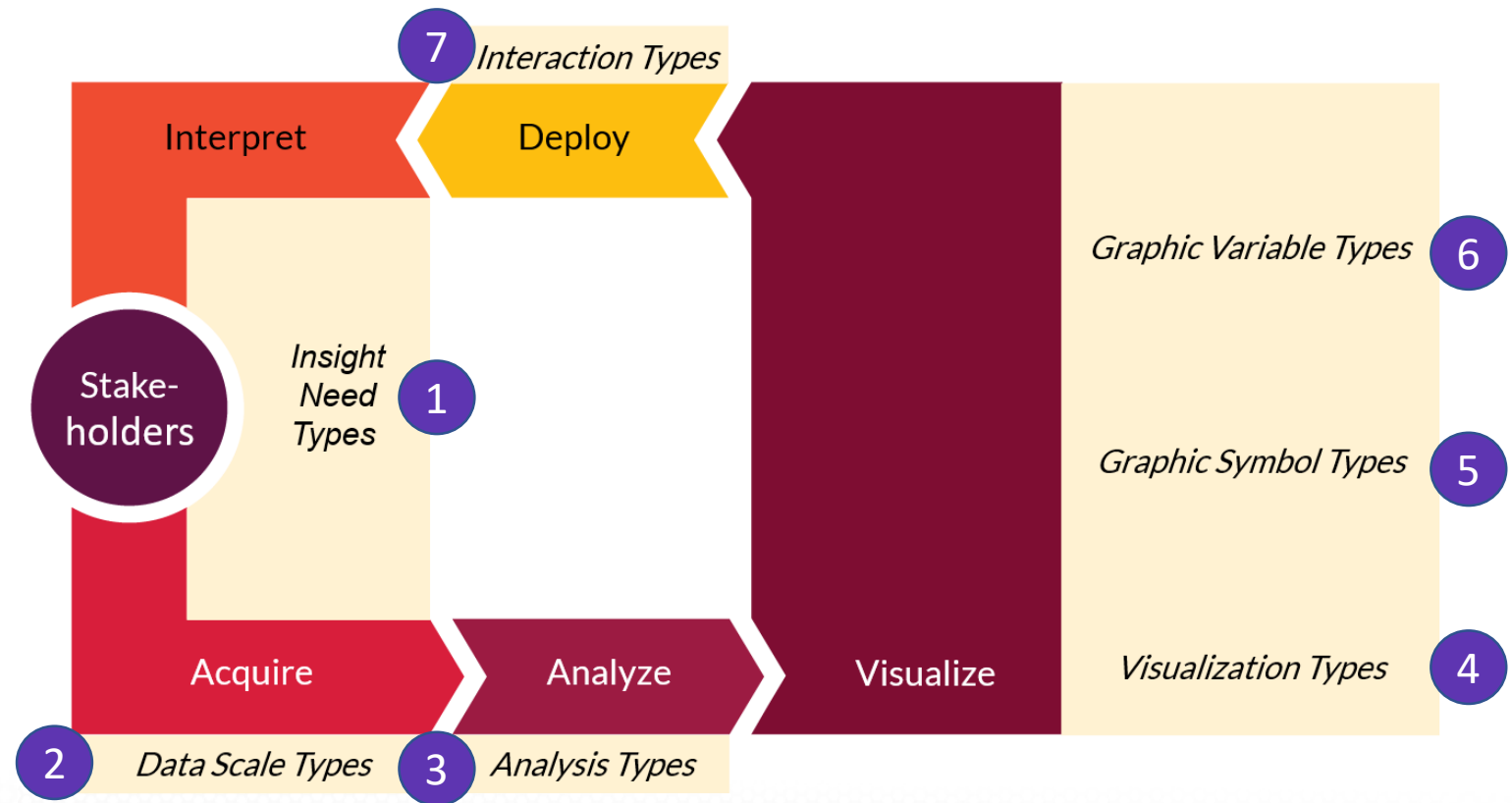
Defines 5 steps required to render data into insights.



# 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.

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:**

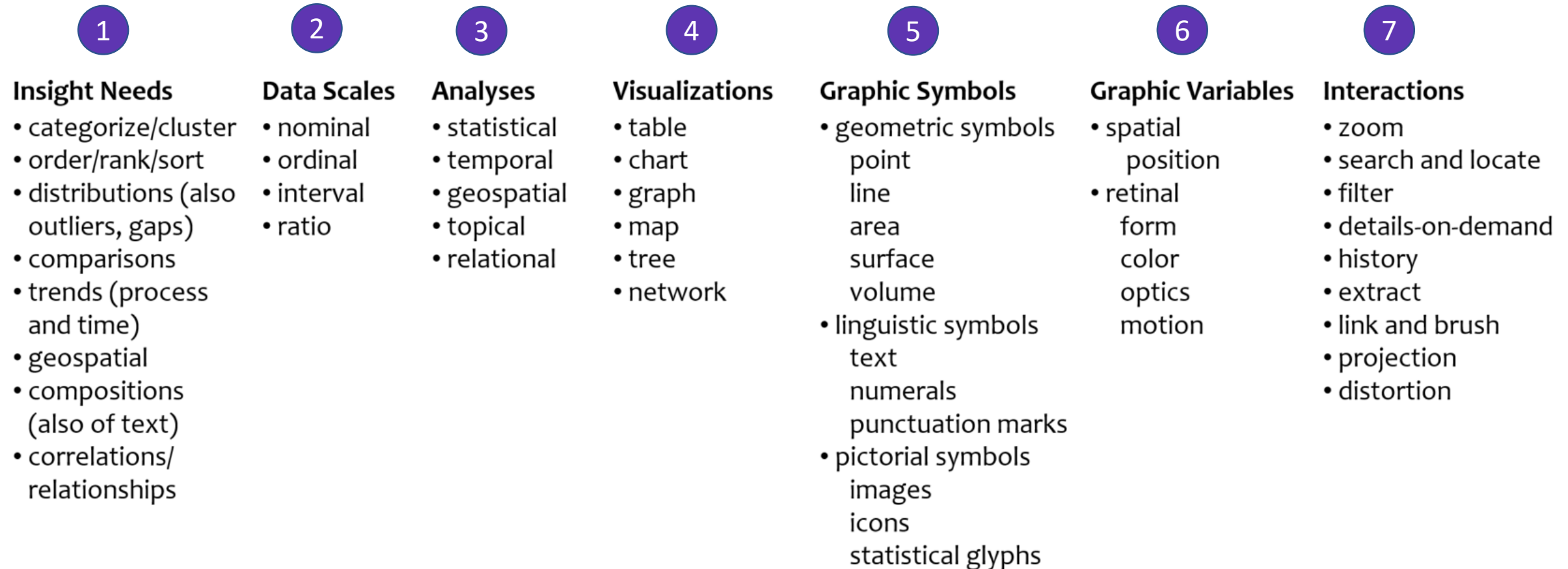
- Select Visualization Type:** Scatter Graph, Geomap, Scimap, Temporal Bar Graph (selected).
- Select Graphic Symbol Type(s):** (Dropdown menu)
- Select Graphic Variable Types:** (Dropdown menu)

**Temporal Bar Graph Preview:**

Temporal Bar Graph

Year	Machine	Big Data	Education	Building	Making	Computing	Web	Form	Smart	Capacity	Algebraic Geometry	Parts	Law	Stem	Analysis	Recovery	Geometry	Computer	Application	Robotics	
1998	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2000	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2002	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2004	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2006	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2008	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2010	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2012	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2014	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2016	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low
2017	High	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low	Low

# 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

2

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# Data Scale Types

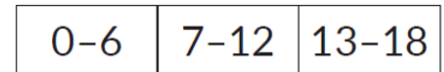
**Nominal:** A categorical scale, also called a nominal or category scale, is **qualitative**. Categories are assumed to be non-overlapping.



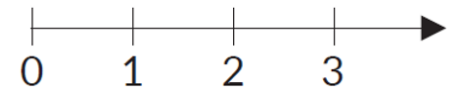
**Ordinal:** An ordinal scale, also called sequence or ordered, is **quantitative**. It rank-orders values representing categories based on some intrinsic ranking, but not at measurable intervals.



**Interval:** An interval scale, also called a value scale, is a **quantitative** numerical scale of measurement where the distance between any two adjacent values (or intervals) is equal, but the zero point is arbitrary.



**Ratio:** A ratio scale, also called a proportional scale, is a **quantitative** numerical scale. It represents values organized as an ordered sequence, with meaningful uniform spacing, and a true zero point.



# Data Scale Types - Examples

**Nominal:** Words or numbers constituting the “categorical” names and descriptions of people, places, things, or events.

**Ordinal:** Days of the week, degree of satisfaction and preference rating scores (e.g., using a Likert scale), or rankings such as low, medium, high.

**Interval:** Temperature in degrees or time in hours. Spatial variables such as latitude and longitude are interval.

**Ratio:** Physical measures such as height, weight, (reaction) time, or intensity of light; number of published papers, co-authors, citations.

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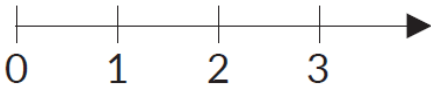
Quantitative

**Interval:** Temperature in degrees or time in hours. Spatial variables such as latitude and longitude are interval.

**Ratio:** Physical measures such as weight, height, (reaction) time, or intensity of light; number of published papers, co-authors, citations.

# Data Scale Types - Mathematical Operations

This table shows the logical mathematical operations permissible, the measure of central tendency, and examples for the different data scale types.

Data Scale Types	Logical Mathematical Operations				Measure of Central Tendency	Examples			
	= ≠	< >	+ -	x ÷					
Nominal	y				mode	  			
Ordinal	y	y			median	  			
Interval	y	y	y		arithmetic mean	<table border="1" data-bbox="1396 1043 1819 1115"> <tr> <td>0-6</td> <td>7-12</td> <td>13-18</td> </tr> </table>	0-6	7-12	13-18
0-6	7-12	13-18							
Ratio	y	y	y	y	geometric mean				

Qualitative

Quantitative

# Typology of the Data Visualization Literacy Framework

3

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  - text
  - numerals
  - punctuation marks
- pictorial symbols
  - images
  - icons
  - statistical glyphs

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

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

# Analysis Types

- When: Temporal Data Analysis + Statistical
- Where: Geospatial Data Analysis
- What: Topical Data Analysis
- With Whom: Network Analysis

# Typology of the Data Visualization Literacy Framework

4

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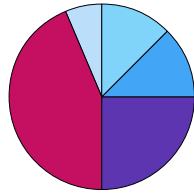
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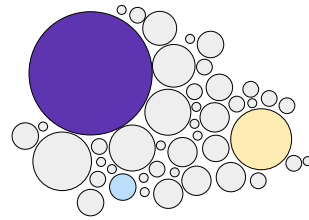
Börner, Katy. 2015. *Atlas of Knowledge: Anyone Can Map*. Cambridge, MA: The MIT Press. 30-31.

# Visualization Types

Chart

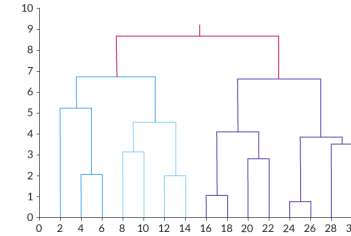


Pie Chart

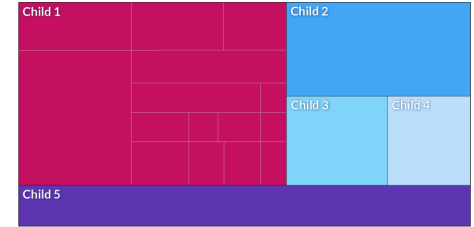


Bubble Chart

Tree

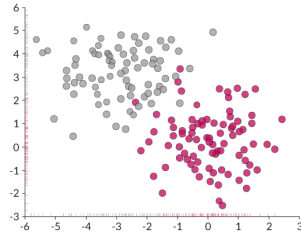


Dendrogram

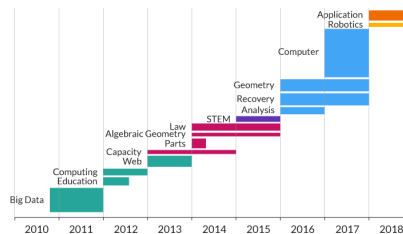


Tree Map

Graph

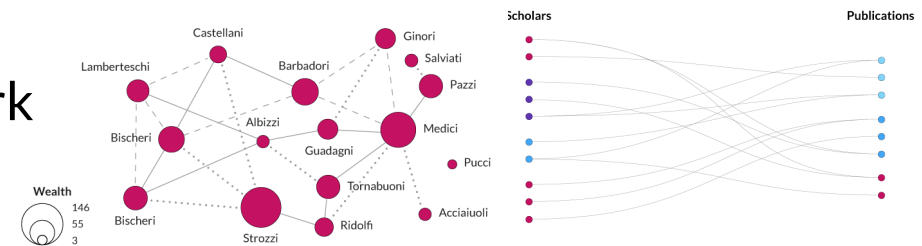


Scatter Graph



Temporal Bar Graph

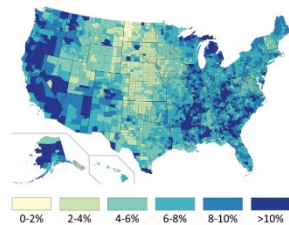
Network



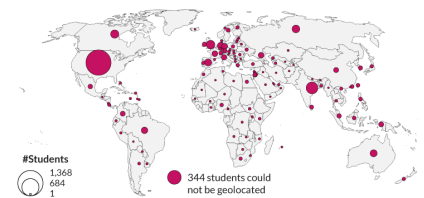
Dendrogram

Tree Map

Map



Choropleth Map

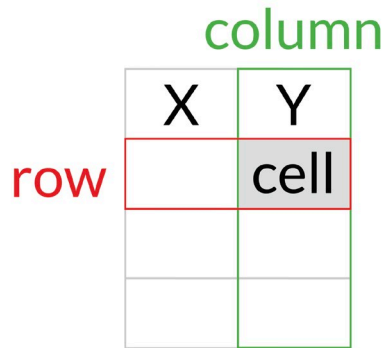


Proportional Symbol Map

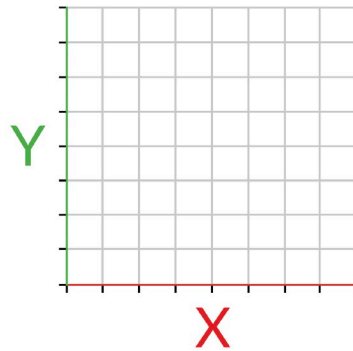


# Visualize: Reference Systems

**Table**  
columns by rows



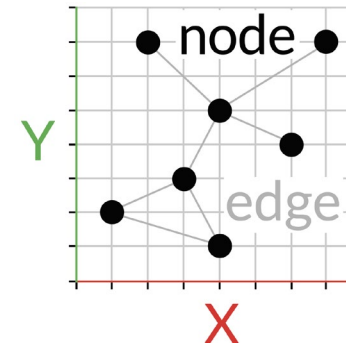
**Graph**  
x-y coordinates



**Map**  
latitude/  
longitude



**Network**  
local similarity

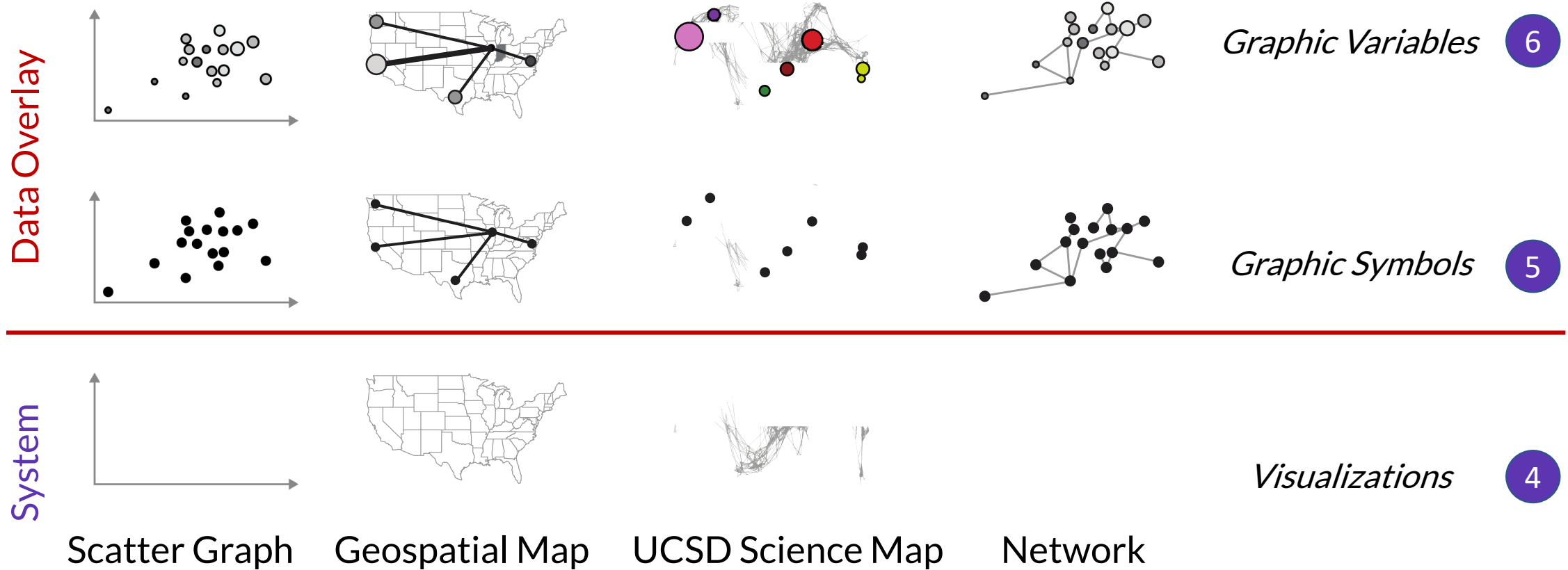


4

**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)
- 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. 32-33.

# Typology of the Data Visualization Literacy Framework

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

# Graphic Variable Types

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

Quantitative

**Form:**

- Size
- Shape
- Rotation (Orientation)

Quantitative

Qualitative

Quantitative

**Color:**

- Value (Lightness)



Quantitative

- Hue (Tint)



Qualitative

- Saturation (Intensity)



Quantitative

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

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

**Motion:** Speed, Velocity, Rhythm

# Graphic Variable Types

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

**Form:**

- Size
- Shape
- Rotation (Orientation)

**Color:**

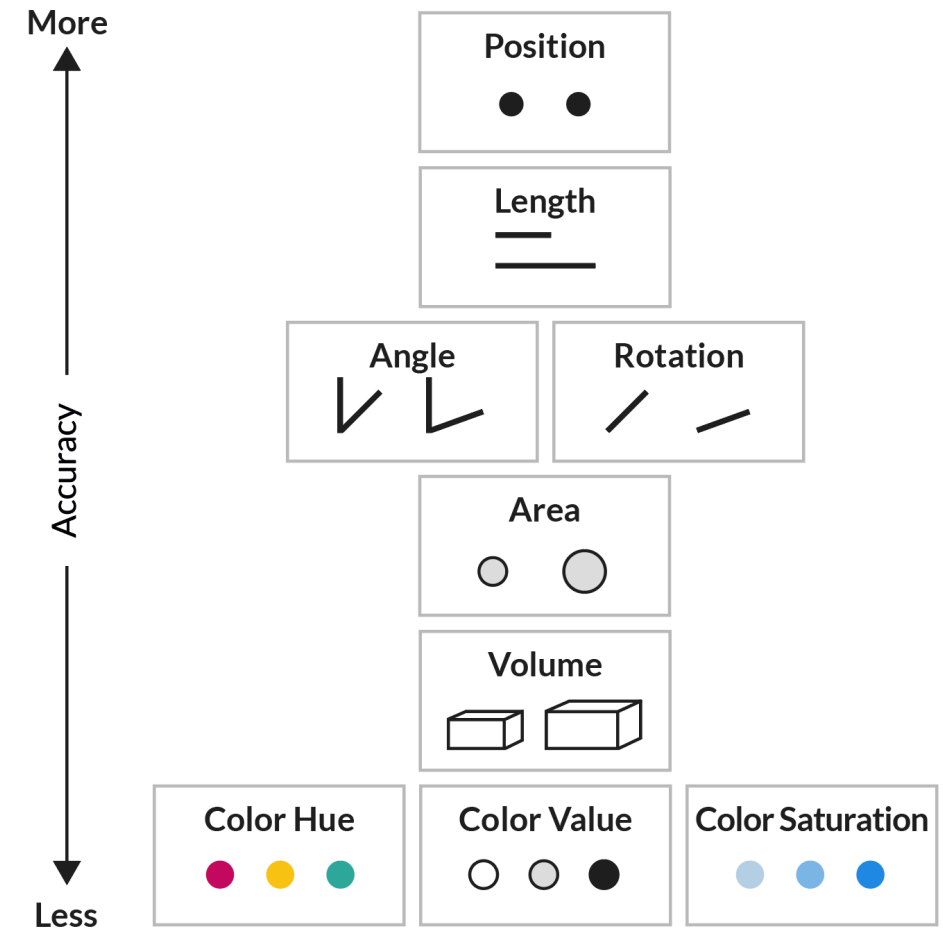
- Value (Lightness)
- Hue (Tint)
- Saturation (Intensity)



**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.



# Typology of the Data Visualization Literacy Framework

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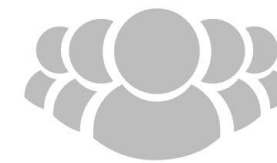
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Connect with industry professionals and leading researchers.



## Evolve Yourself

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# Interdisciplinary Training in Complex Networks and Systems

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Understanding complex networked systems is key to solving some of the most vexing problems confronting humankind, from discovering how dynamic brain connections give rise to thoughts and behaviors, to detecting and preventing the spread of misinformation or unhealthy behaviors across a population. Graduate training, however, typically occurs in one of two dimensions: experimental and observational methods in a specific area such as biology and sociology, or in general methodologies such as machine learning and data science.



## CNS NRT

<https://cns-nrt.indiana.edu>

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