

Computer-Based Visualization

Data Mining and Visualization

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Web site of the course (Remember)

- <http://people.irisa.fr/Alexandre.Termier/dmv/>
- Web site contains:
 - General information
 - Up-to-date schedule (will be the reference)
 - Links to documents

Definition

Computer-based visualization systems provide **visual representations of datasets** designed to **help people carry out tasks more effectively**

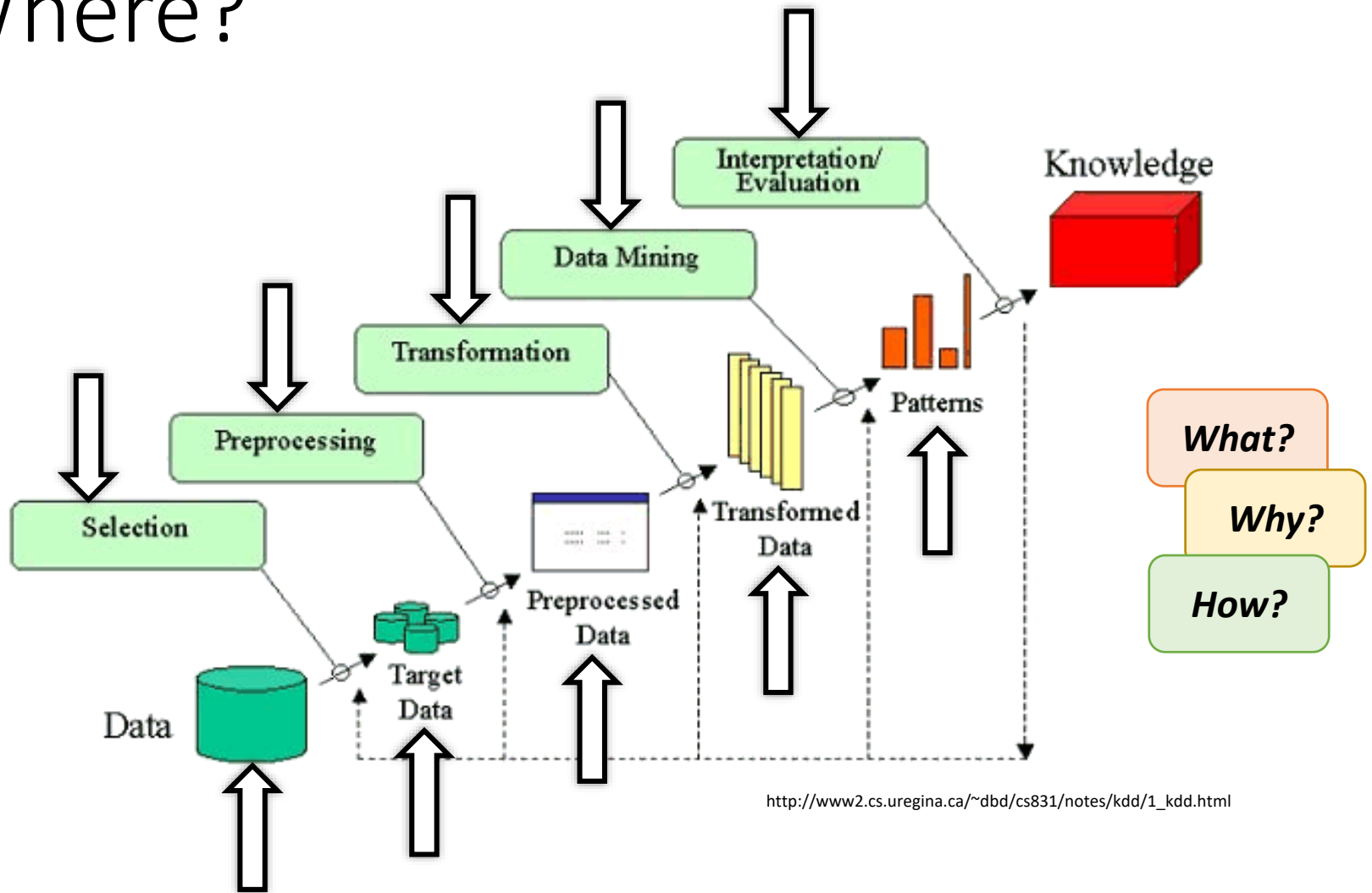


Context

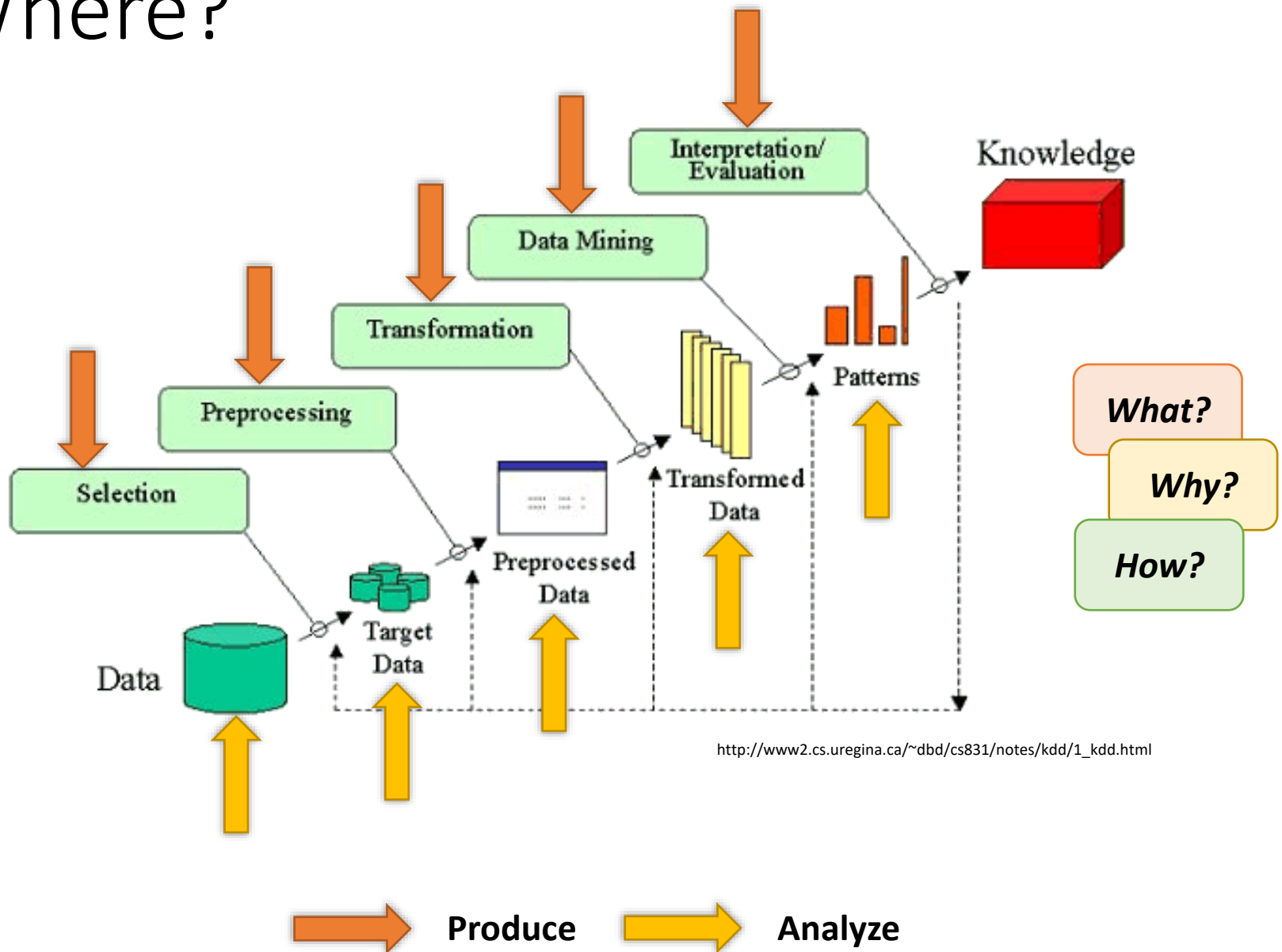
Visualization is suitable when there is a **need to augment human capabilities** rather than replace people with computational decision-making methods

- Need to present results to users -> data visualization
 - Huge lack of communication between data mining / data vis community
 - This course: a small step to improve this communication

Where?



Where?



Simple Example

1.1.1:MY,NP;MY,MB

1.1.2:MY,ME;ME,MB

1.1.3:MY

1.1.4:MY,ME;MY,CL;MY,GE;MY,MC;MY,MB

1.1.5:MY,MB,ME

1.1.6:ME,MY

1.1.7:MY,CV;MY,MB,ME

1.1.8:SN,MY

1.1.9:MB

1.1.10:MY,GG

1.1.11:MY

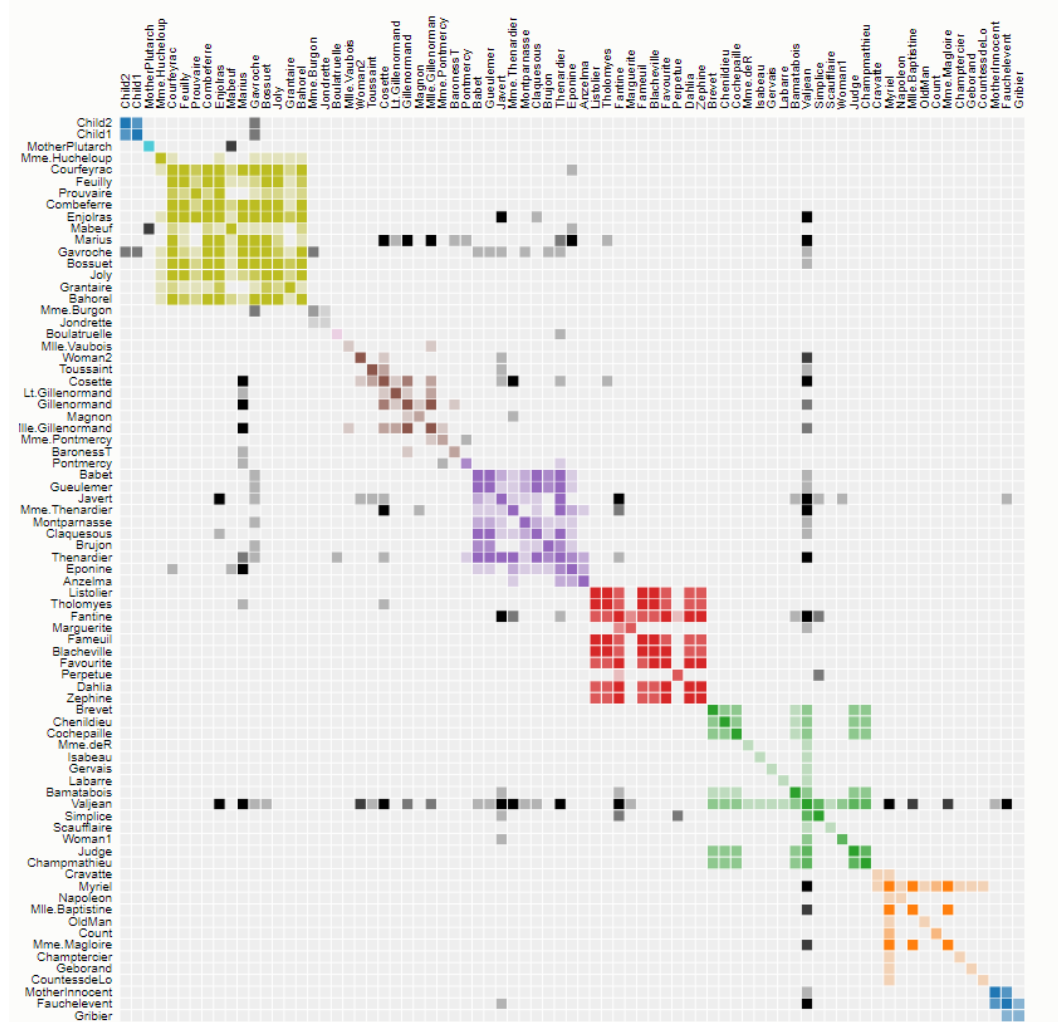
1.1.12:MY

1.1.13:MY

1.1.14:MY,SN

1.2.1:JL,JV;JV,MT;MR,JV

...

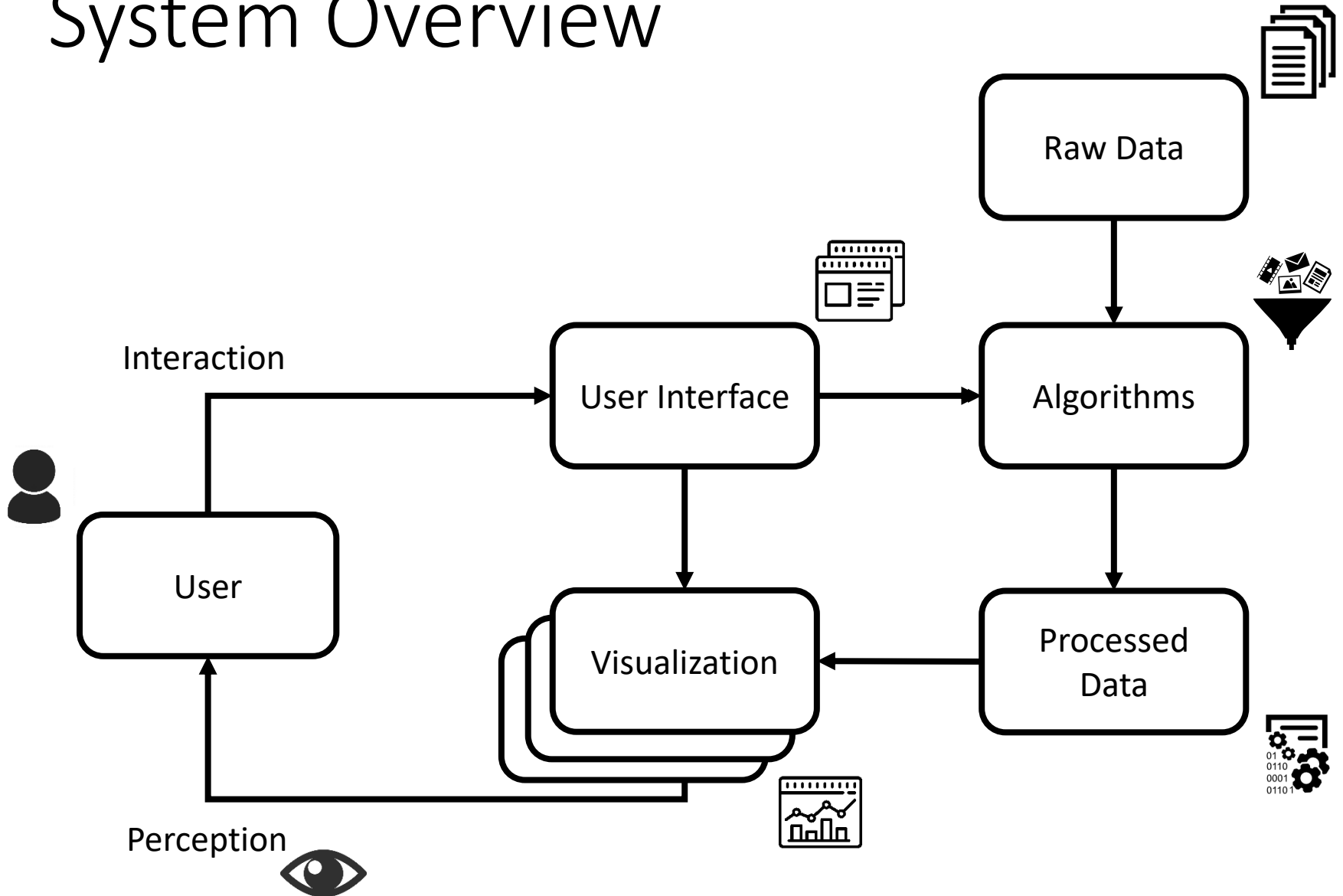


<https://bost.ocks.org/mike/miserables/>

Context

- Scientific visualization is mostly concerned with:
 - 2, 3, 4 dimensional, spatial or spatio-temporal data
 - Discretized data
- Information visualization focuses on:
 - High-dimensional, abstract data
 - Discrete data
 - Financial, statistical, etc
 - Visualization of large trees, networks, graphs
 - Data mining: finding patterns, clusters, voids, outliers

System Overview



Content

- Introduction
- Data abstraction
- Task abstraction
- Visualization Design
- Validation

Introduction

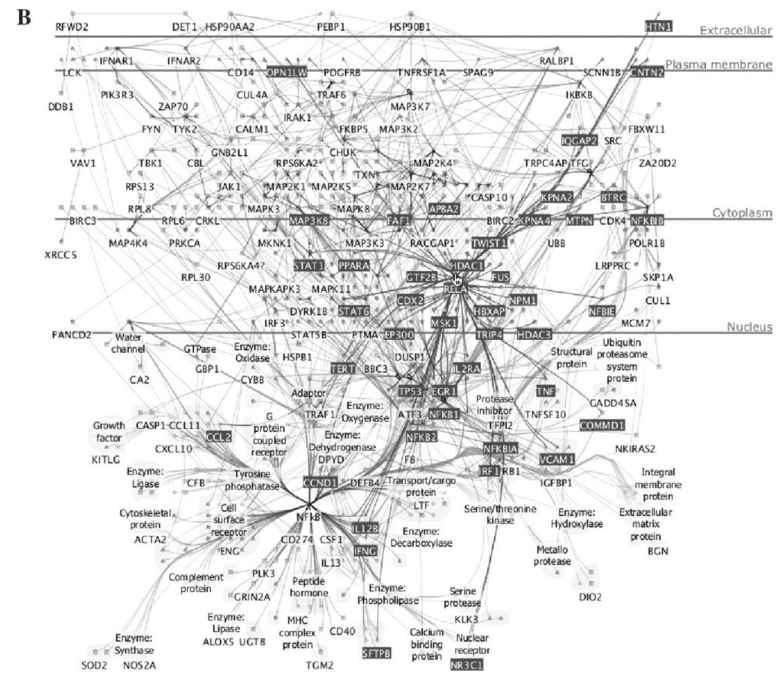
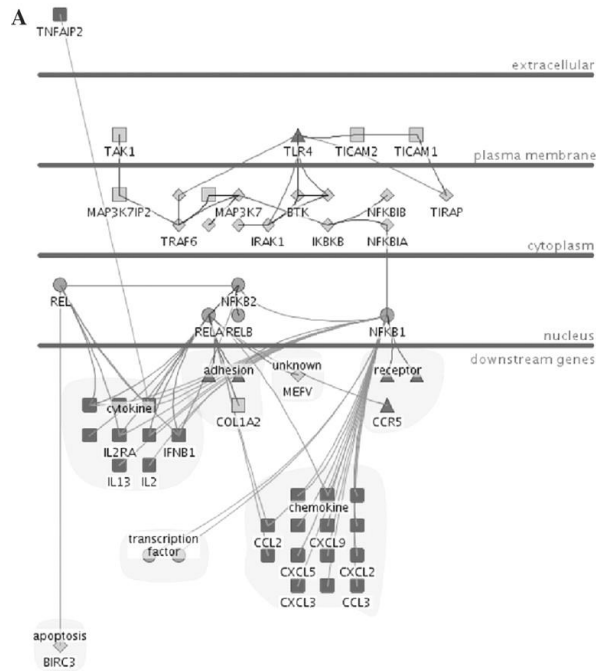
Why data vis (in general)?

Overview



Bring the computer in the loop

- Enable the exploration of large datasets
 - Temporal dimension



Cerebral: a Cytoscape plugin for layout of and interaction with biological networks using subcellular localization annotation.
 Aaron Barsky Jennifer L. Gardy Robert E. W. Hancock Tamara Munzner. *Bioinformatics*, Volume 23, Issue 8, 2007, 1040–1042,



Bring the computer in the loop

- Computational capacity
 - Scalability: How to deal with increasingly data sizes?
- Display capacity
 - Information density: limit of the displayed information



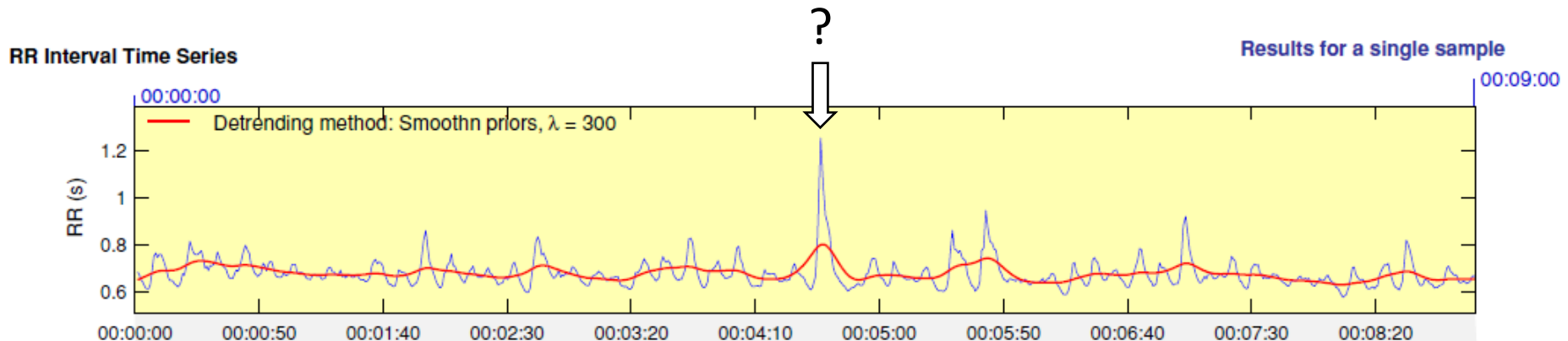
Bring the human in the loop



Bring the human in the loop



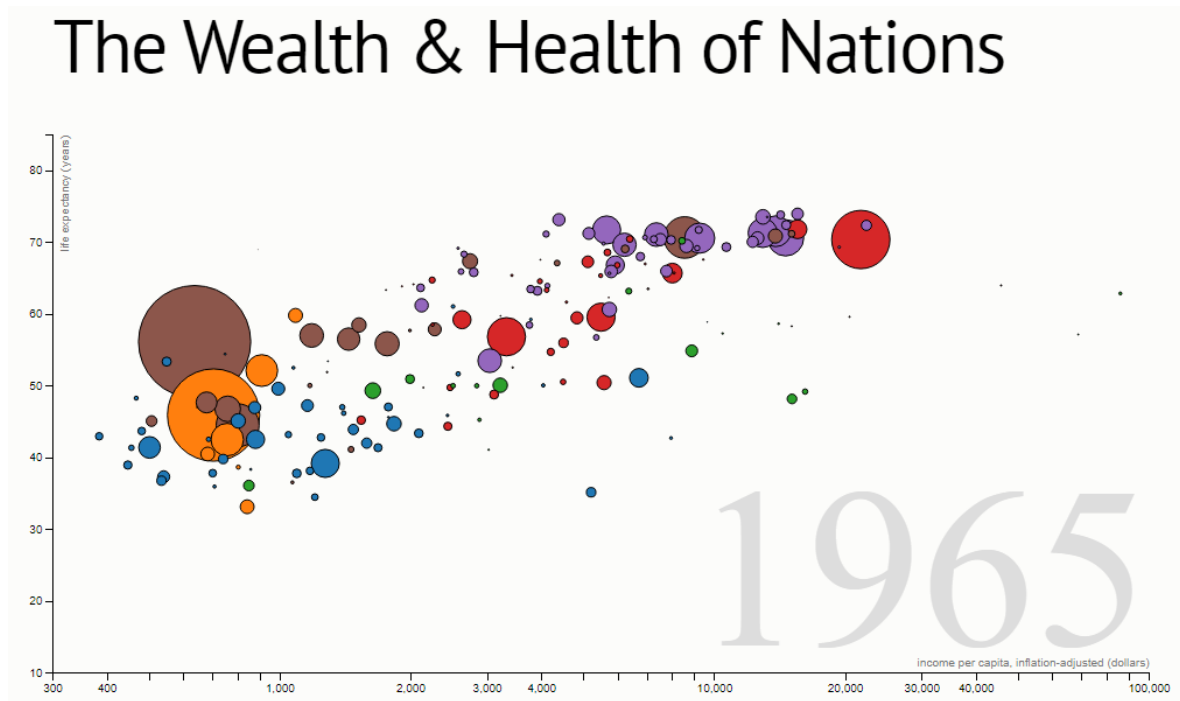
- The question is not known in advance
 - The analysis process is driven by the user
 - Take advantage of the ability of humans to find patterns
- Augment the human decision process
 - Improve decision making





Provide external representations

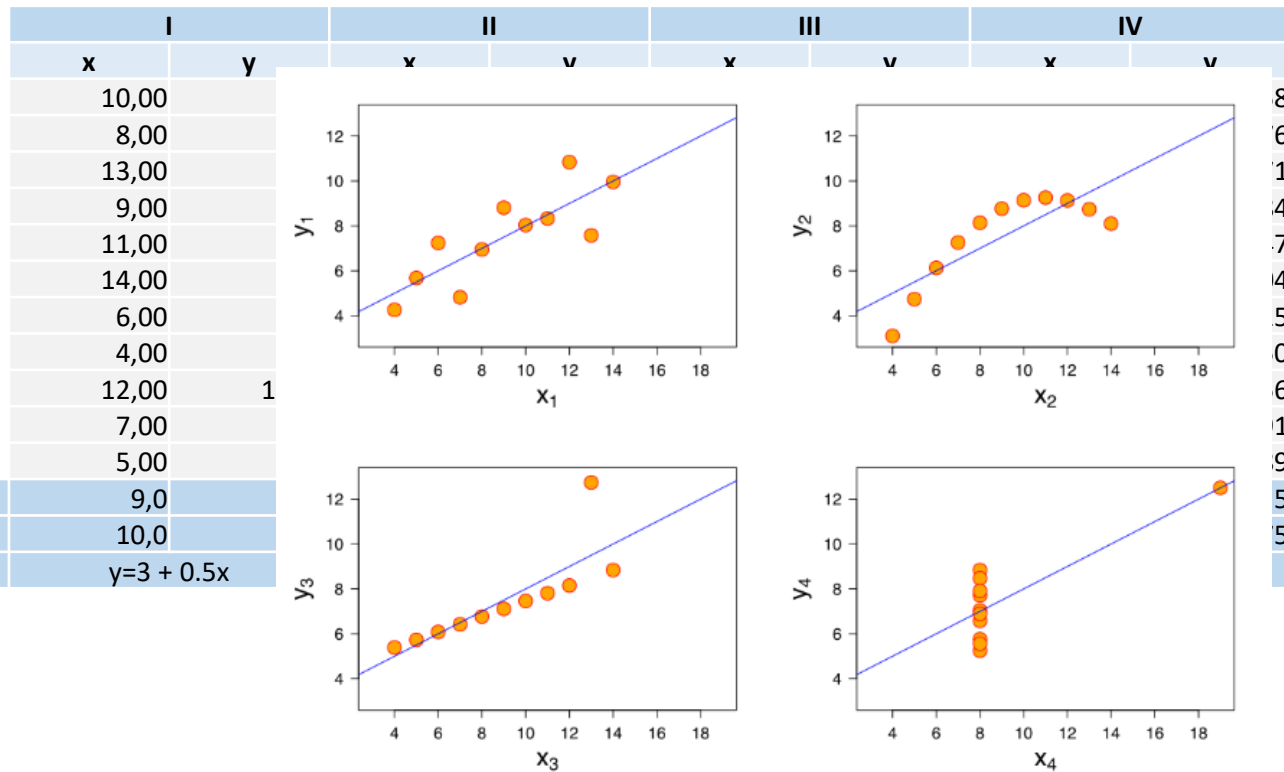
- Offload internal cognition and memory usage to the perceptual system
 - Data organized by spatial location
 - Support perceptual inference
 - Select just the relevant information





Show the data in detail

- The dataset structure in detail is better just a summary
 - Explore the data to find patterns

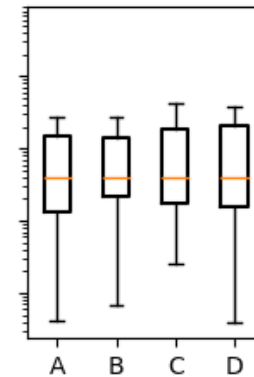
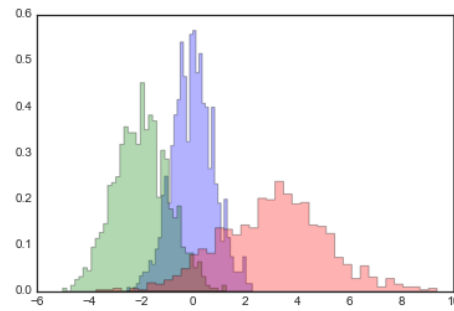
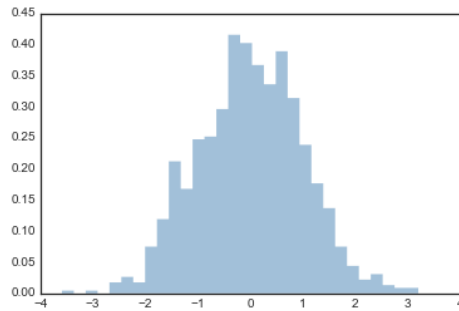


[Anscombe's Quartet]



Show the data in detail

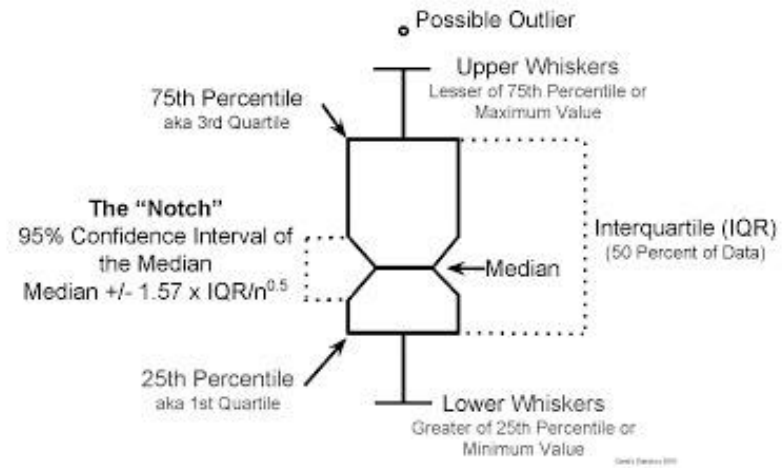
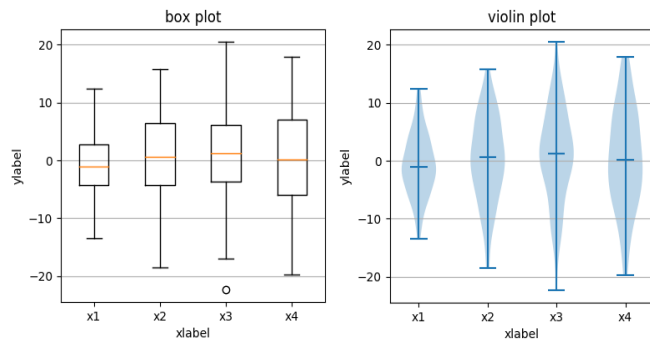
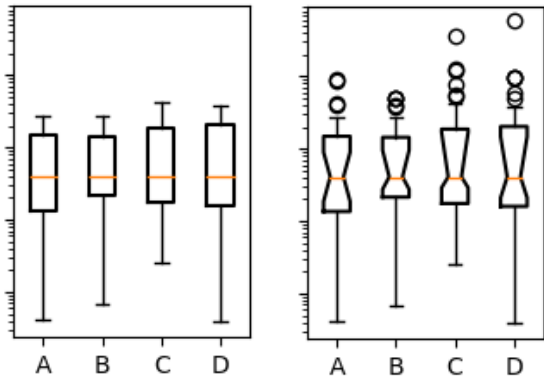
- The dataset structure in detail is better just a summary
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Show the data in detail

- The dataset structure in detail is better just a summary
 - Explore the data to find patterns

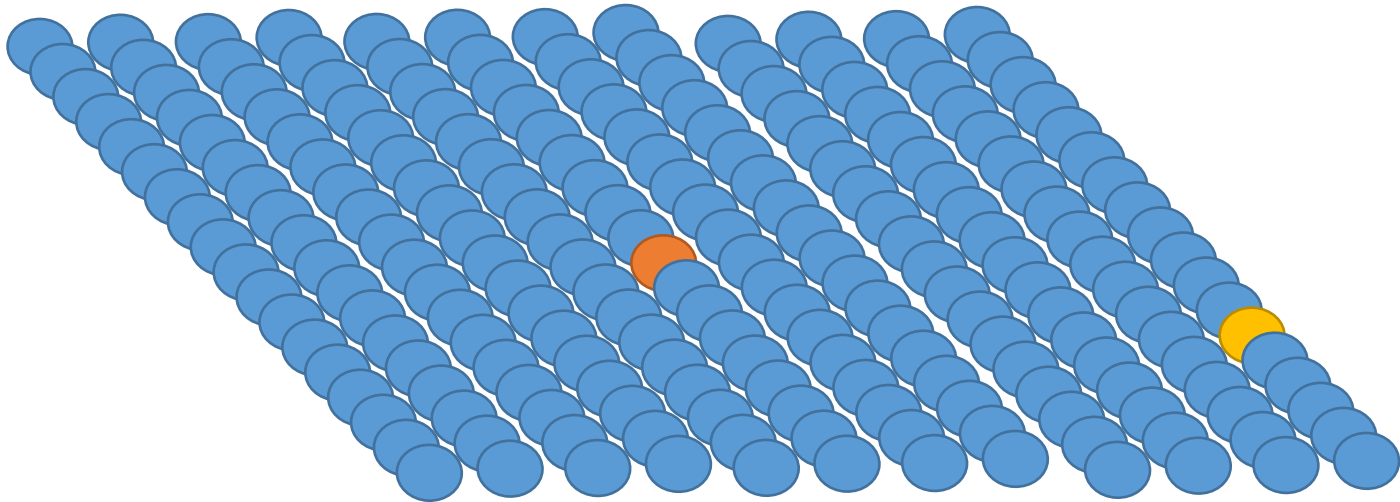


Chambers, John M., William S. Cleveland, Beat Kleiner, and Paul A. Tukey. "Comparing Data Distributions." In Graphical Methods for Data Analysis, 62. Belmont, California: Wadsworth International Group, 1983. ISBN 0-87150-413-8 International ISBN 0-534-98052-X



Depend on vision...

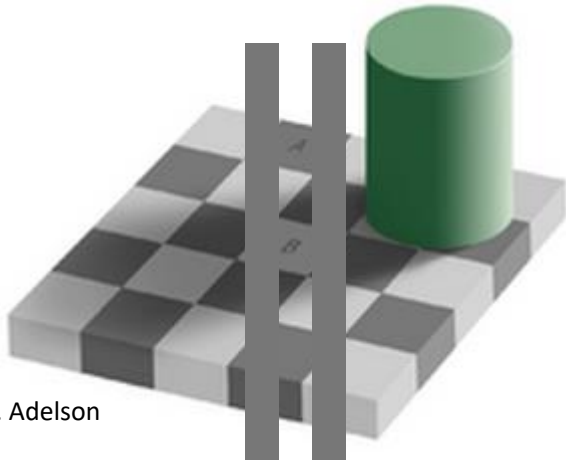
- The Visual system is a high-bandwidth channel of information
 - A significant amount of visual information is done in parallel at the preconscious level



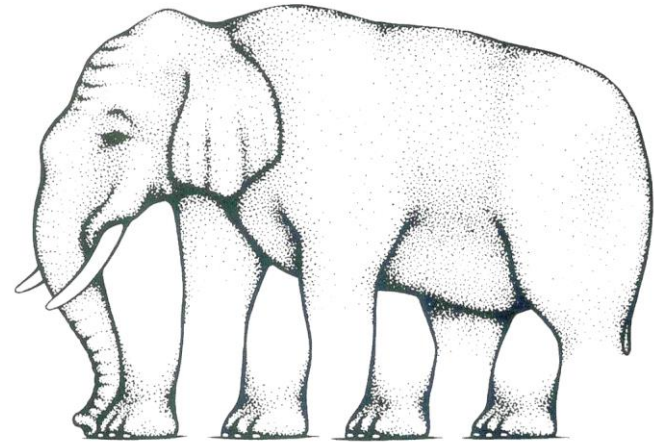


... but be careful

- Need to know its strengths and limitations



Edward H. Adelson





... but be careful

- Need to know its strengths and limitations
- Human perceptual and cognitive capacity limits?

THE ATTENTION TEST

<https://www.youtube.com/watch?v=uO8wpm9HSB0>



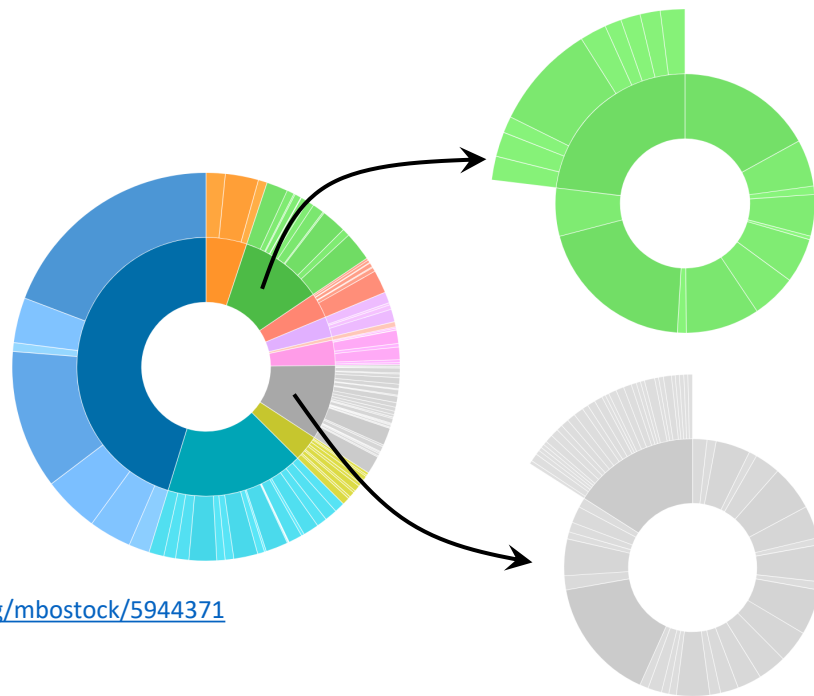
... but be careful

- Need to know its strengths and limitations
- Human perceptual and cognitive capacity limits?
 - Change blindness: the phenomenon where even very large changes are not noticed if we are attending to something else in our view



Interactivity

- For big datasets, the user nor the display can enable the visualization of the entire dataset at once

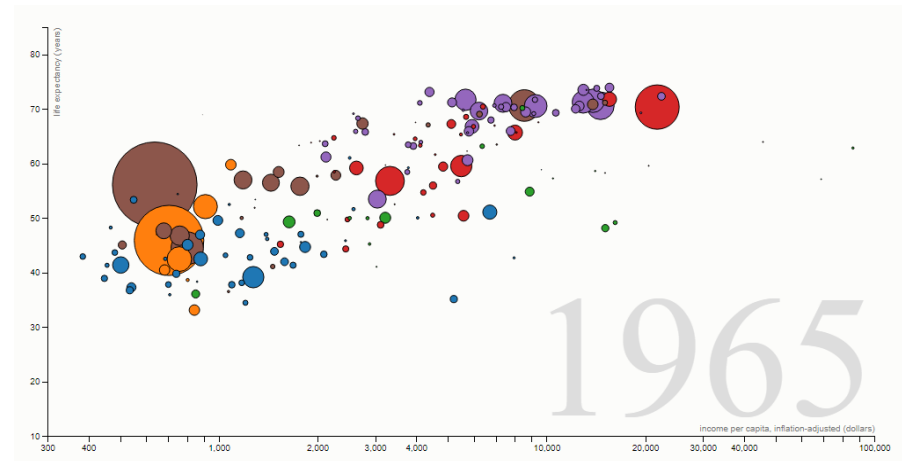
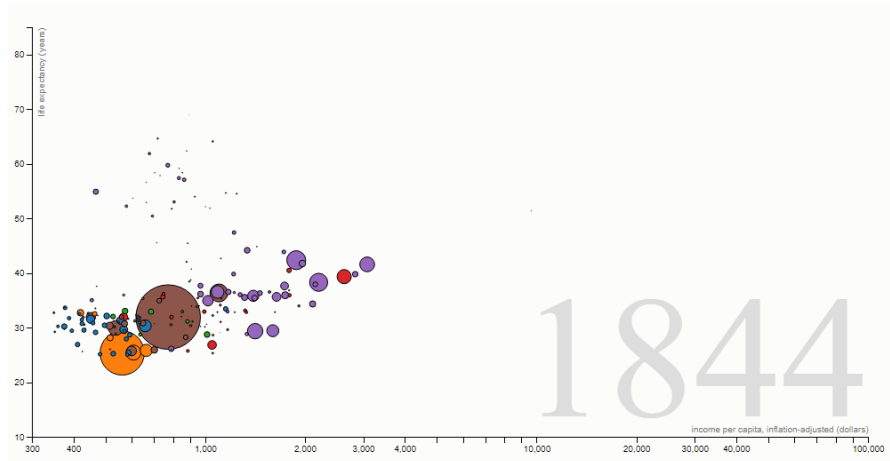


<https://bl.ocks.org/mbostock/5944371>



Interactivity

- A static view only allows for a single view of a limited number of dimensions

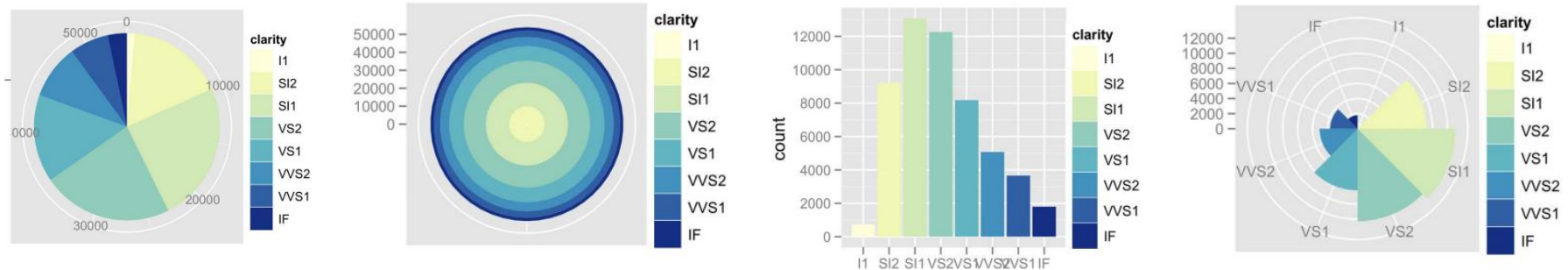


<https://bost.ocks.org/mike/nations/>



Interactivity

- A static view only allows for a single view of a limited number of dimensions
- The same dataset can be sorted using a number of different criteria



Hadley Wickham **A Layered Grammar of Graphics**. Journal of Computational and Graphical Statistics Vol. 19 , Iss. 1,2010



Interactivity

datavoyager Bookmarks (1) Undo Redo

Data
Cars

Fields

- A Cylinders
- A Name
- A Origin
- # Year
- # Acceleration
- # Displacement
- # Horsepower
- # Miles per Gallon
- # Weight in lbs
- # COUNT

Wildcards

- A Categorical Fields
- # Temporal Fields
- # Quantitative Fields

Encoding

x
y
column
row

Marks

size
color
shape
detail
text

Filter

Specified View
No specified visualization yet. Start exploring by dragging a field to encoding pane on the left or examining univariate summaries below.

Related Views

Univariate Summaries

A Cylinders # COUNT

Cylinders	COUNT
3	0
4	200
5	0
6	100
8	100

A Name # COUNT

Name	COUNT
amc ambassador brougham	1
amc ambassador dpi	1
amc ambassador sst	1
amc concord	2
amc concord dl	1
amc concord dl 6	1
amc concord dl 6	1
amc gremlin	1
amc hornet	1
amc hornet sportabout (s...)	1
amc matador	5
amc matador (sw)	1
amc pacer	1
amc pacer dl	1
amc rebel sst	1
amc rebel sst (sw)	1
amc spirit dl	1
audi 100 ls	1
audi 100ls	1
audi 4000	1
audi 5000	1
audi 5000s (diesel)	1
audi fox	1
bmw 2002	1
bmw 320i	1
buick century	1

A Origin # COUNT

Origin	COUNT
Europe	50
Japan	50
USA	250

YEAR (Year) # COUNT

YEAR	COUNT
1970	35
1971	30
1972	30
1973	40
1974	25
1975	30
1976	35
1977	30
1978	35
1979	30
1980	30
1981	30
1982	65

BIN (Acceleration) # COUNT

BIN (Displacement) # COUNT

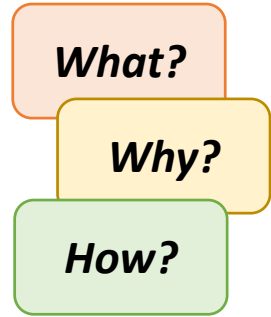
BIN (Horsepower) # COUNT















BIN (Miles per Gallon) # COUNT

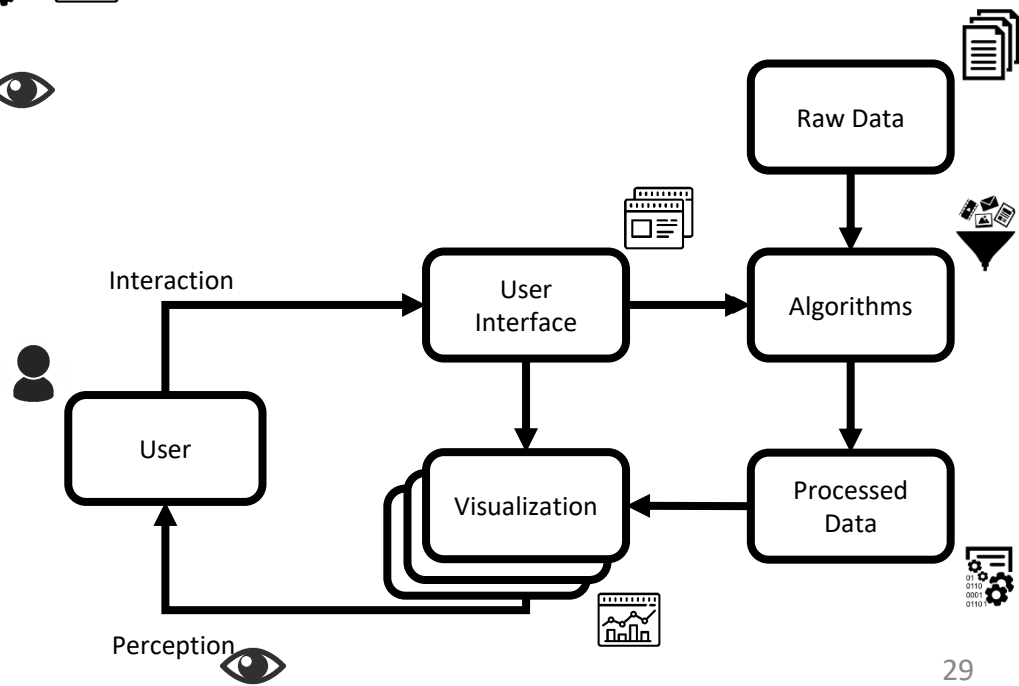
Debug Report an issue

<https://vega.github.io/voyager2/>

Wrap Up



- Bring the computer in the loop  
- Bring the human in the loop   
- Provide external representations  
- Show data in detail  
- Depend on vision  
- Interactivity   



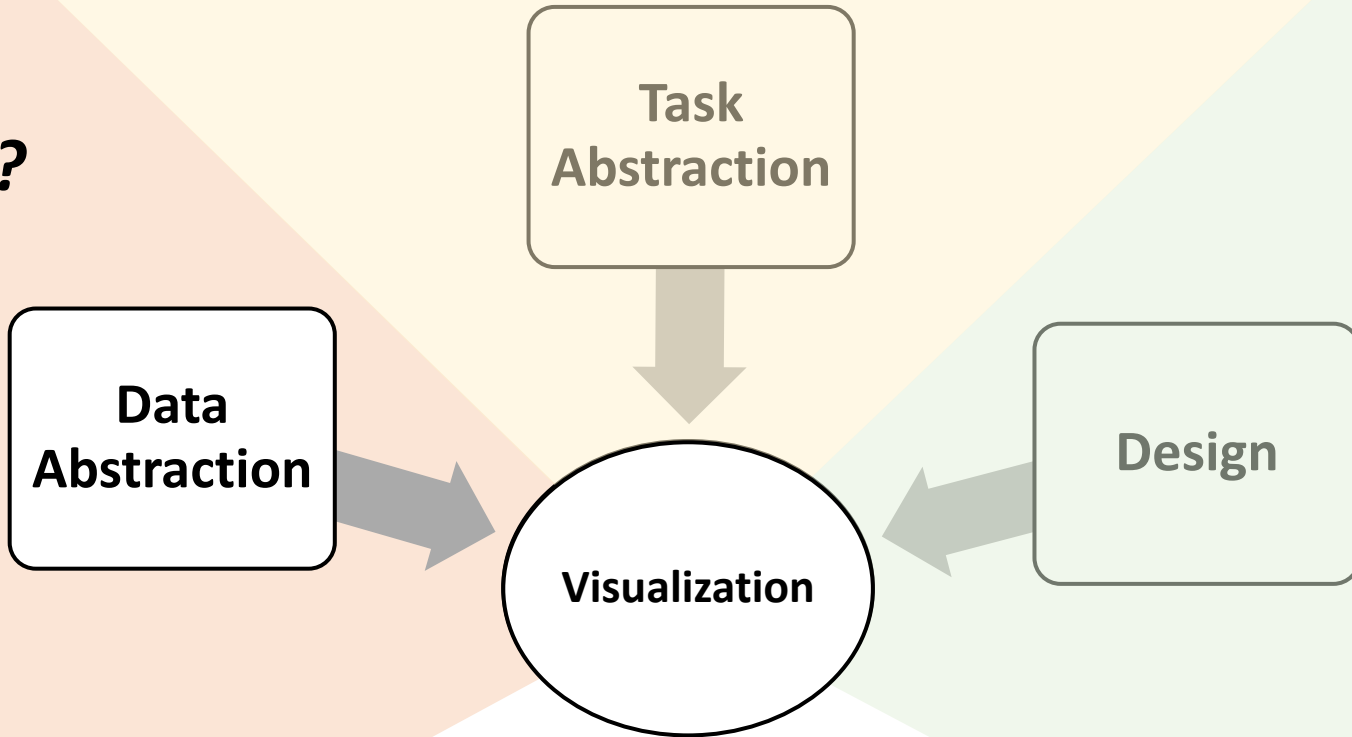
Visualization Design

Methodology to design visualizations

Why?

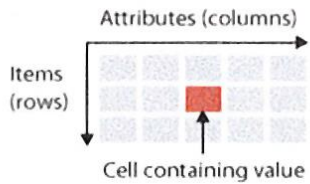
What?

How?

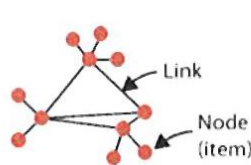


Data and Dataset Types

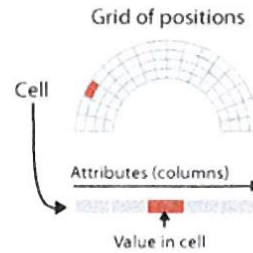
→ Tables



→ Networks

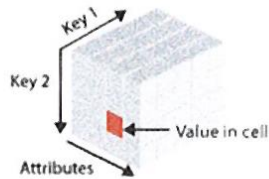


→ Fields (Continuous)



- Items
- Attributes
- Links
- Positions
- Grids

→ *Multidimensional Table*



→ *Trees*



→ Geometry (Spatial)



→ Dataset Availability

→ Static

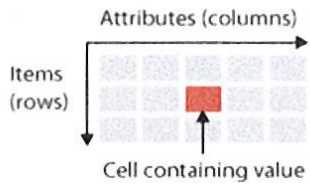


→ Dynamic

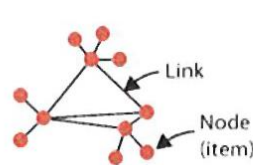


Data and Dataset Types

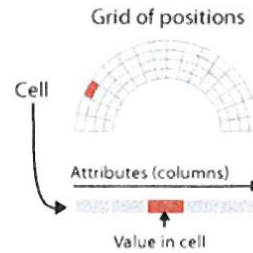
→ Tables



→ Networks

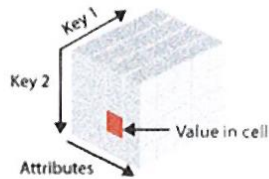


→ Fields (Continuous)



- Items
- **Attributes**
- Links
- Positions
- Grids

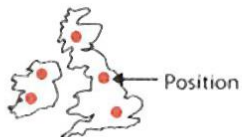
→ *Multidimensional Table*



→ *Trees*



→ Geometry (Spatial)



→ Dataset Availability

→ Static



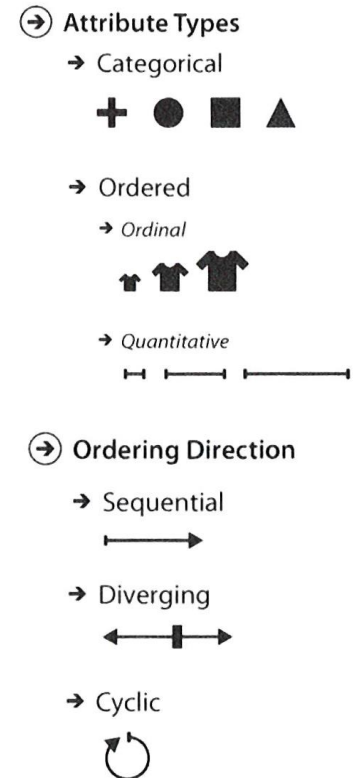
→ Dynamic



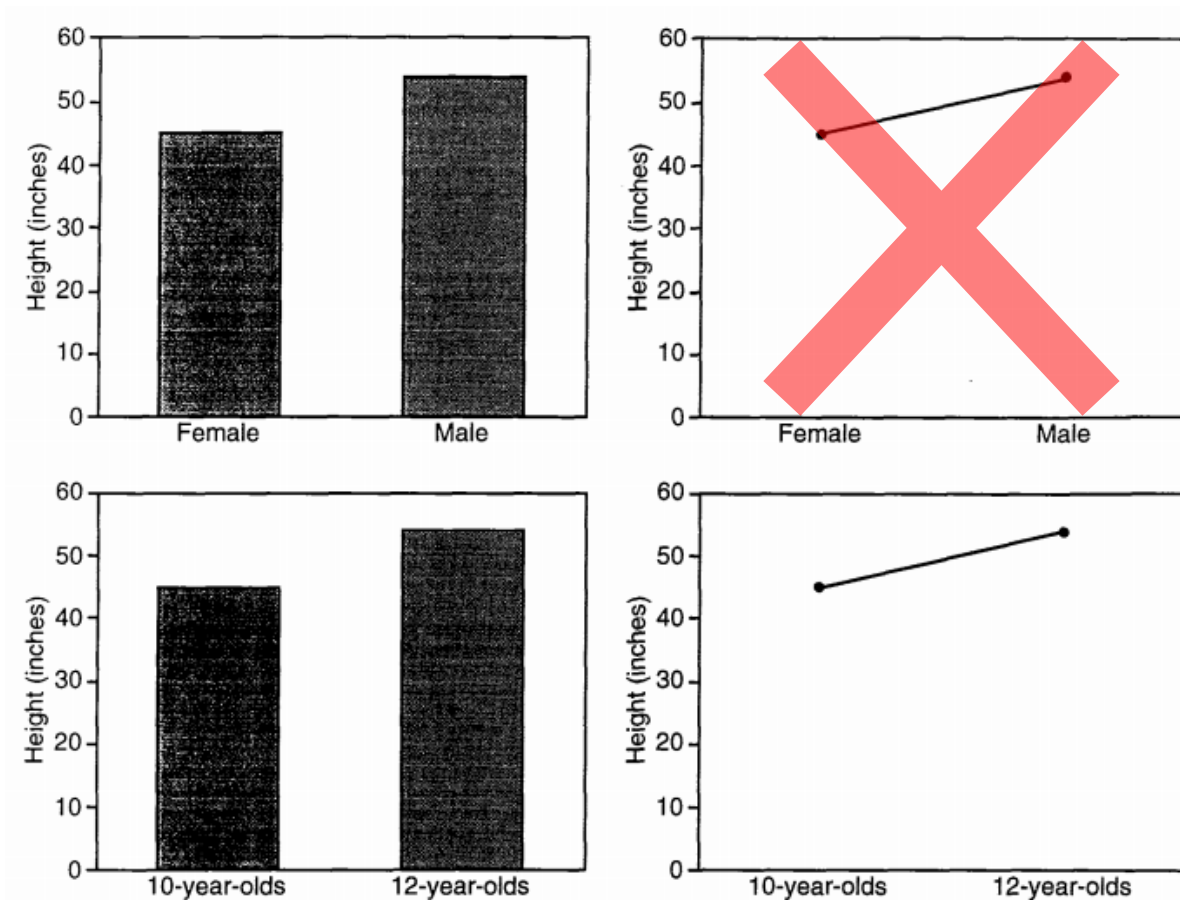
Attributes

An *attribute* is some specific property that can be measured, observed or logged

- ➔ • Categorical
 - Non-numerical, limited number of values, usually fixed
 - Non-ordered, unless explicitly stated
- Ordinal
 - Non-numerical, limited number of values, usually fixed
 - Ordered
- Numerical
 - Quantitative, Infinite number of values
 - Ordered



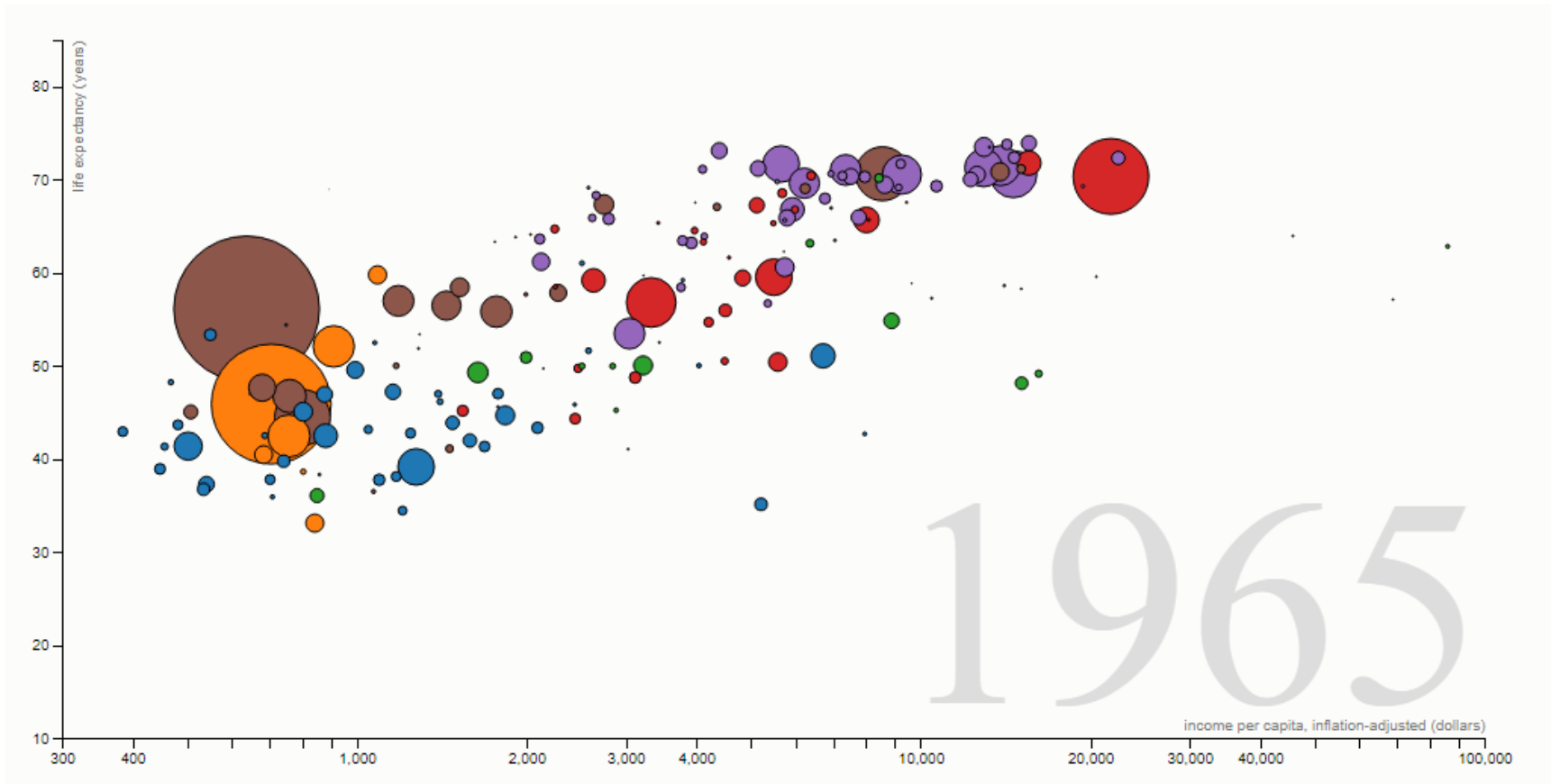
Bar Charts vs Line Charts



Zacks, Jeff, and Barbara Tversky. "Bars and lines: A study of graphic communication." *Memory and Cognition* 27 (1999): 1073-1079.

What?

Scaterplot

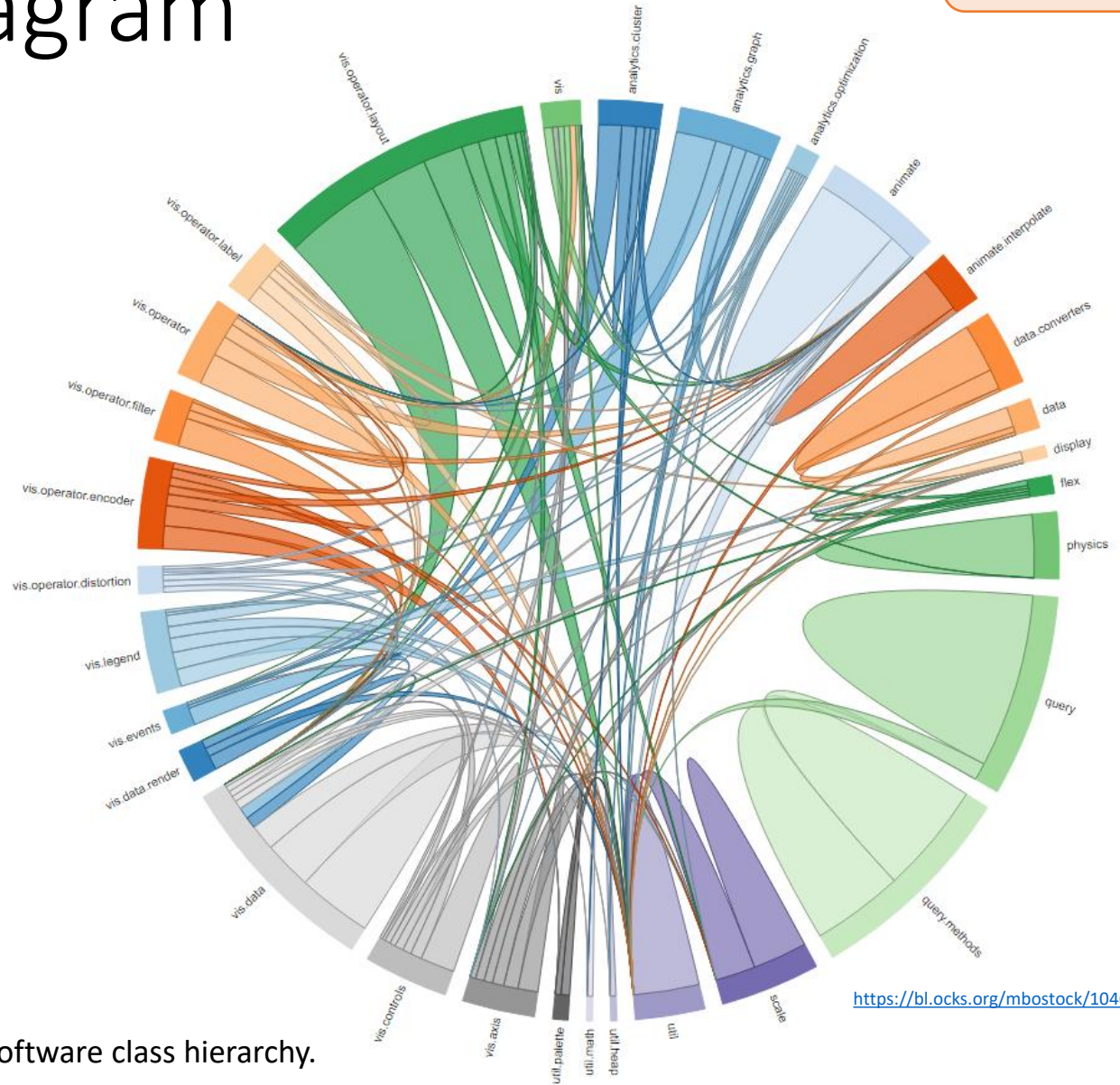


<https://bost.ocks.org/mike/nations/>

Multi dimensional table: N countries * T time
Attributes: 3 Qualitative

What?

Chord Diagram



Network

Dependencies between classes in a software class hierarchy.

<https://bl.ocks.org/mboostock/1046712>

What?

Time Varying Data

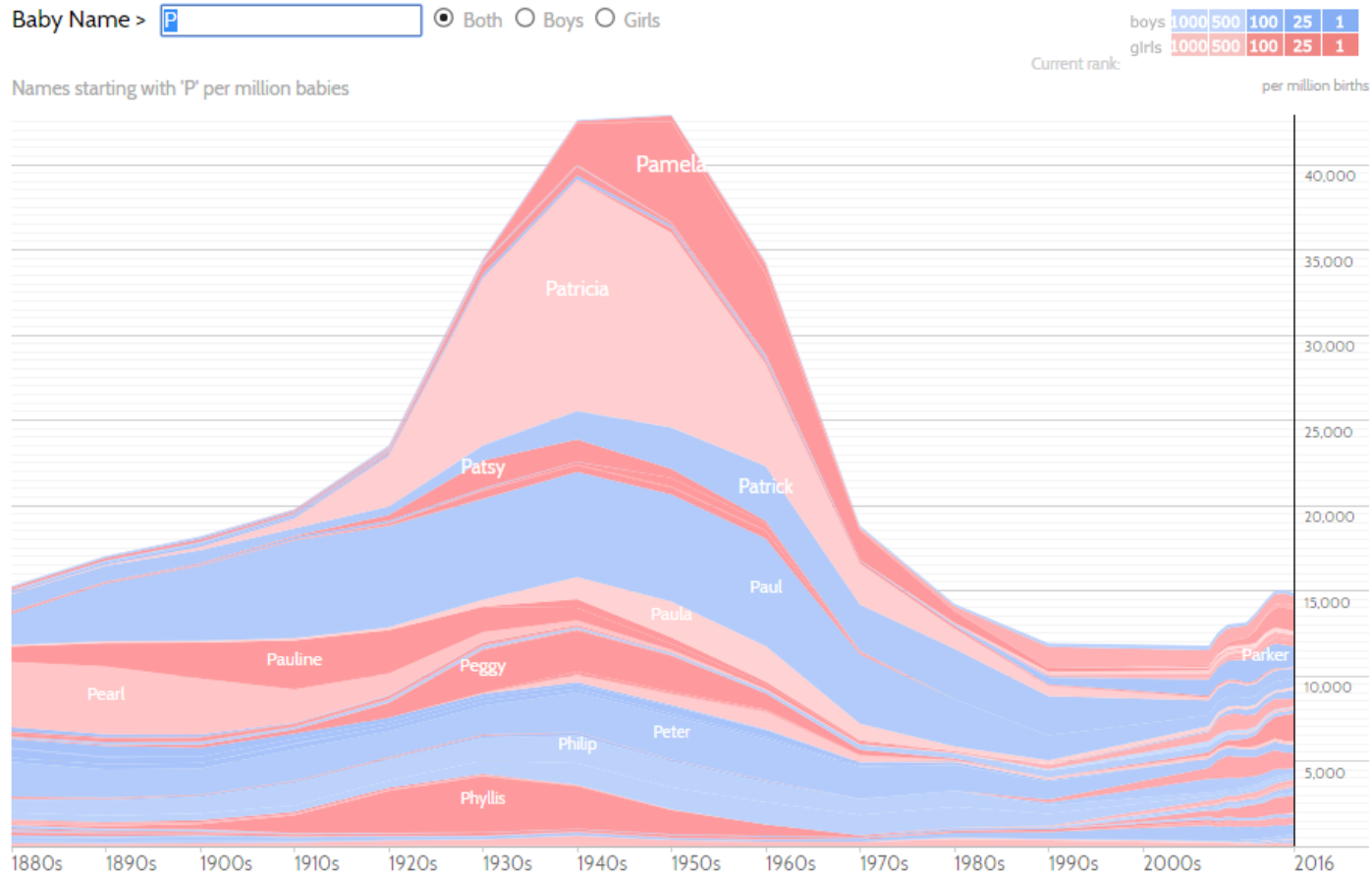


Table: T time

Attributes: 1 Categorical, 1 Quantitative

<http://www.babynamewizard.com/voyager>

Tensor Visualization

Comparison: Ellipsoids vs. superquadrics (Kindlmann)

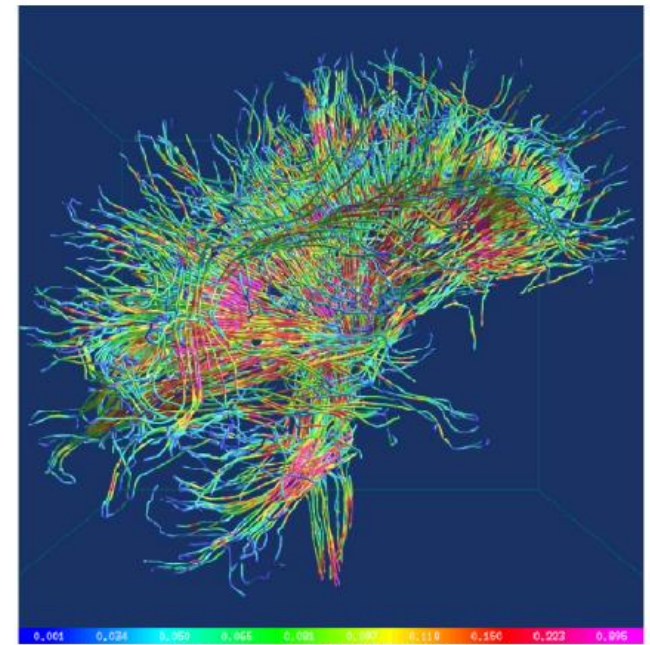
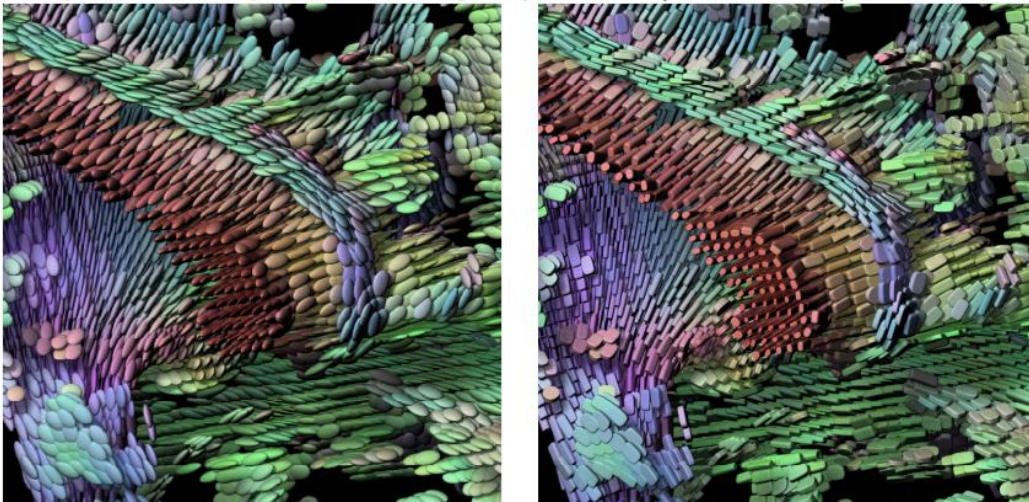


Image credit: W. Shen

Why?

**Task
Abstraction**



How?

Design



What?

**Data
Abstraction**



Visualization

Task Abstraction

- Reframe tasks from domain-specific form into abstract form
 - Ease the design process using references
- The focus on tasks enables the improvement of them
 - Making more effective, user-friendly...
- Tasks can be classified in:
 - Actions – Verb - High-level task
 - Targets – Noun - Precise goal

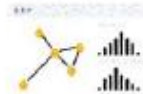
Action Classification

- Analyze
 - Consume

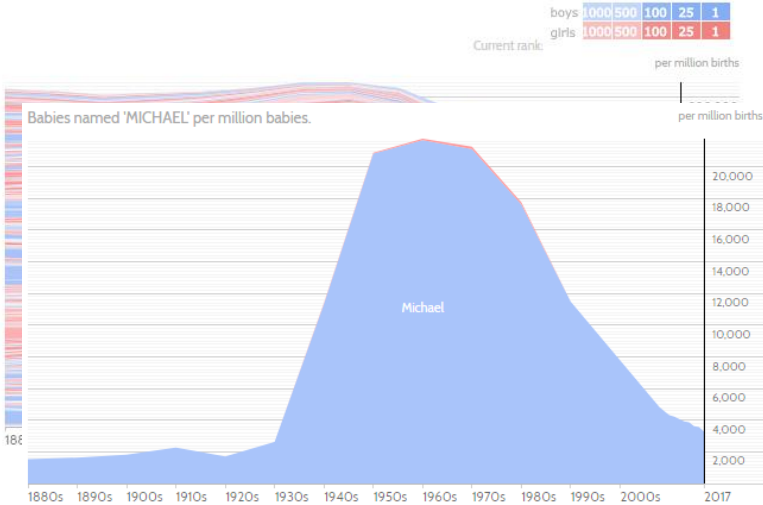
→ Discover



→ Present



→ Enjoy



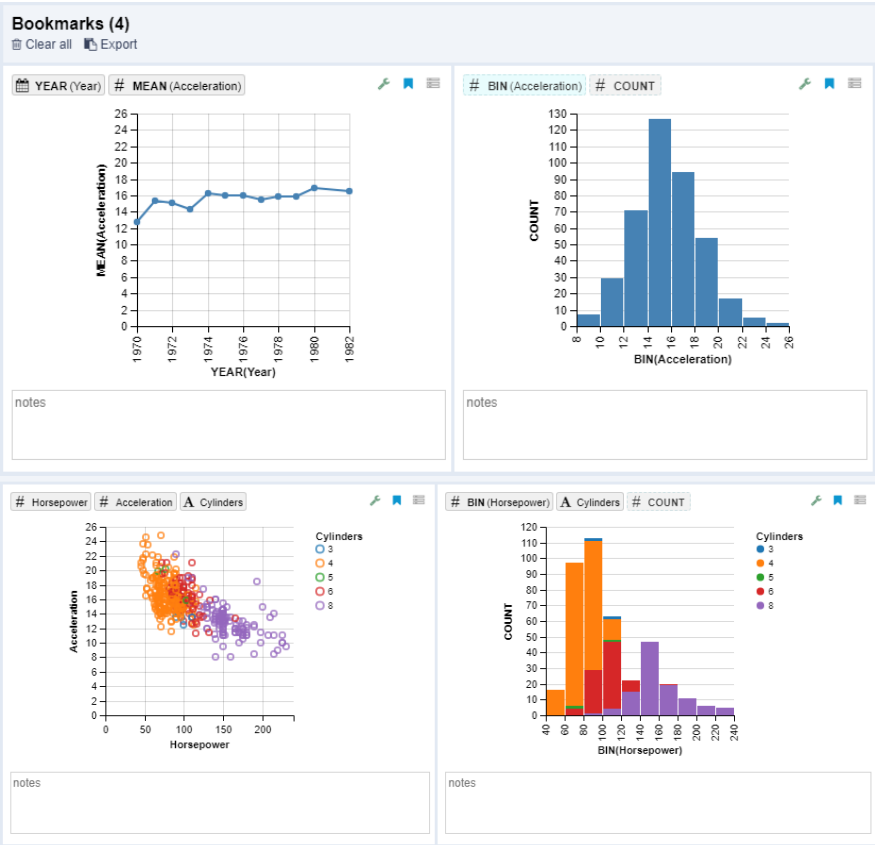
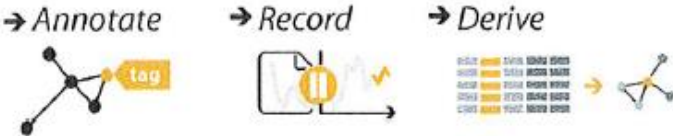
Why?

Action Classification

- Analyze
 - Consume



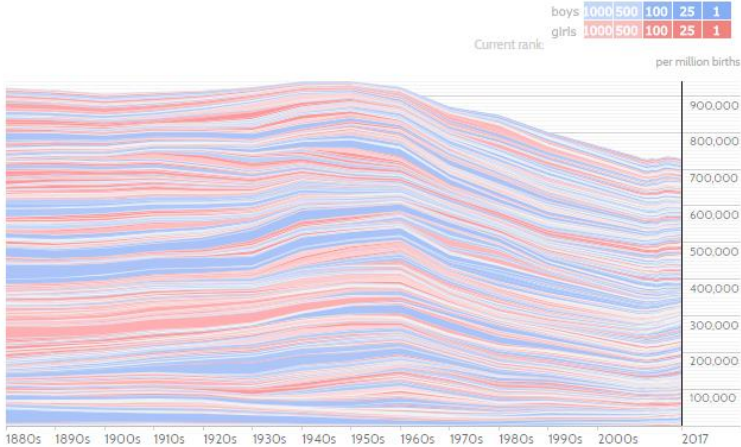
- Produce



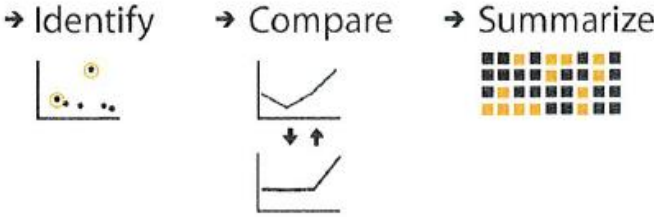
Action Classification

- Search

	Target known	Target unknown
Location known	<i>Lookup</i>	<i>Browse</i>
Location unknown	<i>Locate</i>	<i>Explore</i>



- Query

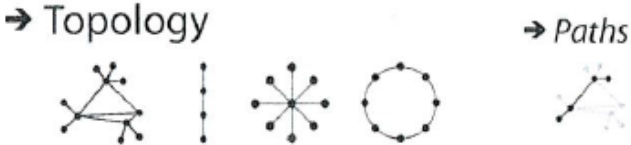


Target Classification

- All Data

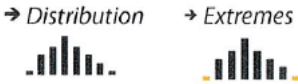


- Network Data

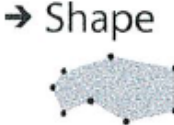


- Attributes

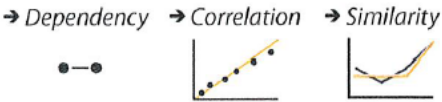
- One



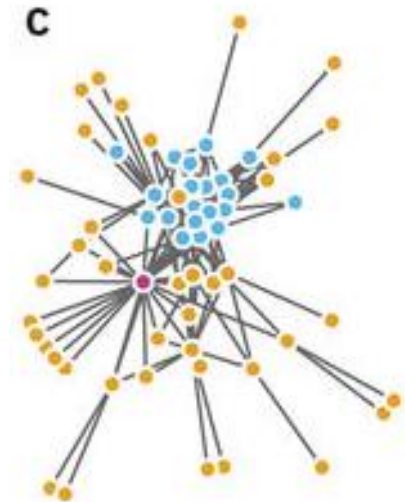
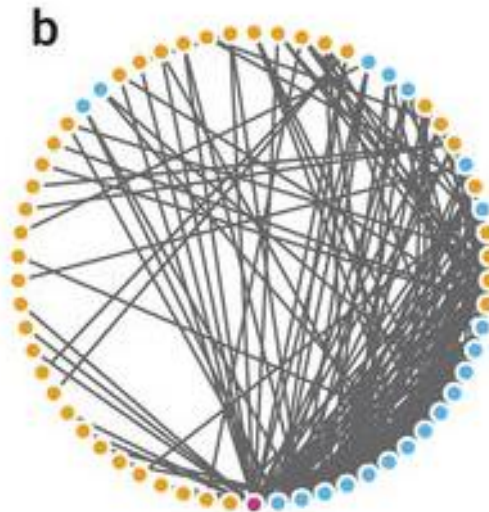
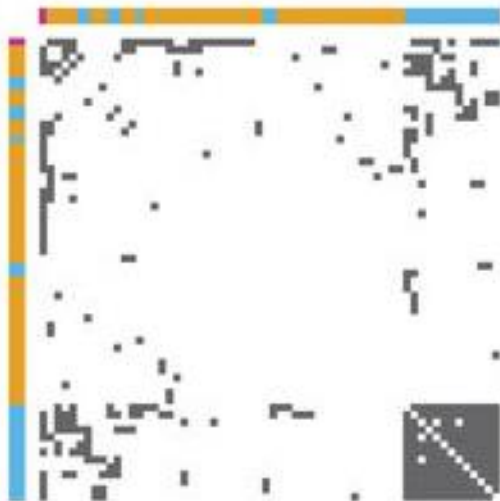
- Spatial Data



- Many



Network Visualization



Gehlenborg, Nils, and Bang Wong. **Points of view: Networks**
Nature methods 9.2 (2012): 115-115.

Why?

Task
Abstraction



How?

Design



What?

Data
Abstraction



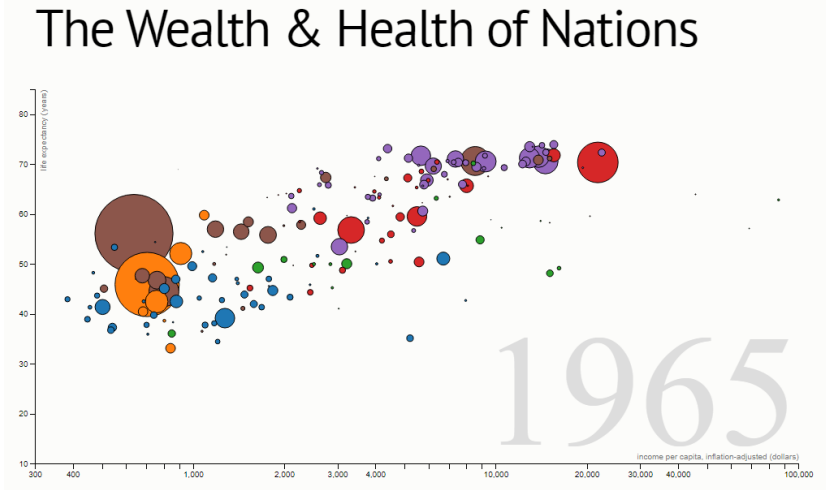
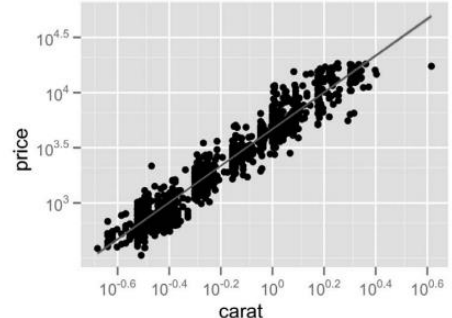
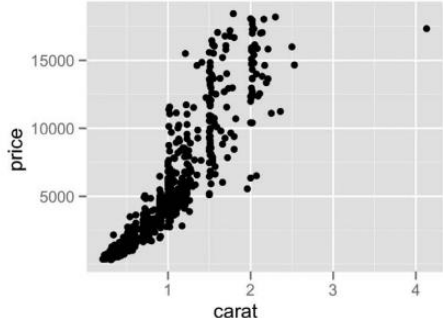
Visualization

How to...

- ... **arrange** the data in the view spatially?

Arrange / Encode

→ Express



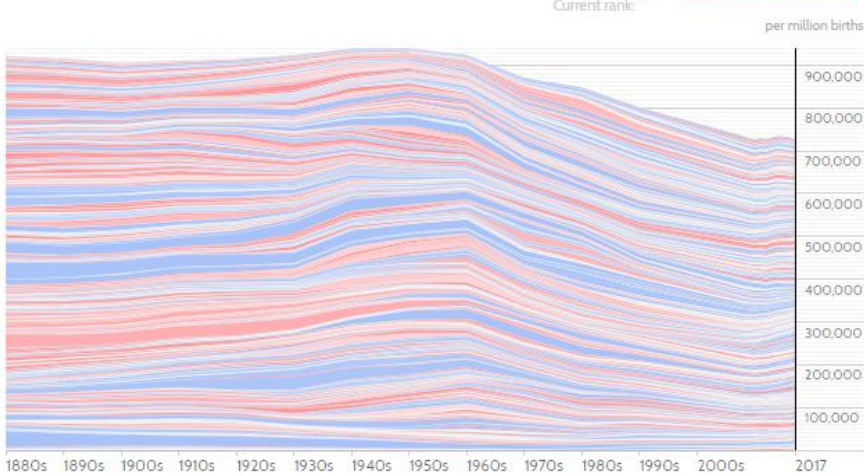
How?

Arrange / Encode

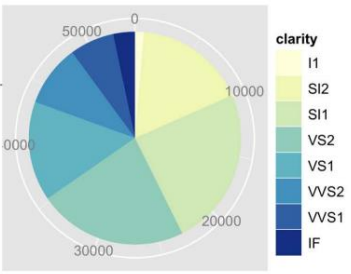
→ Separate



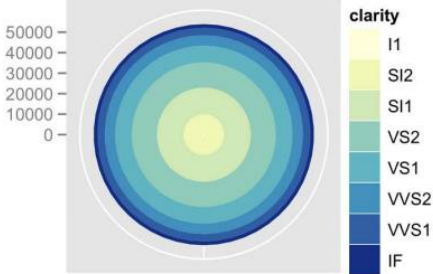
boys	1,000	500	100	25	1
girls	1,000	500	100	25	1



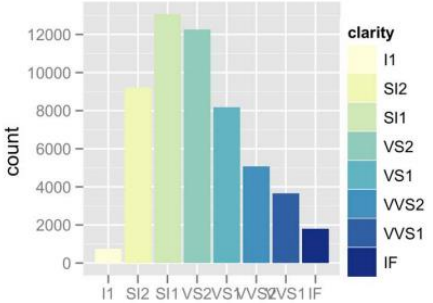
Streamgraphs



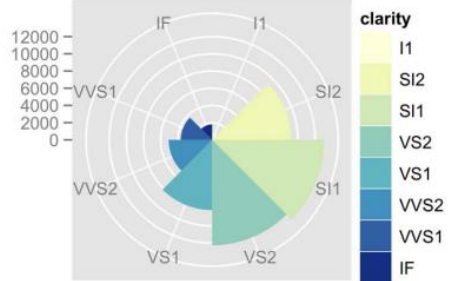
Pie Chart



Bullseye chart



Bar Plot



Coxcomb plot

Hadley Wickham A Layered Grammar of Graphics. Journal of Computational and Graphical Statistics Vol. 19, Iss. 1, 2010

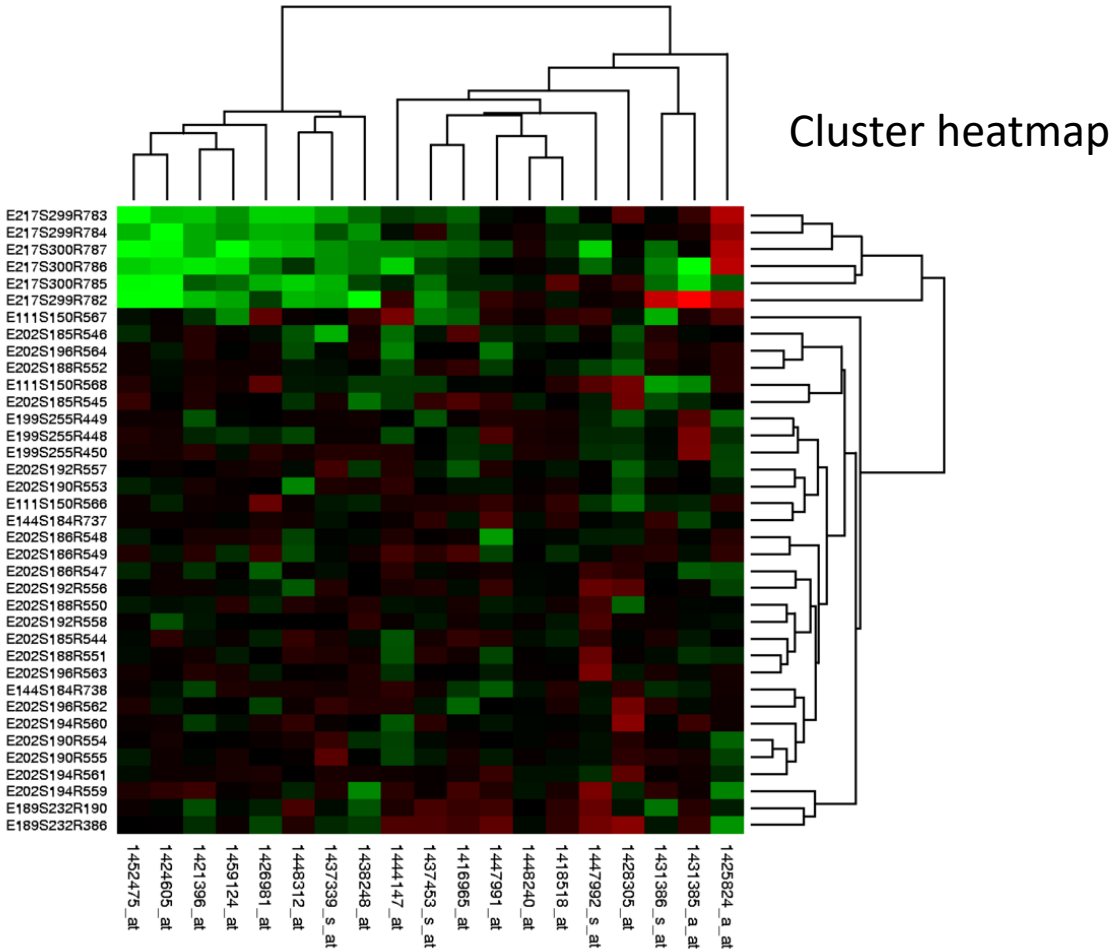
How?

Arrange / Encode

→ Order



→ Align



https://en.wikipedia.org/wiki/Heat_map

Arrange / Encode

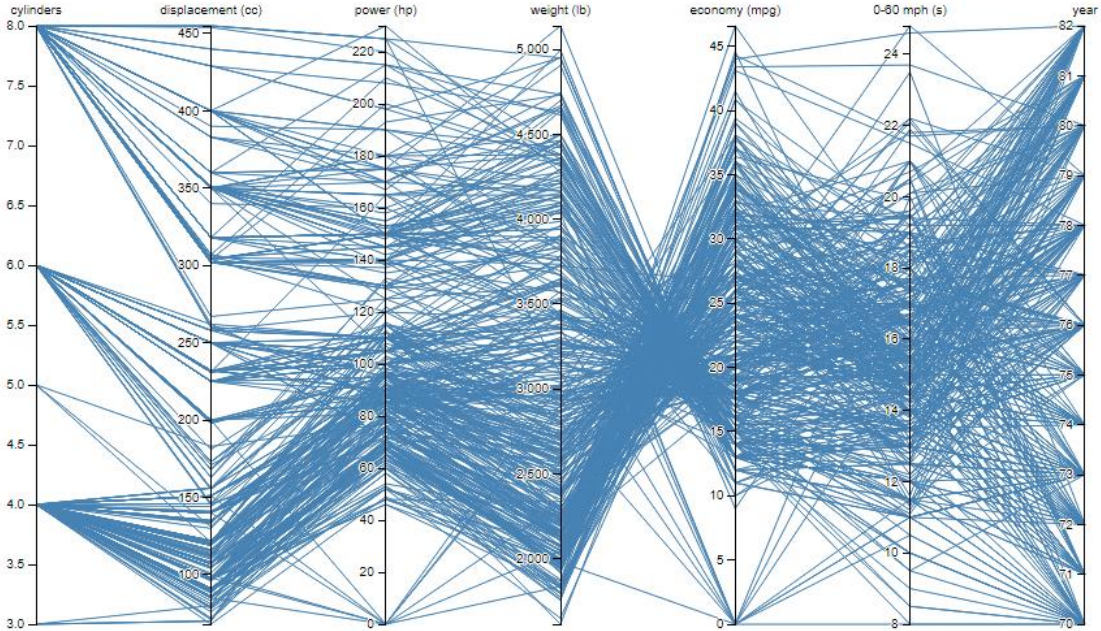
→ Order



→ Align



Parallel Coordinates



<https://bl.ocks.org/jasondavies/1341281>

How to...

- ... **arrange** the data in the view spatially?
- ... **map** the data with all non-spatial channels?

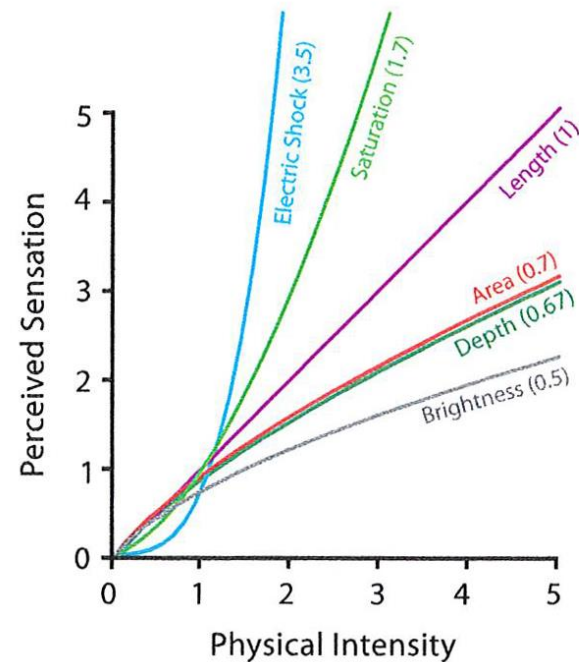
Display

- What is the best way to visually encode the data ?
 - How the human visual system discriminate visual stimuli?

- Two major dimensions

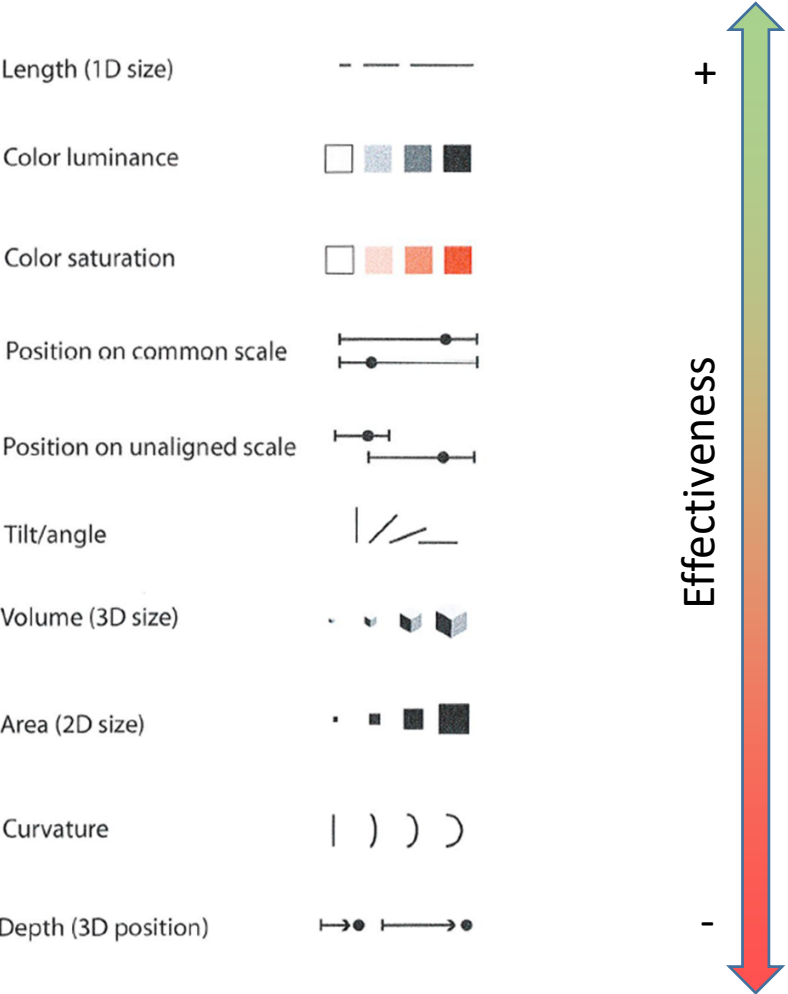
- Color
- Shape

Steven's Psychophysical Power Law: $S = I^N$

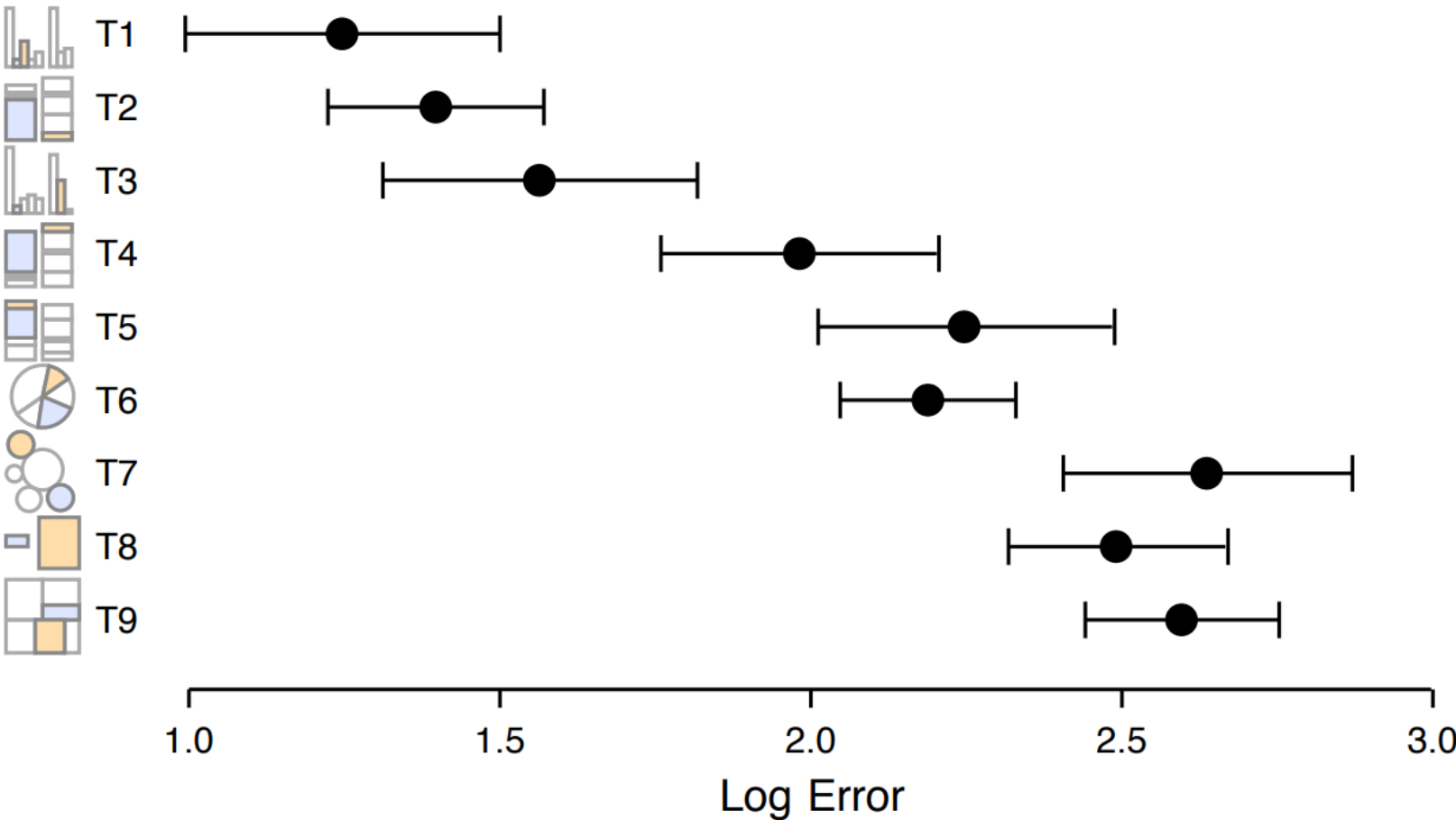


How?

Visual Encoding



Visual discrimination



Crowdsourcing graphical perception: using mechanical turk to assess visualization design. Jeffrey Heer and Michael Bostock. 2010. ACM SIGCHI Conference on Human Factors in Computing Systems, p. 203-212

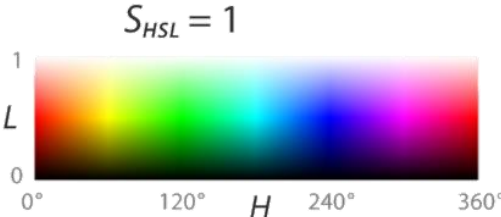
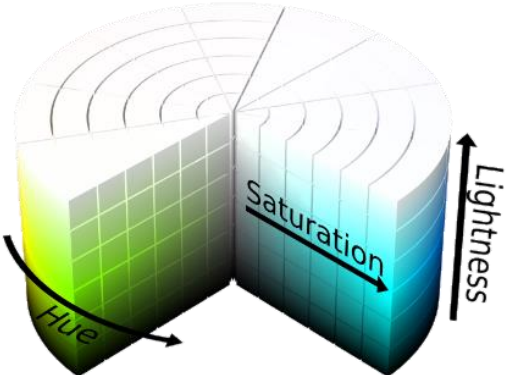
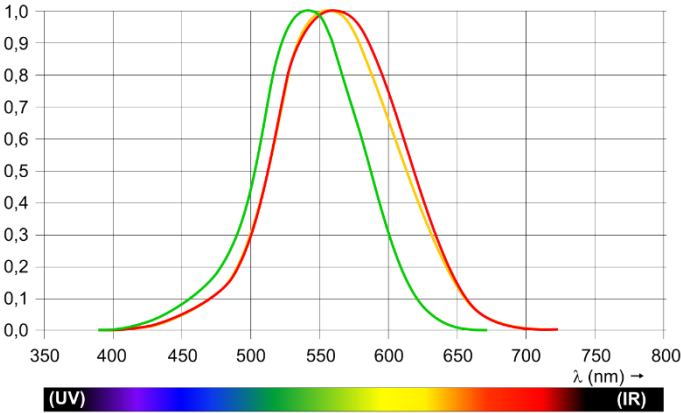
How?

Color Encoding

- Color Components

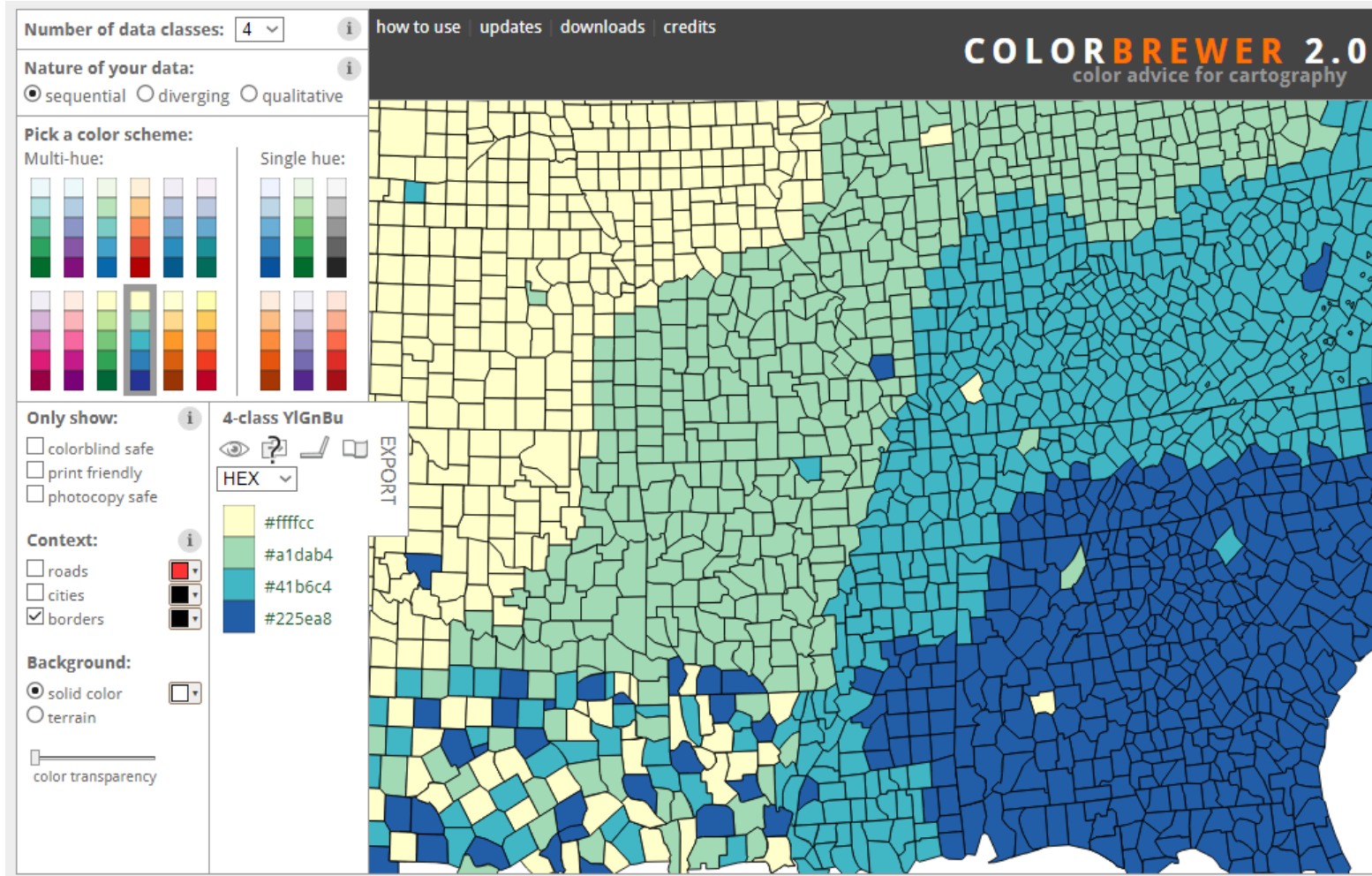


- Color sensitivity



<https://commons.wikimedia.org/wiki/File:LuminosityCurve2.svg>
https://commons.wikimedia.org/wiki/File:Hsl-hsv_models_b.svg

Color Encoding



<http://colorbrewer2.org/>

Color Use Guidelines for Data Representation, Brewer, C. A. 1999. Proceedings of the Section on Statistical Graphics, American Statistical Association, Alexandria VA. pp. 55-60.

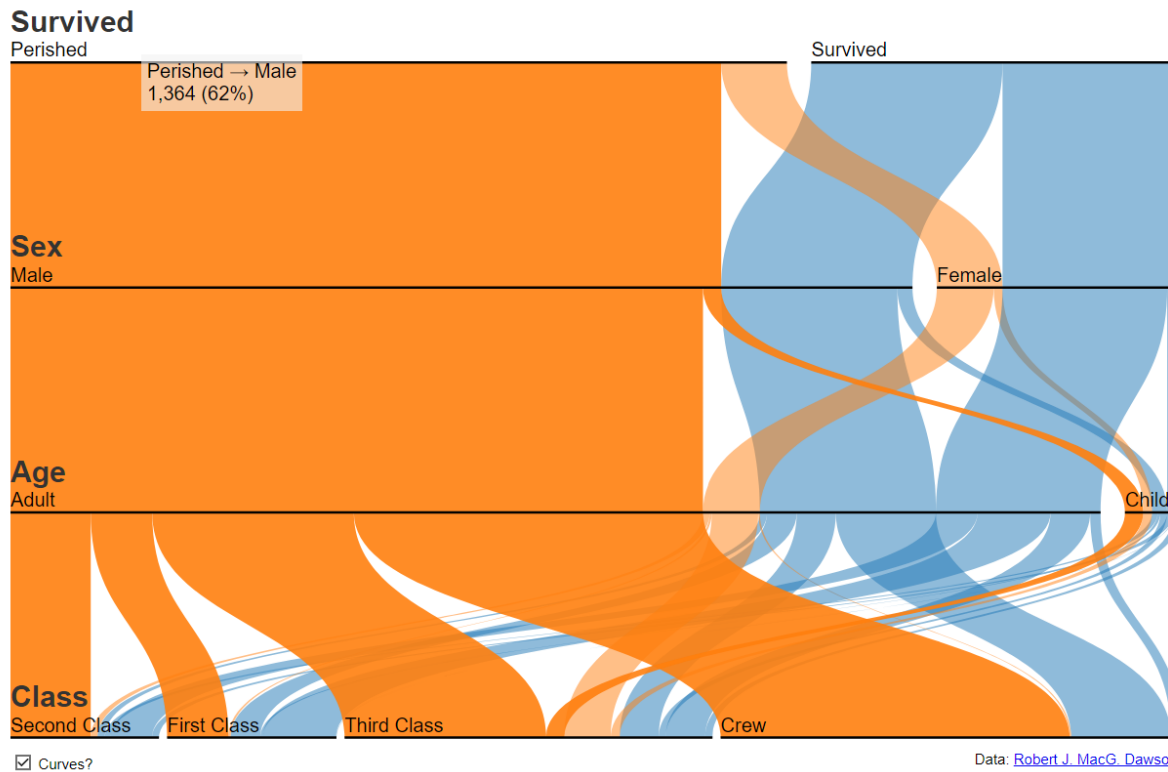
How to...

- ... **arrange** the data in the view spatially?
- ... **map** the data with all non-spatial channels?
- ... **manipulate** the data in the view?

How?

Manipulate

Titanic Survivors



➔ Change



➔ Select

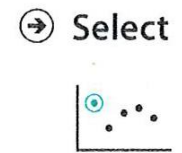
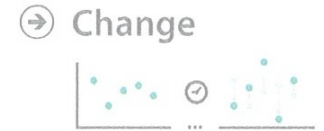
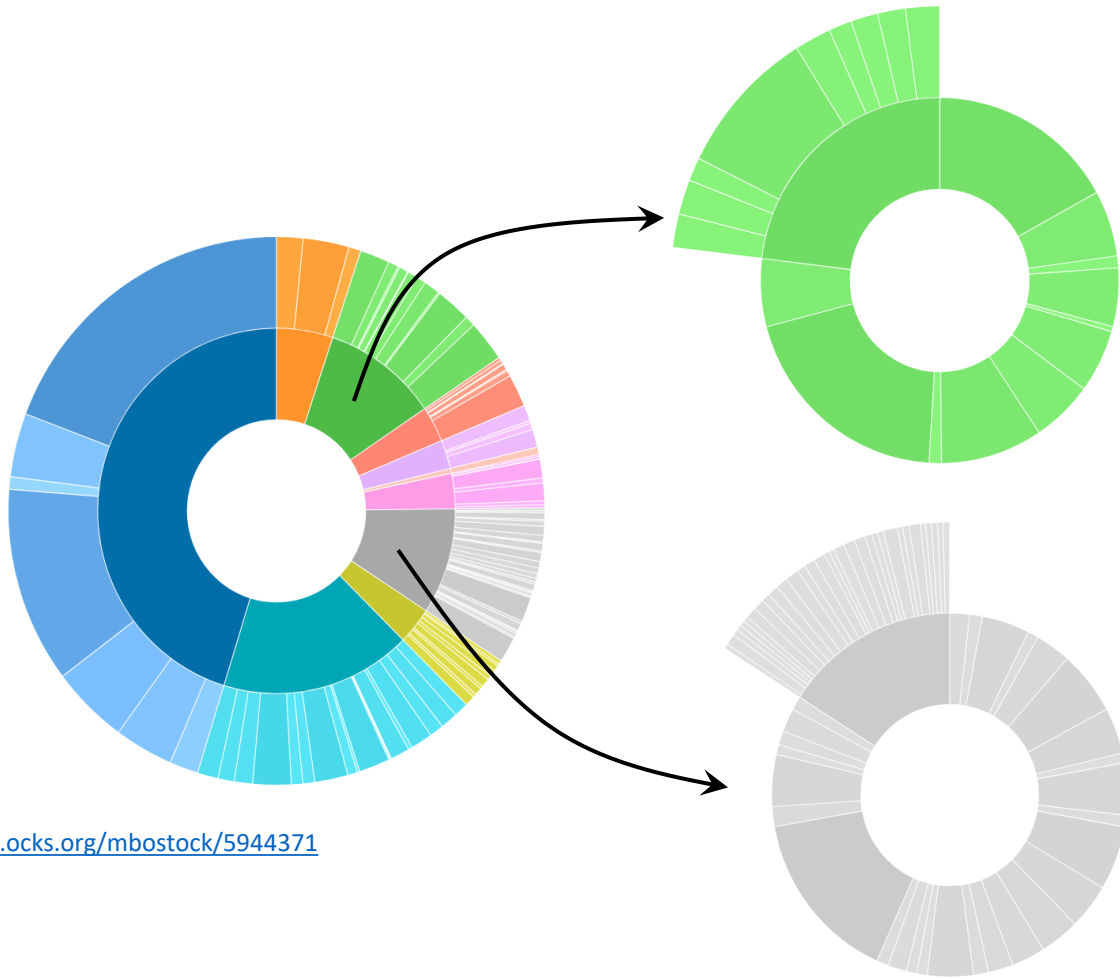


➔ Navigate



<https://www.jasondavies.com/parallel-sets>

Manipulate



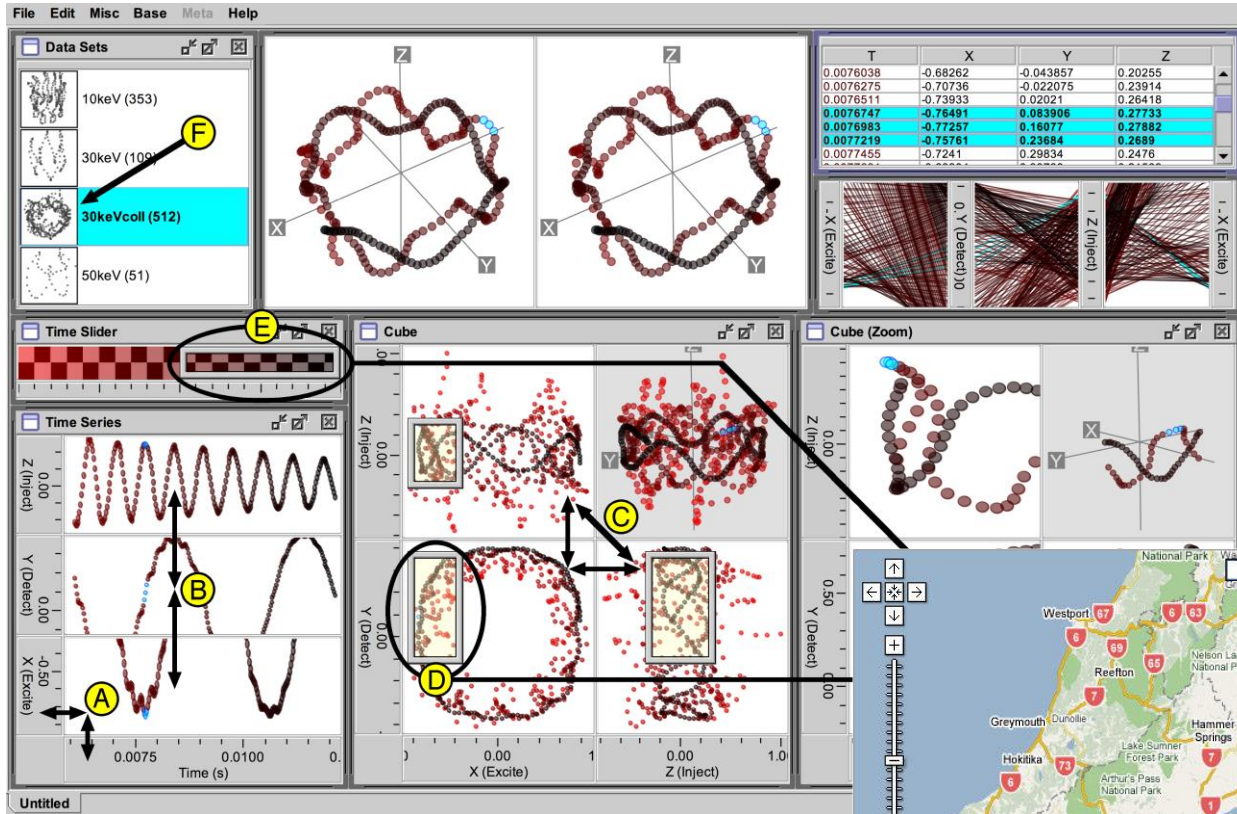
<https://bl.ocks.org/mbostock/5944371>

How to...

- ... **arrange** the data in the view spatially?
- ... **map** the data with all non-spatial channels?
- ... **manipulate** the data in the view?
- ... **facet** data between views?

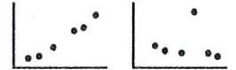
How?

Facet (multi-view)

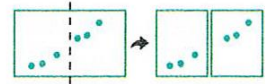


Weaver. Building Highly-Coordinated Visualizations In Improvise. Proc. InfoVis 2004, p. 159-166

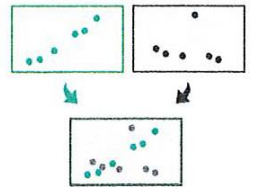
→ Juxtapose



→ Partition



→ Superimpose



Overview + Detail

How?

Facet (multi-view)

The screenshot shows the datavoyager interface with the following components:

- Left Sidebar:**
 - Data:** Cars (Change)
 - Fields:** Cylinders, Name, Origin, Year, Acceleration, Displacement, Horsepower, Miles per Gallon, Weight in lbs, COUNT.
 - Wildcards:** Categorical Fields, Temporal Fields, Quantitative Fields.
- Encoding Panel:** x, y, column, row, size, color, shape, detail, text, any.
- Specified View:** No specified visualization yet. Start exploring by dragging a field to encoding pane on the left or examining univariate summaries below.
- Related Views:** Univariate Summaries (Hide)
- Univariate Summaries:**
 - A Cylinders # COUNT:** Horizontal bar chart showing counts for 3, 4, 5, 6, 8 cylinders.
 - A Name # COUNT:** Horizontal bar chart showing counts for various car models like amc ambassador, audi 100, etc.
 - A Origin # COUNT:** Horizontal bar chart showing counts for Europe, Japan, USA.
 - YEAR (Year) # COUNT:** Line chart showing counts for years from 1970 to 1982.
 - # BIN (Acceleration) # COUNT:** Histogram of acceleration bins.
 - # BIN (Displacement) # COUNT:** Histogram of displacement bins.
 - # BIN (Horsepower) # COUNT:** Histogram of horsepower bins.
 - # BIN (Miles per Gallon) # COUNT:** Histogram of miles per gallon bins.

<https://vega.github.io/voyager2/>

How to...

- ... **arrange** the data in the view spatially?
- ... **map** the data with all non-spatial channels?
- ... **manipulate** the data in the view?
- ... **facet** data between views?
- ... **reduce** the data in the view?

Reduce

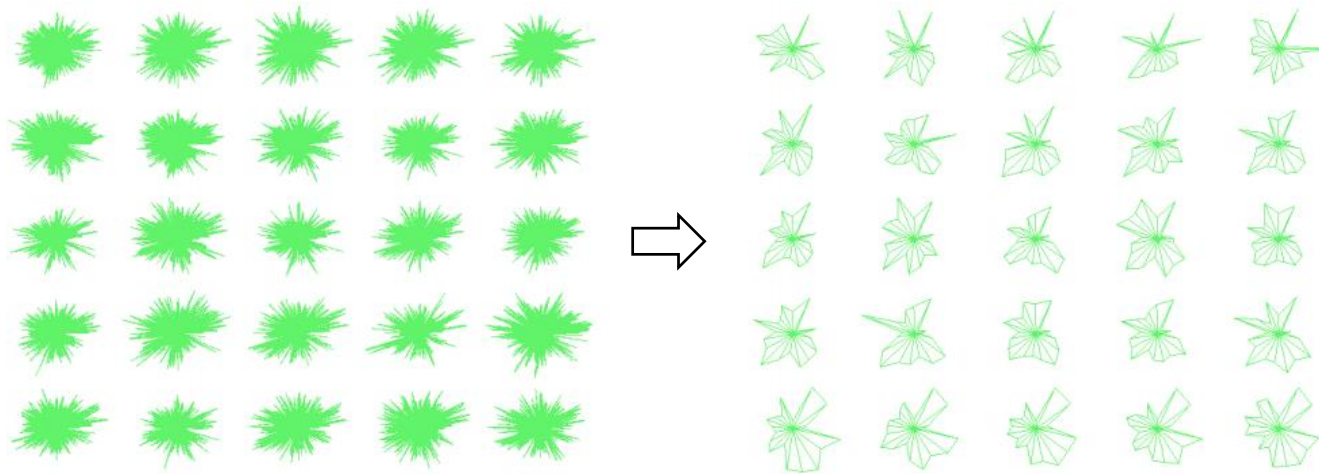
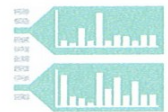
→ Filter



→ Aggregate



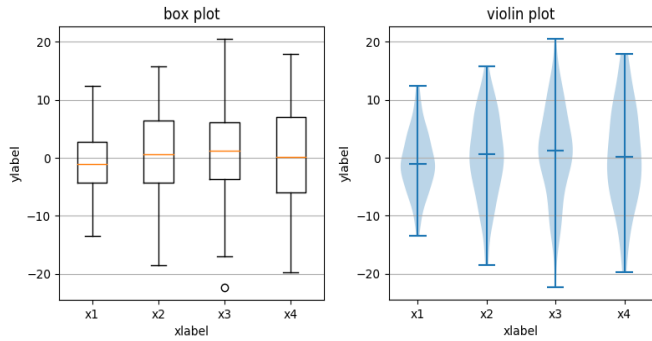
→ Embed



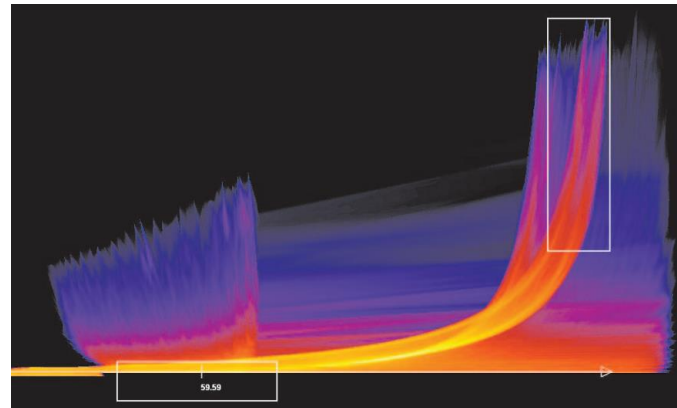
Yang, Jing, et al. "Interactive hierarchical dimension ordering, spacing and filtering for exploration of high dimensional datasets." *IEEE Symposium on Information Visualization, 2003*

How?

Reduce



<https://matplotlib.org/gallery.html>



S. Bachthaler and D. Weiskopf, "Continuous Scatterplots," in IEEE Transactions on Visualization and Computer Graphics, vol. 14, no. 6, pp. 1428-1435, 2008.

➔ Filter



➔ Aggregate



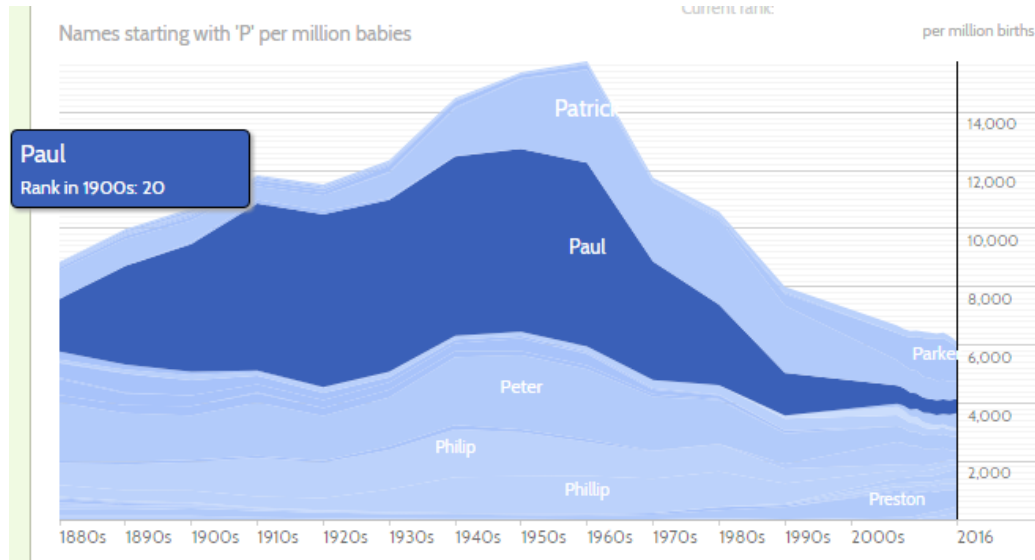
➔ Embed



S. Ingram, T. Munzner and M. Olano, "Glimmer: Multilevel MDS on the GPU," in IEEE Transactions on Visualization and Computer Graphics, vol. 15, no. 2, pp. 249-261, 2009.

How?

Reduce



<http://www.babynamewizard.com/voyager>

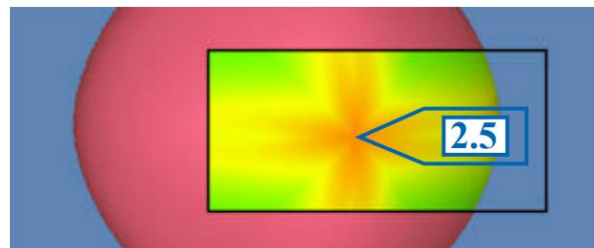
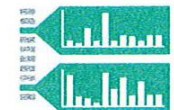
➔ Filter



➔ Aggregate

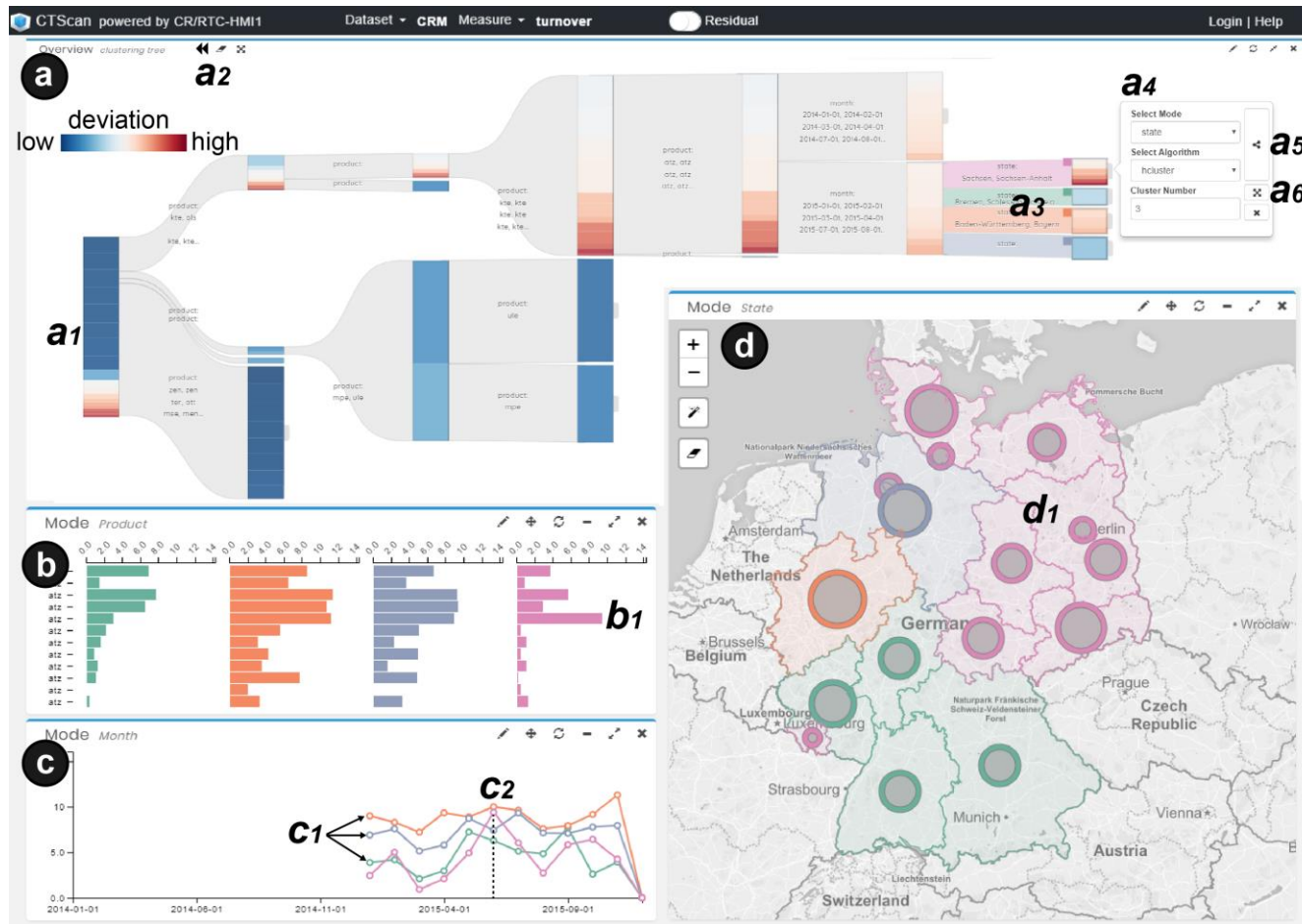


➔ Embed



Eric A. Bier, Maureen C. Stone, Ken Pier, William Buxton, and Tony D. DeRose. 1993. **Toolglass and magic lenses: the see-through interface.** ACM SIGGRAPH '93

Example: Progressive Partition and Multidimensional Pattern Extraction for Large-Scale Spatio-Temporal Data Analysis



Liu, Dongyu, Panpan Xu, and Liu Ren. "TPFlow: Progressive partition and multidimensional pattern extraction for large-scale spatio-temporal data analysis." IEEE transactions on visualization and computer graphics 25.1 (2018): 1-11.

Pattern Comparison

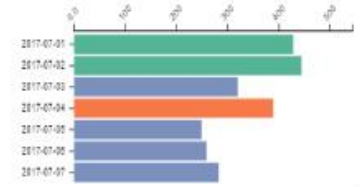
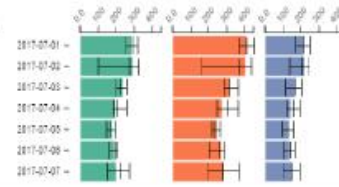
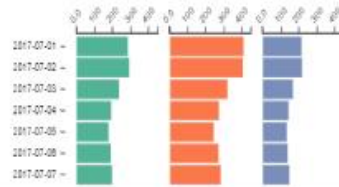
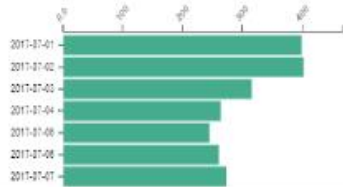
#1 Basic chart

#2 Enable comparison

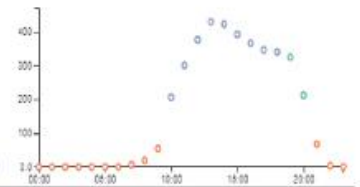
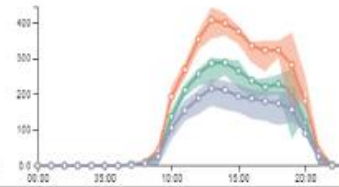
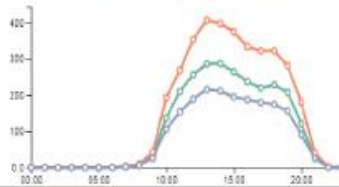
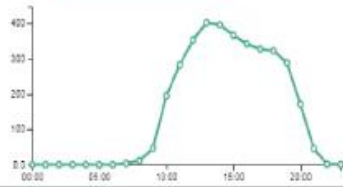
#3 Encode deviation

#4 Encode partition

#1 Temporal
Categorical
Numerical



#2 Temporal



#3 Spatial



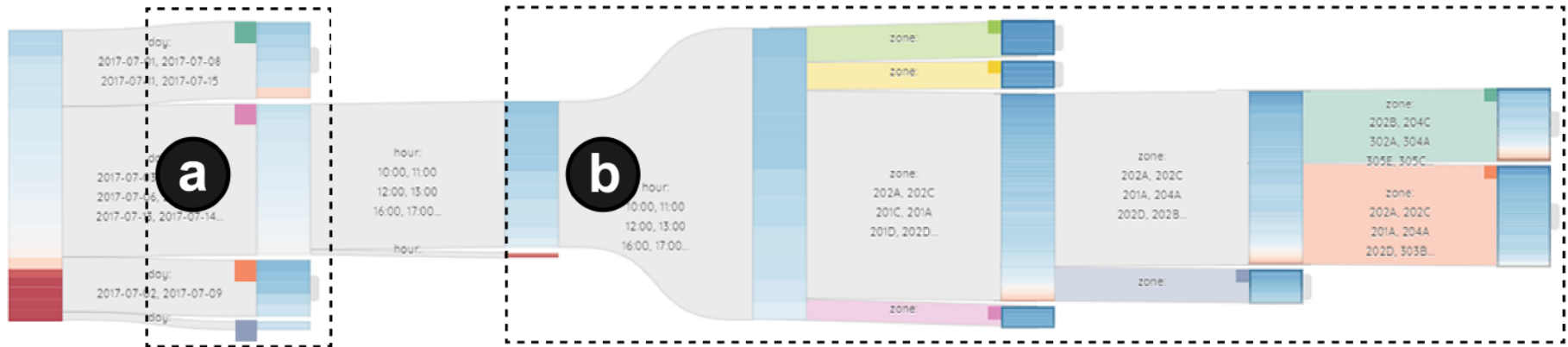
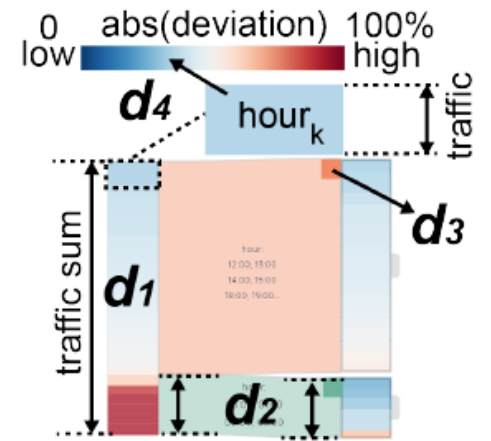
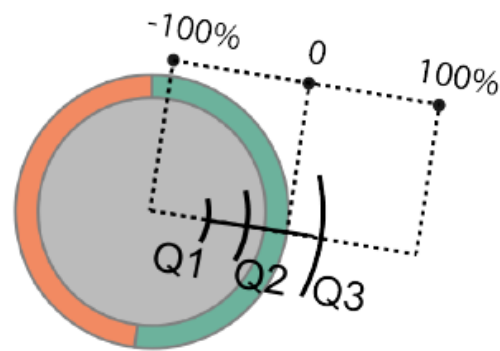
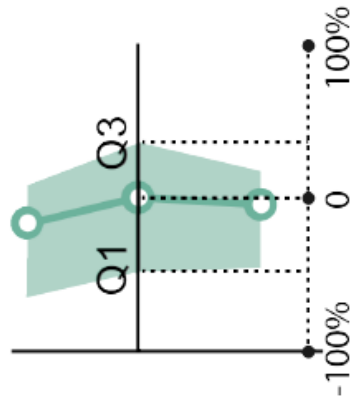
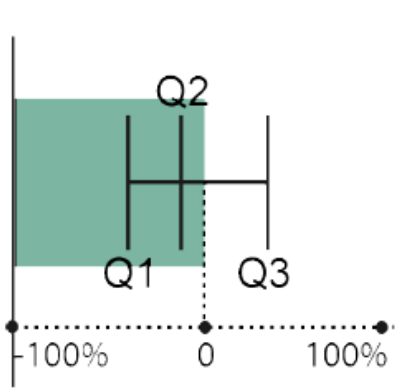
#4 OD



NULL



Data Encoding



TPFlow: Progressive Partition and Multidimensional Pattern Extraction for Large-Scale Spatio-Temporal Data Analysis

IEEE VAST 2018 - Conference on Visual Analytics Science and Technology
(TVCG Track*)

Dongyu Liu
Hong Kong University of
Science and Technology, Hong Kong

Panpan Xu Liu Ren
Bosch Research North America
Sunnyvale, CA

(With Audio)

*Paper selected to IEEE Transactions on Visualization and Computer Graphics (TVCG)



Why?

**Task
Abstraction**



How?

Design



**Data
Abstraction**



Visualization

Validation



Works?

What?

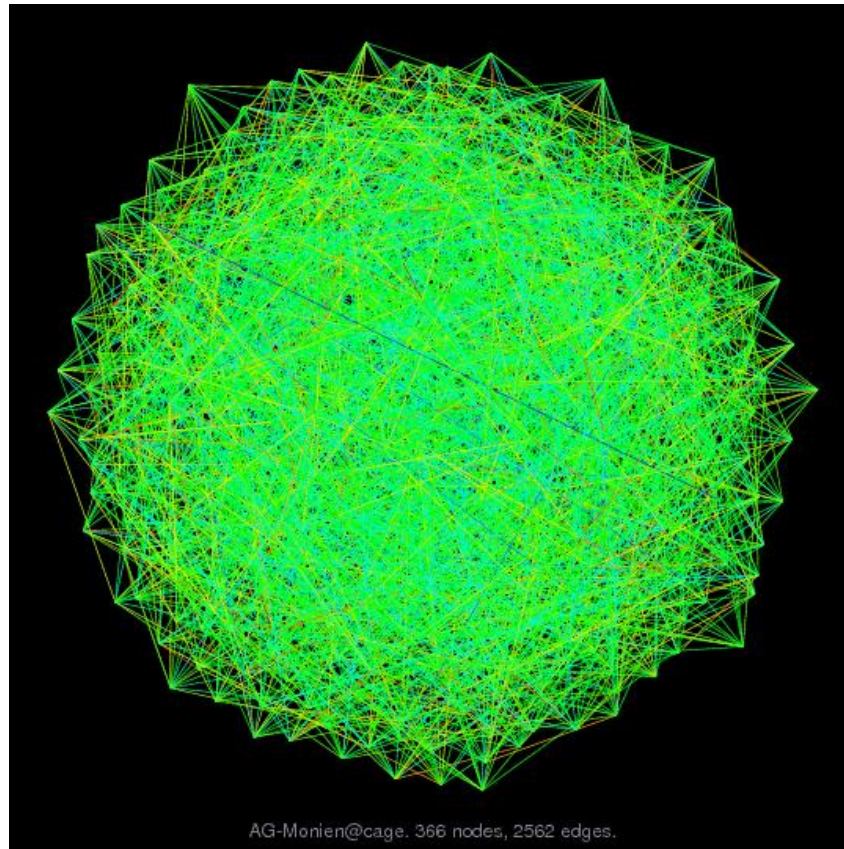
Validation

- How do you know if it works?
- How do you measure it?
 - How do you decide the benchmark data?
- How do you argue that one design is better or worse than another?
- What does it better mean?
 - Do user get something done faster?

Why it can go wrong?

- I do not understand
 - Poor match with the properties of the human perceptual and cognitive system
- Lack of context
 - Comprehensible by a human in some other setting
- Scalability problems

Why it can go wrong?



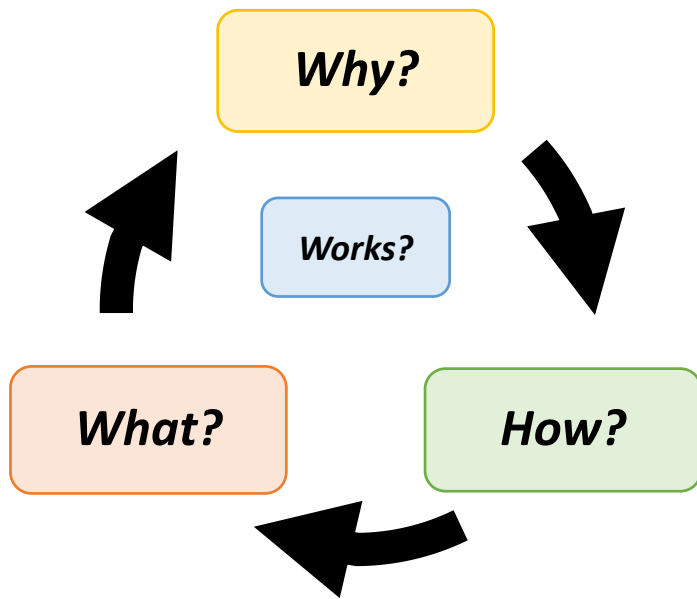
<http://yifanhu.net/GALLERY/GRAPHS>

Good Practices

- Do not optimize, satisfy
 - **The perfect visualization does not exist!!**
- Need to know a wide variety of vis techniques
 - Ease the design process using references
- Justification and Alternatives
 - 3D vs 2D?
- Eyes Beat Memory
 - Comparing views better than switch views
- Interactivity and Responsiveness

Life Cycle

- Follow the **life cycle** of a visualization tool
 - Profile of the user (expert, novice,...)



Prototype!

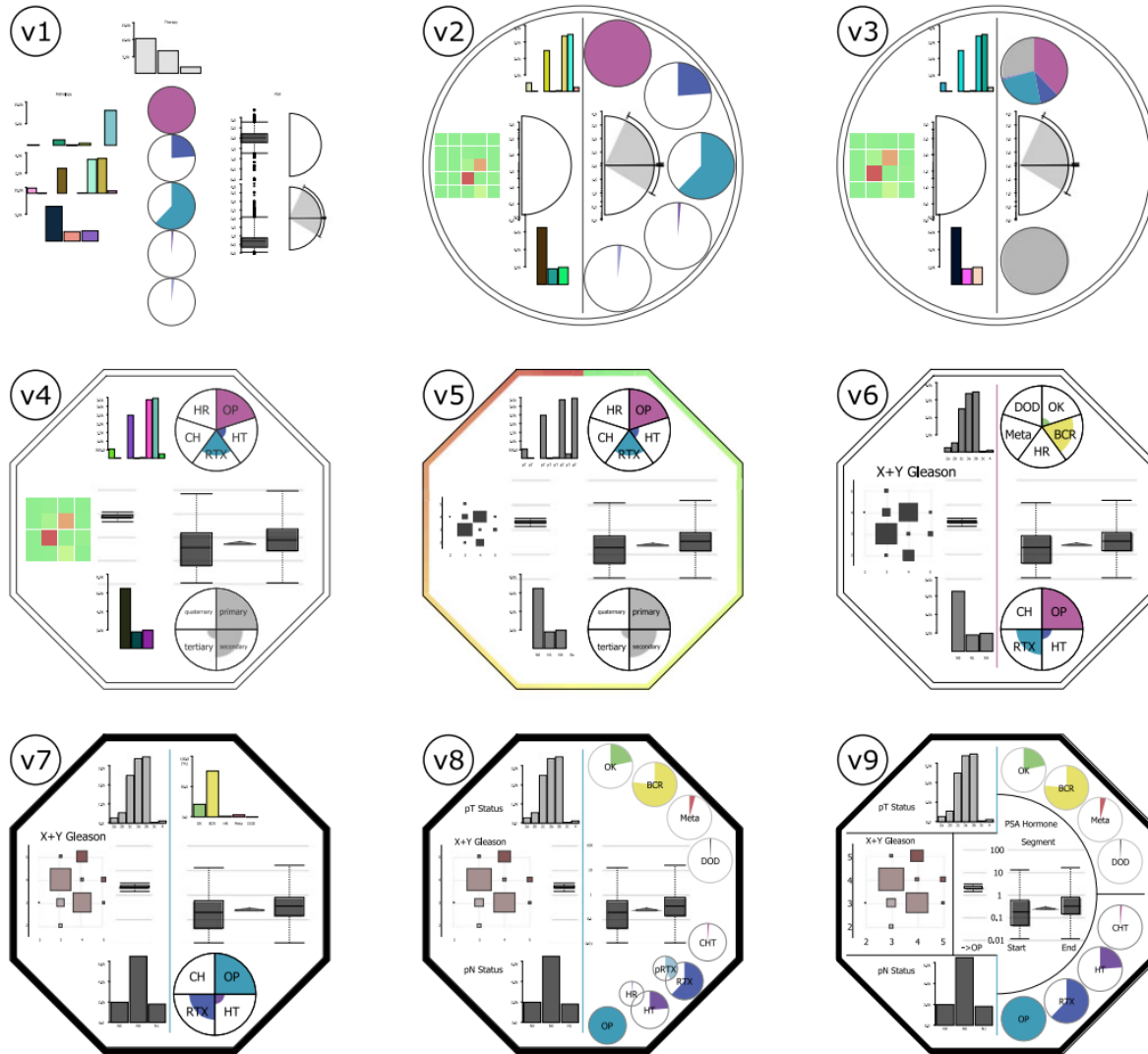


Trust the experts!

















Evaluate!

Example: Using Dashboard Networks to Visualize Multiple Patient Histories



Overview of 15 Most Relevant Attributes Describing the Condition of a Patient

Attribute	Description	Att. Type	Value Domain	Visual Enc.
PSA Pre-OP	The value of the PSA hormone is a most important biological indicator for prostate cancer. Last measurement of the PSA hormone before OP (surgery).	Numerical	$[0; 200.0]_{\frac{ng}{ml}}$ <i>median</i> : 8.7	
PSA at segment start	Value of the PSA hormone at start of the visualized temporal segment.	Numerical	$[0; 4,472.0]_{\frac{ng}{ml}}$ <i>median</i> : 0.21	
PSA at segment end	Value of the PSA hormone at the end of the visualized temporal segment. Used to assess the progression of the PSA value within the segment interval.	Numerical	$[0; 4,472.0]_{\frac{ng}{ml}}$ <i>median</i> : 0.26	
PSA trend in segment	Trend of the PSA hormone value within the duration of a temporal segment. Depicts the delta between the PSA values in a relative way.	Numerical	$] - 70.3; 4381.68]_{\frac{ng}{ml}}$ <i>median</i> : 0.0	
pT-status	Histopathological status of the tumor, obtained by pathology report (pT - pathological tumor). Used for prognoses about the progress of the disease.	Ordinal	<i>pt2a, pt2b, pt2c, pt3a, pt2b, pt3c, pt4</i>	
pN-status	Histopathological characteristics of the tumor, obtained by pathology report. Used for prognoses about the progress of the disease.	Ordinal	<i>NX, N0, N1</i>	
Gleason Score	Histopathological characteristics of the tumor, obtained by pathology report. Used for prognoses about the progress of the disease.	Ordinal	<i>2+3, 3+2, 3+3, 3+4, 3+5, 4+3, 4+4, 4+5, 5+3, 5+4</i>	
OK after OP	Outcome variable (biological condition) reflecting a good condition after OP. Is active until relapse or metastases.	Boolean	<i>[false, true]</i>	
BCR	Outcome variable (biological condition) describing the biochemical recurrence (relapse) of the tumor. Used for prognoses about the progress of the disease.	Boolean	<i>[false, true]</i>	
Metastases	Outcome variable (biological condition) describing whether a patient got metastases. Metastases are the most severe biological indicator in the dataset.	Boolean	<i>[false, true]</i>	
DOD	Biological end point when a patient died of disease. Often used for cause-effect analyses.	Boolean	<i>[false, true]</i>	
OP	Treatment attribute reflecting whether a patient had surgery. In the dataset all patients received OP.	Boolean	<i>[false, true]</i>	
RTX	Treatment attribute reflecting whether a patient got radiation therapy. Local treatment, only affecting the prostate area.	Boolean	<i>[false, true]</i>	
HT	Treatment attribute reflecting whether a patient had hormone therapy. Systemic treatment, affecting the whole human body.	Boolean	<i>[false, true]</i>	
CHT	Treatment attribute reflecting whether a patient got chemotherapy. Often applied in severe situations, e.g., when metastases are detected.	Boolean	<i>[false, true]</i>	

Immersive Analytics

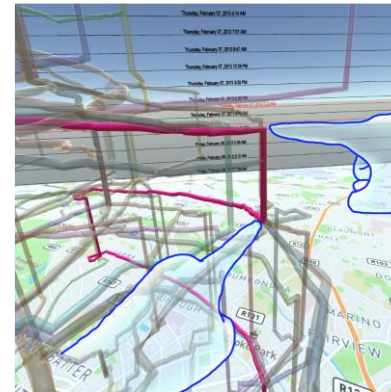
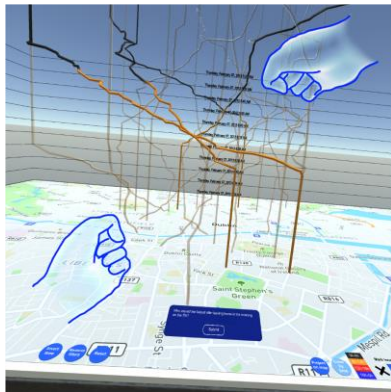
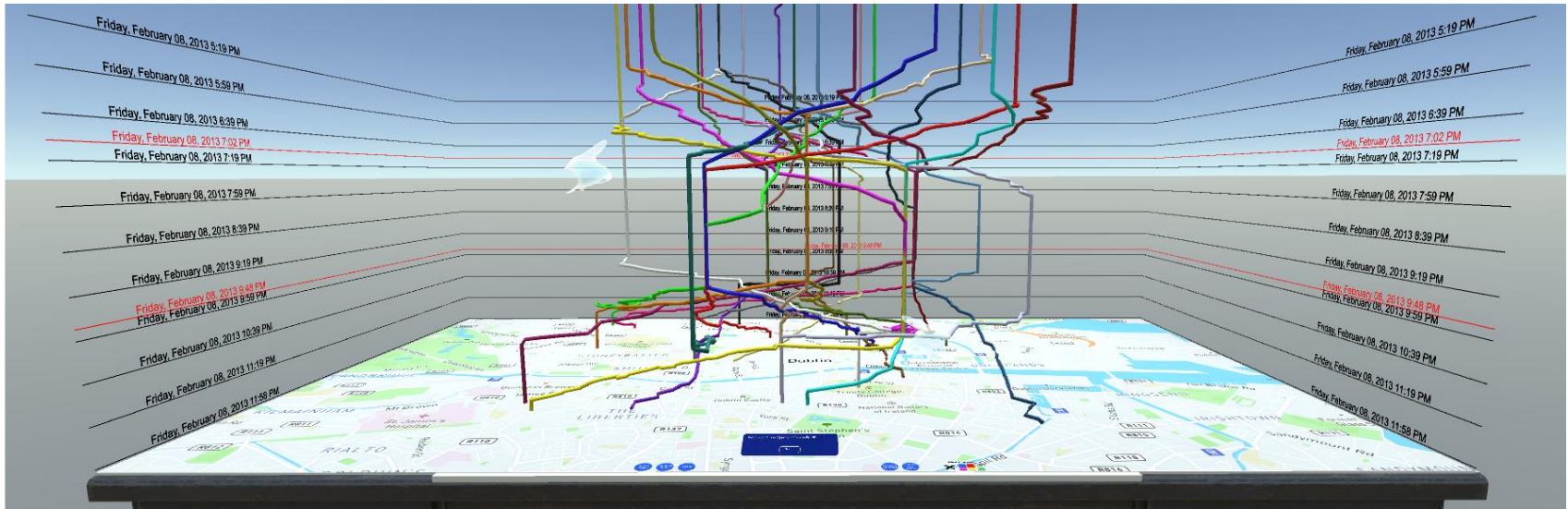
Focus

Immersive Analytics

“Science of analytical reasoning facilitated by immersive human-computer interfaces”



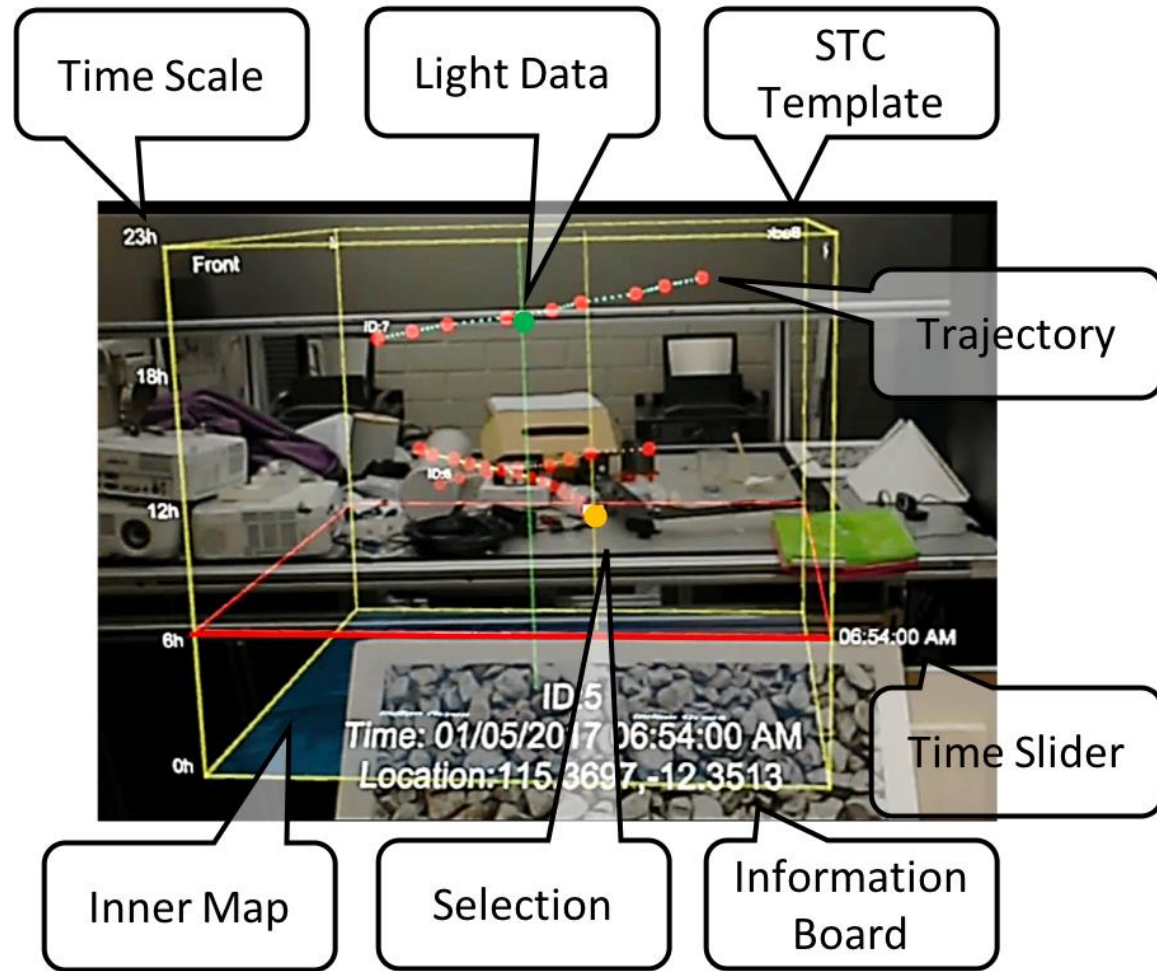
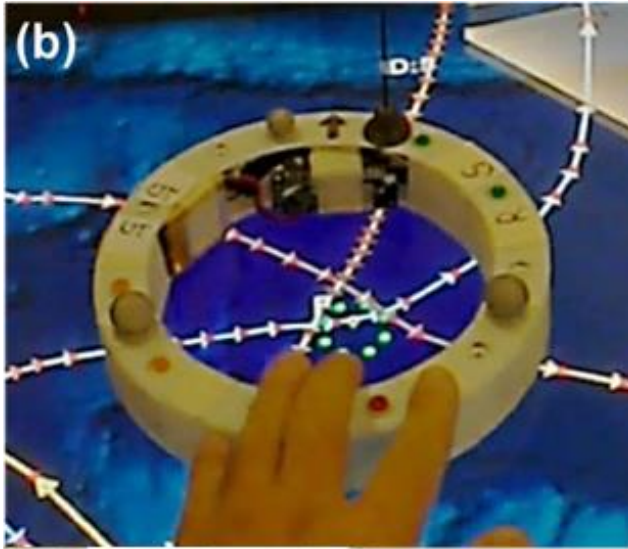
Space-Time Cube Geovisualization



Evaluating an Immersive Space-Time Cube Geovisualization for Intuitive Trajectory Data Exploration

Jorge Wagner, Wolfgang Stuerzlinger, Luciana Nedel

GeoGate



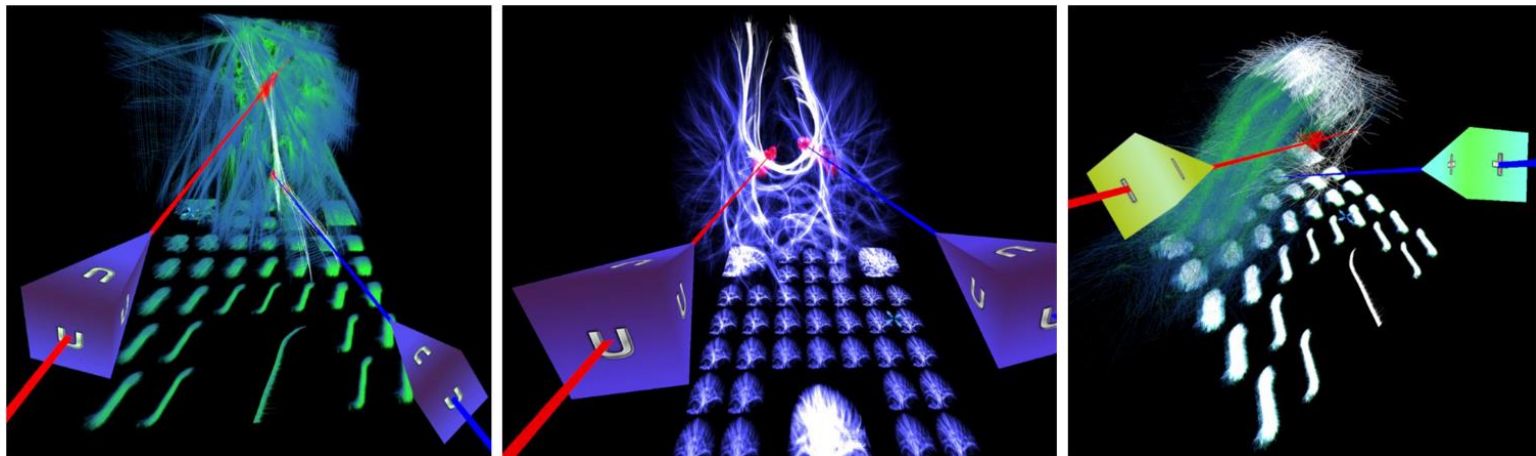
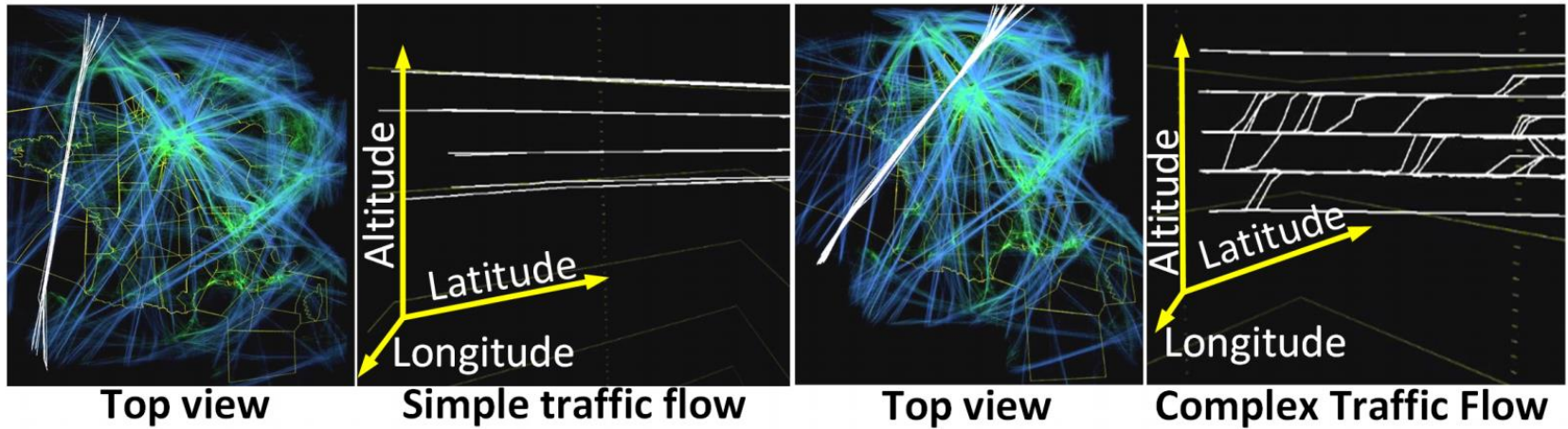
S. Y. Ssin, J. A. Walsh, R. T. Smith, A. Cunningham and B. H. Thomas, "GeoGate: Correlating Geo-Temporal Datasets Using an Augmented Reality Space-Time Cube and Tangible Interactions," *2019 IEEE Conference on Virtual Reality and 3D User Interfaces (VR)*, Osaka, Japan, 2019, pp. 210-219, doi: 10.1109/VR.2019.8797812.

Bento Box



Johnson S, Orban D, Runesha HB, Meng L, Juhnke B, Erdman A, Samsel F and Keefe DF (2019) Bento Box: An Interactive and Zoomable Small Multiples Technique for Visualizing 4D Simulation Ensembles in Virtual Reality. *Front. Robot. AI* 6:61. doi: 10.3389/frobt.2019.00061

FiberClay



FiberClay: Sculpting Three Dimensional Trajectories to Reveal Structural Insights

Christophe Hurter, Nathalie Henry Riche, Steven Drucker, Maxime Cordeil, Richard Alligier, Romain Vuillemot

Toolkits for Immersive Analytics

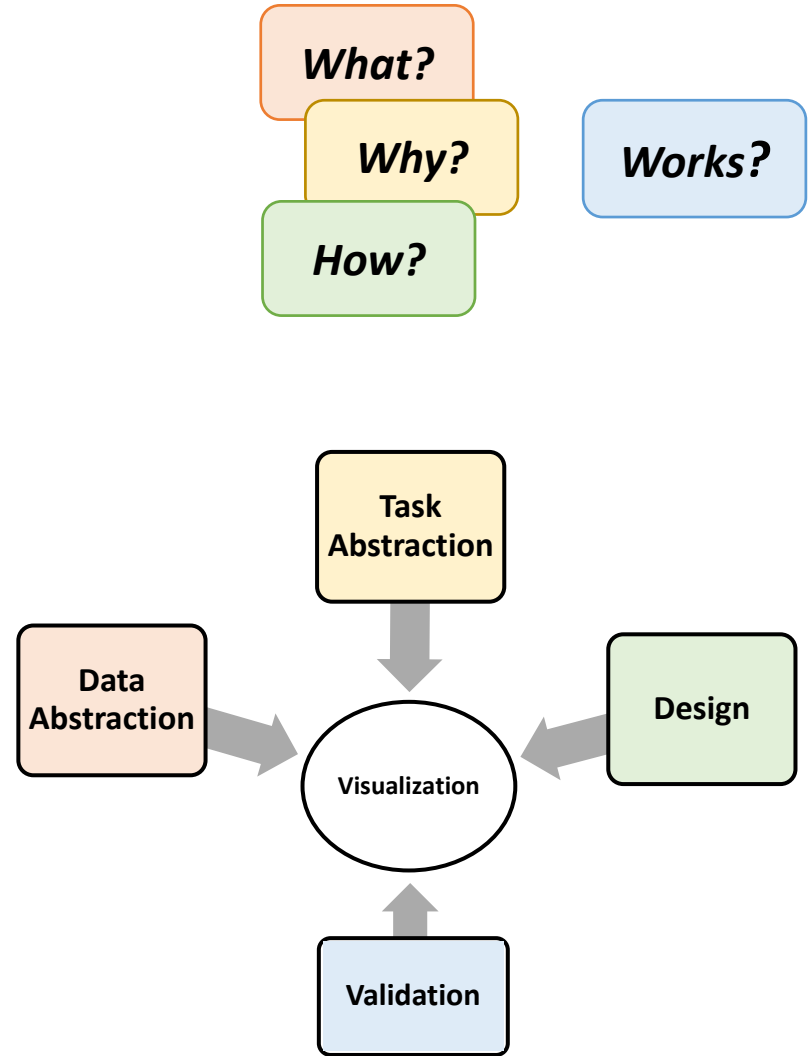
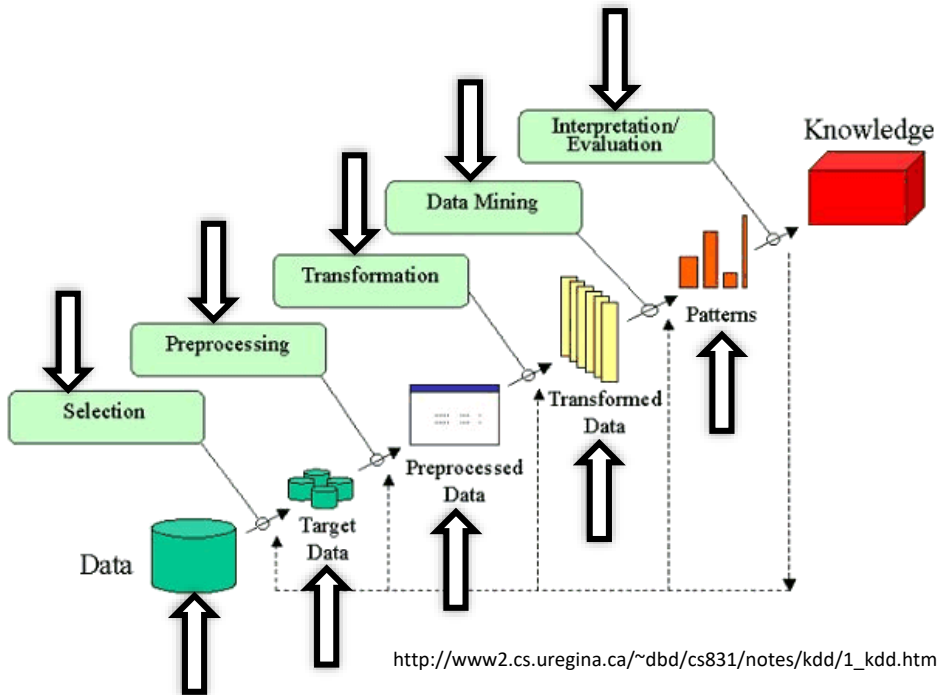
← GUI-based prototyping is easy to learn and use but is more templated → Grammar-based prototyping requires learning the grammar but is more flexible →



Research Challenges in Immersive Analytics

- Combining Human and Computer Intelligence
- The Utility of Immersion
- Designing Immersive Analytics Systems
- Facilitating Collaboration through Immersion
- Changing the Process of Analysis with Immersion

Wrap Up



Further reading

- Remy Dautriche. Multi-scale Interaction Techniques for the Interactive Visualization of Execution Traces. Chapter 3.3, Pattern Visualization, 2016
- T. Munzner. [Visualization](#). Chapter 27, p 675-707, of Fundamentals of Graphics, Third Edition, by Peter Shirley et al. AK Peters, 2009.
 - <http://www.cs.ubc.ca/labs/imager/tr/2009/VisChapter/akp-vischapter.pdf>
- T. Munzner. "**A Nested Model for Visualization Design and Validation**," in IEEE Transactions on Visualization and Computer Graphics, vol. 15, no. 6, pp. 921-928, Nov.-Dec. 2009.
 - <http://ieeexplore.ieee.org/abstract/document/5290695/>
- J. Heer, M. Bostock, and V. Ogievetsky. "**A tour through the visualization zoo**." Queue 8.5 (2010): 20.
 - <http://cacm.acm.org/magazines/2010/6/92482-a-tour-through-the-visualization-zoo/pdf>
- I. Herman, G. Melançon, and M. S. Marshall. "**Graph visualization and navigation in information visualization: A survey**." IEEE Transactions on visualization and computer graphics 6.1 (2000): 24-43.
 - <http://ieeexplore.ieee.org/abstract/document/841119/>
- F. Beck et al. **A taxonomy and survey of dynamic graph visualization**. Computer Graphics Forum. Vol. 36. No. 1. 2017.
 - <http://onlinelibrary.wiley.com/doi/10.1111/cgf.12791/full>

Further reading

- **Visualization Analysis and Design**
- **Design for Information: An Introduction to the Histories, Theories, and Best Practices Behind Effective Information Visualizations**
- **The Functional Art: An introduction to information graphics and visualization**

