

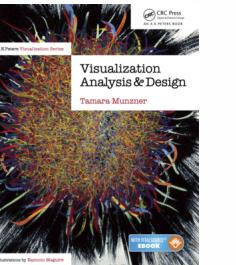
Visualization Analysis & Design

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<http://www.cs.ubc.ca/~tmm/talks.html#vad15seattle>



Defining visualization (vis)

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

Why?...

Why have a human in the loop?

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

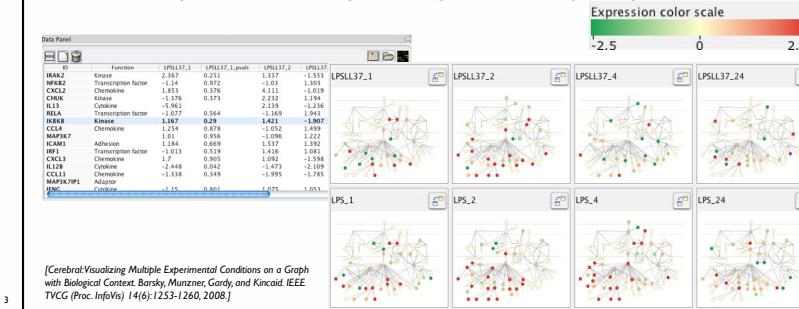
Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

- don't need vis when fully automatic solution exists and is trusted
- many analysis problems ill-specified
 - don't know exactly what questions to ask in advance
- possibilities
 - long-term use for end users (e.g. exploratory analysis of scientific data)
 - presentation of known results
 - stepping stone to better understanding of requirements before developing models
 - help developers of automatic solution refine/debug, determine parameters
 - help end users of automatic solutions verify, build trust

Why use an external representation?

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

- external representation: replace cognition with perception



[Cerebral Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Brehmer, Munzner, Gandy, and Kincaid. IEEE TVCG (Proc. InfoVis) 14(6):1233-1240, 2008.]

Why represent all the data?

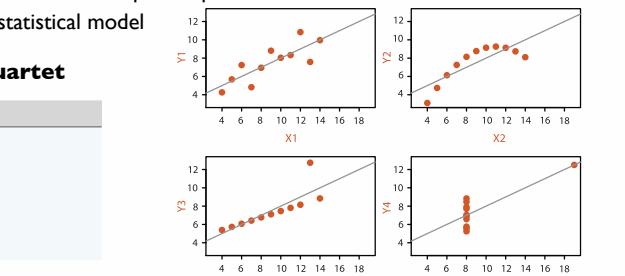
Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.

- summaries lose information, details matter
 - confirm expected and find unexpected patterns
 - assess validity of statistical model

Anscombe's Quartet

Identical statistics

x mean	9
x variance	10
y mean	8
y variance	4
x/y correlation	1



Why are there resource limitations?

Vis designers must take into account three very different kinds of resource limitations: those of computers, of humans, and of displays.

- computational limits
 - processing time
 - system memory
- human limits
 - human attention and memory
- display limits
 - pixels are precious resource, the most constrained resource
 - **information density**: ratio of space used to encode info vs unused whitespace
 - tradeoff between clutter and wasting space, find sweet spot between dense and sparse

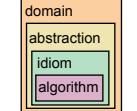
Why focus on tasks and effectiveness?

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

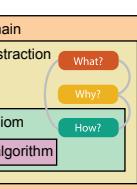
- what counts as effective?
 - novel: enable entirely new kinds of analysis
 - faster: speed up existing workflows
- most possibilities ineffective
 - increase chance of finding good solutions by understanding full space of possibilities
- tasks serve as constraint on design (as does data)
 - representations do not serve all tasks equally!

Analysis framework: Four levels, three questions

- **domain situation**
 - who are the target users?
- **abstraction**
 - translate from specifics of domain to vocabulary of vis
 - **what** is shown? **data abstraction**
 - **why** is the user looking at it? **task abstraction**
- **idiom**
 - **how** is it shown?
 - **visual encoding idiom**: how to draw
 - **interaction idiom**: how to manipulate
- **algorithm**
 - efficient computation



[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]



[A Multi-Level Typology of Abstract Visualization Tasks. Brehmer and Munzner. IEEE TVCG 19(12):2376-2385, 2013 (Proc. InfoVis 2013).]

Validation methods from different fields for each level



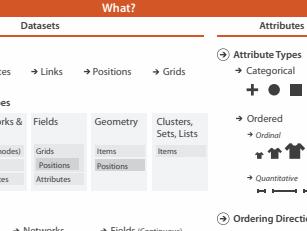
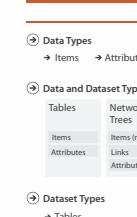
- mismatch: cannot show idiom good with system timings
- mismatch: cannot show abstraction good with lab study

Why analyze?

- imposes a structure on huge design space
 - scaffold to help you think systematically about choices
 - analyzing existing as stepping stone to designing new

SpaceTree

TreeJuxtaposer



What?

Attributes

Ordering Direction

Dataset Availability

Geometry

Grids

Items

Links

Positions

Tables

Clusters, Sets, Lists

Items

Attributes

Categorical

Ordinal

Quantitative

Sequential

Diverging

Cyclic

Static

Dynamic

Position

Link

Node

Cell

Value in cell

Grid of positions

Fields (Continuous)

Geometry (Spatial)

Spatial

Dataset and data types

Dataset Types

Networks

Fields (Spatial)

Geometry (Spatial)

Spatial

Attribute Types

Ordered

Ordinal

Quantitative

Position

Actions I: Analyze

- consume

Search

Summarize

- discover vs present

Compare

Identify

- aka explore vs explain

Derive

Compare

- enjoy

Explore

Summarize

- newcomer

Compare

Identify

- aka casual, social

Compare

Identify

- produce

Compare

Identify

- annotate, record

Compare

Identify

- derive

Compare

Identify

- crucial design choice

Compare

Identify

Actions II: Search

- what does user know?

Search

Summarize

- target, location

Compare

Identify

- target known

Compare

Identify

- target unknown

Compare

Identify

- location known

Compare

Identify

- location unknown

Compare

Identify

Actions III: Query

- what does user know?

Search

Summarize

- target, location

Compare

Identify

- how much of the data matters?

Compare

Identify

- one, some, all

Compare

Identify

Actions IV: Produce

Produce

Targets

- All Data
 - Trends → Outliers → Features
- Attributes
 - One → Many → Dependency → Correlation → Similarity
 - Distribution → Extremes
- Network Data
 - Topology
 - Paths
- Spatial Data
 - Shape

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How?

Encode	Manipulate	Facet	Reduce
<ul style="list-style-type: none"> Arrange → Express → Separate Order → Align Use → 	<ul style="list-style-type: none"> Map from categorical and ordered attributes → Color → Hue → Saturation → Luminance Size, Angle, Curvature, ... Shape → + ● □ ▲ Motion → Direction, Rate, Frequency, ... 	<ul style="list-style-type: none"> Change → Juxtapose → Filter Select → Partition → Aggregate Navigate → Superimpose → Embed 	<ul style="list-style-type: none"> Facet Reduce

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What?
Why?
How?

How to encode: Arrange space, map channels

Encode
<ul style="list-style-type: none"> Arrange → Express → Separate Order → Align Use →

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Encoding visually

- analyze idiom structure

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Definitions: Marks and channels

- marks
 - geometric primitives
 - Points, Lines, Areas
- channels
 - control appearance of marks
 - Position (Horizontal, Vertical, Both), Color, Shape, Tilt, Size (Length, Area, Volume)

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Encoding visually with marks and channels

- analyze idiom structure
 - as combination of marks and channels

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1: vertical position
2: vertical position horizontal position
3: vertical position horizontal position color hue
4: vertical position horizontal position color hue size (area)

mark: line mark: point mark: point mark: point

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Channels

Position on common scale	Spatial region
Position on unaligned scale	Color hue
Length (1D size)	Motion
Tilt/angle	Shape
Area (2D size)	
Depth (3D position)	
Color luminance	
Color saturation	
Curvature	
Volume (3D size)	

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Channels: Rankings

	Effectiveness
Magnitude Channels: Ordered Attributes	Best
Position on common scale	Best
Position on unaligned scale	Good
Length (1D size)	Medium
Tilt/angle	Medium
Area (2D size)	Medium
Depth (3D position)	Medium
Color luminance	Medium
Color saturation	Medium
Curvature	Medium
Volume (3D size)	Least

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Identity Channels: Categorical Attributes

- expressiveness principle
 - match channel and data characteristics
- effectiveness principle
 - encode most important attributes with highest ranked channels

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How?

Encode	Manipulate	Facet	Reduce
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What?
Why?
How?

How to handle complexity: 3 more strategies + 1 previous

Manipulate	Facet	Reduce	→ Derive
<ul style="list-style-type: none"> Change Select Navigate 	<ul style="list-style-type: none"> Juxtapose Partition Superimpose 	<ul style="list-style-type: none"> Filter Aggregate Embed 	<ul style="list-style-type: none"> Derive

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• change view over time
• facet across multiple views
• reduce items/attributes within single view
• derive new data to show within view

How to handle complexity: 3 more strategies + 1 previous

Manipulate	Facet	Reduce	→ Derive
<ul style="list-style-type: none"> Change Select Navigate 	<ul style="list-style-type: none"> Juxtapose Partition Superimpose 	<ul style="list-style-type: none"> Filter Aggregate Embed 	<ul style="list-style-type: none"> Derive

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• change over time
- most obvious & flexible of the 4 strategies

Facet

- Juxtapose
- Partition
- Superimpose

Coordinate Multiple Side By Side Views

- Share Encoding: Same/Different → Linked Highlighting
- Share Data: All/Subset/None
- Share Navigation

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How to handle complexity: 3 more strategies + 1 previous

Manipulate	Facet	Reduce	→ Derive
<ul style="list-style-type: none"> Change Select Navigate 	<ul style="list-style-type: none"> Juxtapose Partition Superimpose 	<ul style="list-style-type: none"> Filter Aggregate Embed 	<ul style="list-style-type: none"> Derive

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• facet data across multiple views

Idiom: Linked highlighting

- see how regions contiguous in one view are distributed within another
- powerful and pervasive interaction idiom
- encoding: different - **multiform**
- data: all shared

System: EDV

[Visual Exploration of Large Structured Datasets, Wills, Proc. New Techniques and Trends in Statistics (NTTS), pp. 237–246. IOS Press, 1995.]

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Idiom: bird's-eye maps

• encoding: same
• data: subset shared
• navigation: shared
– bidirectional linking

• differences
– viewpoint
– (size)

• **overview-detail**

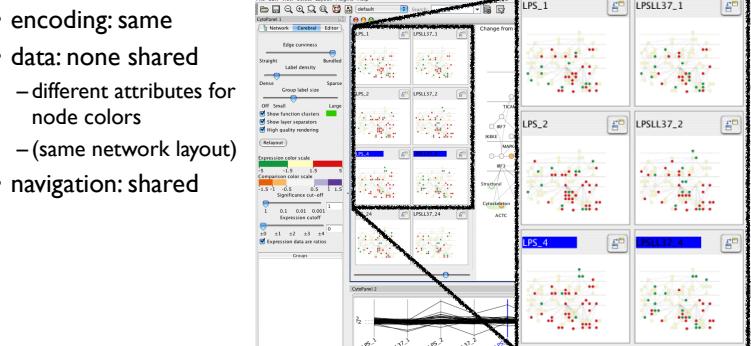


[A Review of Overview+Detail, Zooming, and Focus+Context Interfaces. Cockburn, Karlson, and Bederson. ACM Computing Surveys 41:1 (2008), 1-31.]

System: Google Maps

Idiom: Small multiples

• encoding: same
• data: none shared
– different attributes for node colors
– (same network layout)
• navigation: shared



[Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gandy, and Kincaid. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2008) 14:6 (2008), 1253-1260.]

System: Cerebral

Coordinate views: Design choice interaction

		Data		
		All	Subset	None
Encoding	Same	Redundant	Overview/ Detail	Small Multiples
	Different	Multiform	Multiform, Overview/ Detail	No Linkage

why juxtapose views?
– benefits: eyes vs memory
• lower cognitive load to move eyes between 2 views than remembering previous state with single changing view
– costs: display area, 2 views side by side each have only half the area of one view

Partition into views

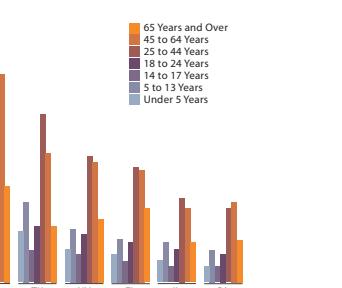
- how to divide data between views
 - encodes association between items using spatial proximity
 - major implications for what patterns are visible
 - split according to attributes
- design choices
 - how many splits
 - all the way down: one mark per region?
 - stop earlier, for more complex structure within region?
 - order in which attrs used to split
 - how many views

Partition into Side-by-Side Views



Partitioning: List alignment

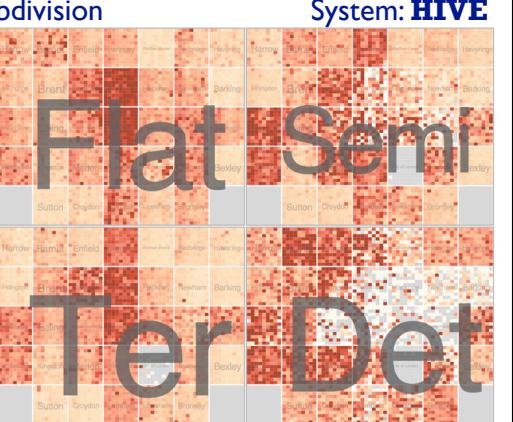
- single bar chart with grouped bars
 - split by state into regions
 - complex glyph within each region showing all ages
 - compare: easy within state, hard across ages



[Configuring Hierarchical Layouts to Address Research Questions. Slingsby, Dykes, and Wood. IEEE Transactions on Visualization and Computer Graphics (Proc. InfoVis 2009) 15:6 (2009), 977-984.]

Partitioning: Recursive subdivision

• split by type
• then by neighborhood
• then time
– years as rows
– months as columns

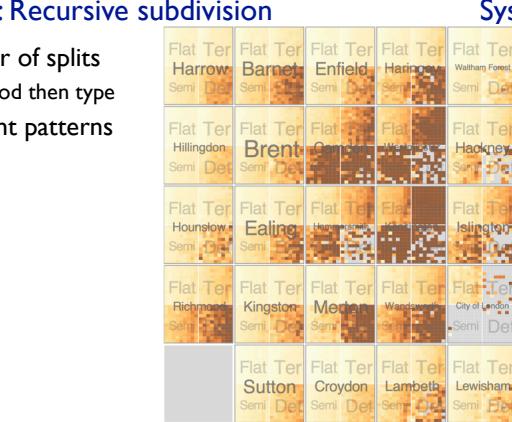


[Configuring Hierarchical Layouts to Address Research Questions. Slingsby, Dykes, and Wood. IEEE Transactions on Visualization and Computer Graphics (Proc. InfoVis 2009) 15:6 (2009), 977-984.]

System: HIVE

Partitioning: Recursive subdivision

• switch order of splits
– neighborhood then type
• very different patterns

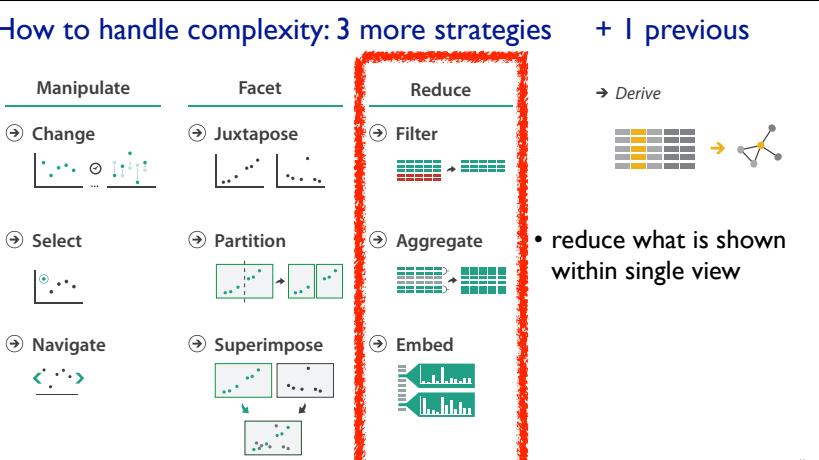


[Configuring Hierarchical Layouts to Address Research Questions. Slingsby, Dykes, and Wood. IEEE Transactions on Visualization and Computer Graphics (Proc. InfoVis 2009) 15:6 (2009), 977-984.]

System: HIVE

How to handle complexity: 3 more strategies

+ 1 previous



Manipulate
Facet
Reduce
Juxtapose
Filter
Partition
Superimpose
Embed
Select
Aggregate
Embed
Navigate

• reduce what is shown within single view

Reduce items and attributes

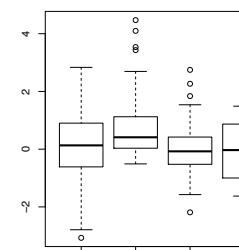
- reduce/increase: inverses
- filter
 - pro: straightforward and intuitive
 - to understand and compute
 - con: out of sight, out of mind
- aggregation
 - pro: inform about whole set
 - con: difficult to avoid losing signal
- not mutually exclusive
 - combine filter, aggregate
 - combine reduce, facet, change, derive

Reducing Items and Attributes

- Filter
 - Items
 - Attributes
- Aggregate
 - Items
 - Attributes
- Reduce
 - Filter
 - Aggregate
 - Embed

Idiom: boxplot

- static item aggregation
- task: find distribution
- data: table
- derived data
 - 5 quant attrs
 - median: central line
 - lower and upper quartile: boxes
 - lower upper fences: whiskers
 - values beyond which items are outliers
 - outliers beyond fence cutoffs explicitly shown



[40 years of boxplots. Wickham and Stryjewski. 2012. had.co.nz]

Idiom: Dimensionality reduction for documents

- attribute aggregation
 - derive low-dimensional target space from high-dimensional measured space

Task 1: Item 1, ..., Item n → In HD data → Out 2D data

Task 2: Item 1, ..., Item n → In 2D data → Out Scatterplot Clusters & points

Task 3: Item 1, ..., Item n → In Scatterplot Clusters & points → Out Labels for clusters

What? Why? How?

In HD data → In 2D data → In Scatterplot Clusters & points → Out Labels for clusters

More Information

- this talk <http://www.cs.ubc.ca/~tmm/talks.html#vad15seattle>
- book page (including tutorial lecture slides) <http://www.cs.ubc.ca/~tmm/vadbook>
 - 20% promo code for book+ebook combo: HVN17
 - <http://www.crcpress.com/product/isbn/9781466508910>
- illustrations: Eamonn Maguire
- papers, videos, software, talks, full courses <http://www.cs.ubc.ca/group/infovis> <http://www.cs.ubc.ca/~tmm>



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Tamara Munzner
A K Peters Visualization Series, CRC Press, Visualization Series, 2014.