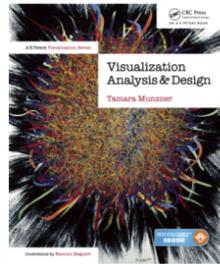


Visualization Analysis & Design

Full-Day Tutorial

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 ACT
 August 2016, Iowa City IA



<http://www.cs.ubc.ca/~tmm/talks.html#vad16act>

@tamaramunzner

Outline

- Session 1 8:30-10:00am**
Visualization Analysis Framework
 - Introduction: Definitions
 - Analysis: What, Why, How
 - Marks and Channels
- Session 2 10:30am-12:00pm**
Spatial Layout
 - Arrange Tables
 - Arrange Spatial Data
 - Arrange Networks and Trees
- Session 3 1:00-2:30pm**
Color & Interaction
 - Map Color
 - Manipulate: Change, Select, Navigate
 - Facet: Juxtapose, Partition, Superimpose
- Session 4 3:00-4:30pm**
Guidelines and Examples
 - Reduce: Filter, Aggregate
 - Rules of Thumb
 - Q&A

<http://www.cs.ubc.ca/~tmm/talks.html#vad16act>

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Defining visualization (vis)

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

Why?...

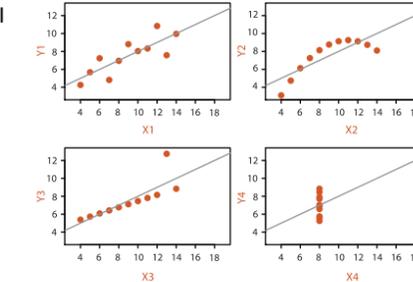
Why represent all the data?

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	7.5
y variance	3.75
x/y correlation	0.816



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.

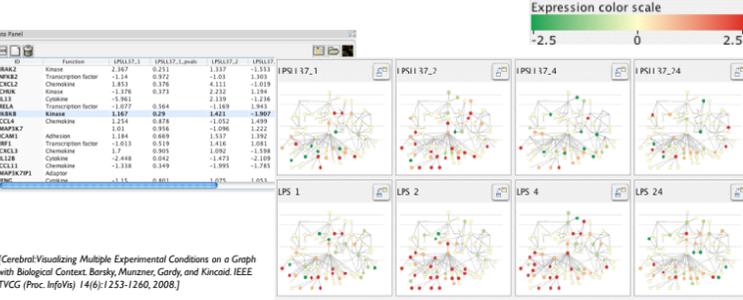
Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

- 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



[Cerebra] Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Bersky, Munzner, Gandy, and Kincaid. IEEE TVCG (Proc. InfoVis) 14(6):1253-1260, 2008.

Why depend on vision?

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

- human visual system is high-bandwidth channel to brain
 - overview possible due to background processing
 - subjective experience of seeing everything simultaneously
 - significant processing occurs in parallel and pre-attentively
- sound: lower bandwidth and different semantics
 - overview not supported
 - subjective experience of sequential stream
- touch/haptics: impoverished record/replay capacity
 - only very low-bandwidth communication thus far
- taste, smell: no viable record/replay devices

Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
 - Chap 1: What's Vis, and Why Do It?

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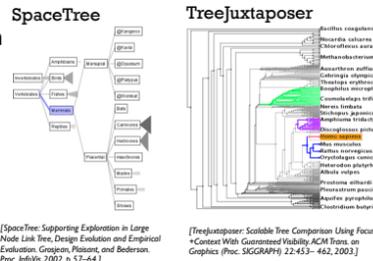
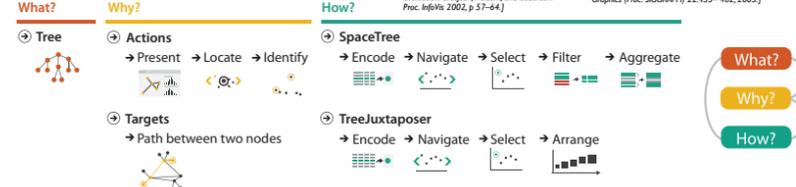
What resource limitations are we faced with?

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 analyze?

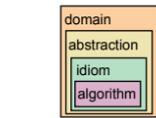
- imposes structure on huge design space
 - scaffold to help you think systematically about choices
 - analyzing existing as stepping stone to designing new
 - most possibilities ineffective for particular task/data combination



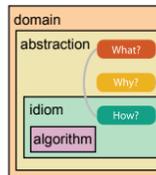
[SpaceTree: Supporting Exploration in Large Node Link Trees, Design Evaluation and Empirical Evaluation. Gandy, Munzner, and Bederson. Proc. InfoVis 2002, p. 57-64.]
 [TreeJuxtaposer: Scalable Tree Comparison Using Focus + Context With Guaranteed Visibility. ACM Trans. on Graphics (Proc. SIGGRAPH) 22:453-462, 2003.]

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**
 - often don't just draw what you're given: transform to new form
- 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).]

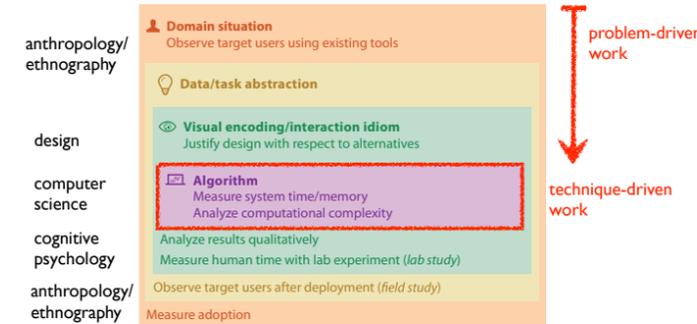
Why is validation difficult?

- different ways to get it wrong at each level

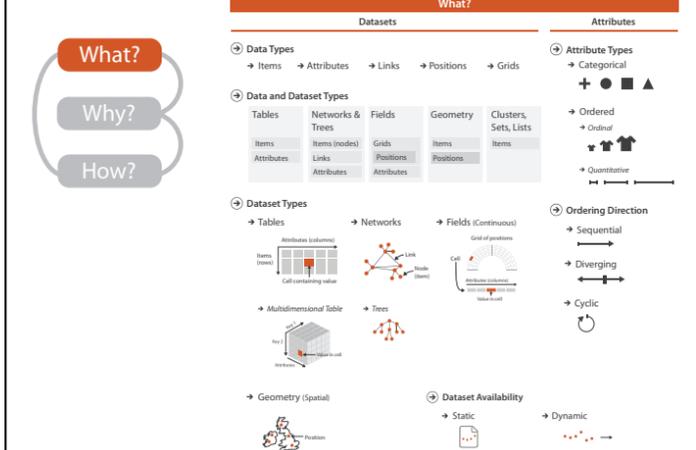


Why is validation difficult?

- solution: use methods from different fields at each level



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



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

Three major datatypes

Dataset Types

- Tables**
 - Attributes (columns)
 - Items (rows)
 - Cell containing value
 - Multidimensional Table
- Networks**
 - Link
 - Node (Item)
 - Trees
- Spatial**
 - Fields (Continuous)
 - Geometry (Spatial)
 - Grid of positions
 - Cell
 - Attributes (columns)
 - Value in cell
 - Position

• visualization vs computer graphics
– geometry is design decision

Dataset and data types

Data and Dataset Types

Tables	Networks & Trees	Fields	Geometry	Clusters, Sets, Lists
Items	Items (nodes)	Grids	Items	Items
Attributes	Links	Positions	Positions	
	Attributes	Attributes		

Data Types

- Items
- Attributes
- Links
- Positions
- Grids

Dataset Availability

- Static
- Dynamic

Attribute types

Attribute Types

- Categorical: +, ●, ■, ▲
- Ordered: → Ordinal, → Quantitative

Ordering Direction

- Sequential: →
- Diverging: ↔
- Cyclic: ↻

Why?

What? Why? How?

Actions

- Analyze: Consume (Discover, Present, Enjoy), Produce (Annotate, Record, Derive)
- Search: Location known (Lookup, Browse), Location unknown (Locate, Explore)
- Query: Identify, Compare, Summarize

Targets

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

• {action, target} pairs

- discover distribution
- compare trends
- locate outliers
- browse topology

Actions: Analyze

- consume
 - discover vs present
 - classic split
 - aka explore vs explain
 - enjoy
 - newcomer
 - aka casual, social
- produce
 - annotate, record
 - derive
 - crucial design choice

Derive

- don't just draw what you're given!
- decide what the right thing to show is
- create it with a series of transformations from the original dataset
- draw that
- one of the four major strategies for handling complexity

Original Data: exports, imports
Derived Data: trade balance = exports - imports

Actions: Search, query

- what does user know? Search
 - target, location
- how much of the data matters?
 - one, some, all
- independent choices for each of these three levels
 - analyze, search, query
 - mix and match

	Target known	Target unknown
Location known	Lookup	Browse
Location unknown	Locate	Explore

Query: Identify, Compare, Summarize

Analysis example: Derive one attribute

- Strahler number
 - centrality metric for trees/networks
 - derived quantitative attribute
 - draw top 5K of 500K for good skeleton

[Using Strahler numbers for real time visual exploration of huge graphs. Auber. Proc. Intl. Conf. Computer Vision and Graphics, pp. 56-69, 2002.]

Why: Targets

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

How?

Encode	Manipulate	Facet	Reduce
<ul style="list-style-type: none"> Arrange: Express, Separate, Order, Use Map: Color (Hue, Saturation, Luminance), Size, Angle, Curvature, Shape, Motion 	<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

What? Why? How?

Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
 - Chap 2: What: Data Abstraction
 - Chap 3: Why: Task Abstraction
- A Multi-Level Typology of Abstract Visualization Tasks. Brehmer and Munzner. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis) 19:12 (2013), 2376–2385.
- Low-Level Components of Analytic Activity in Information Visualization. Amar, Eagan, and Stasko. Proc. IEEE InfoVis 2005, p 111–117.
- A taxonomy of tools that support the fluent and flexible use of visualizations. Heer and Shneiderman. Communications of the ACM 55:4 (2012), 45–54.
- Rethinking Visualization: A High-Level Taxonomy. Tory and Möller. Proc. IEEE InfoVis 2004, p 151–158.
- Visualization of Time-Oriented Data. Aigner, Miksch, Schumann, and Tominski. Springer, 2011.

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Visual encoding

- analyze idiom structure

Definitions: Marks and channels

- marks
 - geometric primitives
- channels
 - control appearance of marks
 - can redundantly code with multiple channels

Position	Color
Shape	Tilt
Size	

Visual encoding

- analyze idiom structure
- as combination of marks and channels

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

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	Same
Color saturation	
Curvature	Same
Volume (3D size)	

Channels: Matching Types

Magnitude Channels: Ordered Attributes

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

Identity Channels: Categorical Attributes

- Spatial region
- Color hue
- Motion
- Shape

• expressiveness principle
– match channel and data characteristics

Channels: Rankings

Magnitude Channels: Ordered Attributes

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

Identity Channels: Categorical Attributes

- Spatial region
- Color hue
- Motion
- Shape

• expressiveness principle
– match channel and data characteristics

• effectiveness principle
– encode most important attributes with highest ranked channels

Channels: Expressiveness types and effectiveness rankings

Magnitude Channels: Ordered Attributes

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

Identity Channels: Categorical Attributes

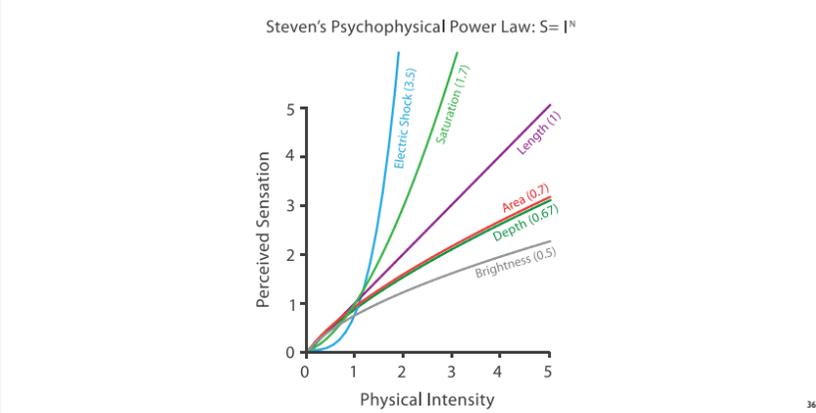
- Spatial region
- Color hue
- Motion
- Shape

• expressiveness principle
– match channel and data characteristics

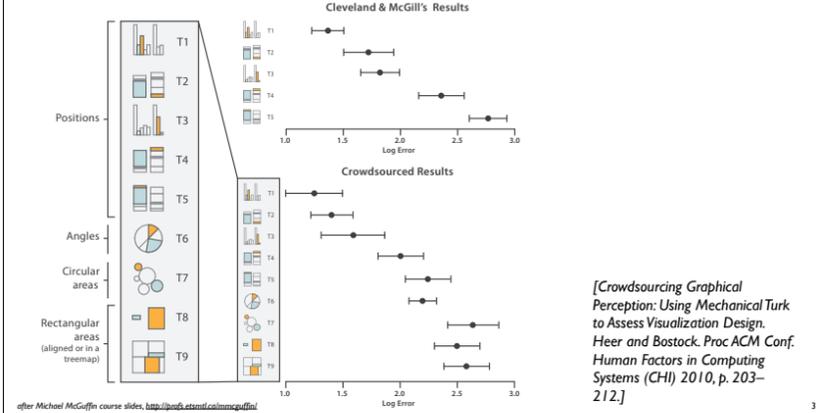
• effectiveness principle
– encode most important attributes with highest ranked channels

– spatial position ranks high for both

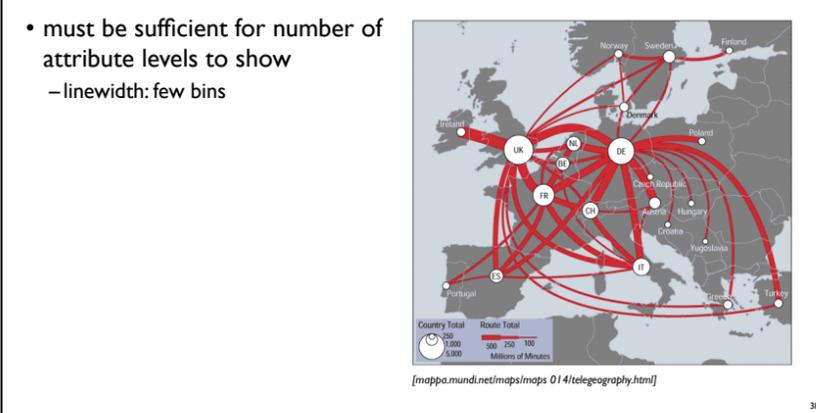
Accuracy: Fundamental Theory



Accuracy: Vis experiments



Discriminability: How many usable steps?



Separability vs. Integrality

Position + Hue (Color)

Size + Hue (Color)

Width + Height

Red + Green

Fully separable Some interference Some/significant interference Major interference

2 groups each 2 groups each 3 groups total: integral area 4 groups total: integral hue

Popout

• find the red dot
– how long does it take?

• parallel processing on many individual channels
– speed independent of distractor count
– speed depends on channel and amount of difference from distractors

• serial search for (almost all) combinations
– speed depends on number of distractors

Popout

• many channels: tilt, size, shape, proximity, shadow direction, ...
• but not all! parallel line pairs do not pop out from tilted pairs

Grouping

Marks as Links

- Containment
- Connection

Identity Channels: Categorical Attributes

- Spatial region
- Color hue
- Motion
- Shape

• containment
• connection

• proximity
– same spatial region

• similarity
– same values as other categorical channels

Relative vs. absolute judgements

• perceptual system mostly operates with relative judgements, not absolute
– that's why accuracy increases with common frame/scale and alignment

– Weber's Law: ratio of increment to background is constant

- filled rectangles differ in length by 1:9, difficult judgement
- white rectangles differ in length by 1:2, easy judgement

Relative luminance judgements

• perception of luminance is contextual based on contrast with surroundings

Relative color judgements

• color constancy across broad range of illumination conditions

Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014. – Chap 5: Marks and Channels
- On the Theory of Scales of Measurement. Stevens. Science 103:2684 (1946), 677–680.
- Psychophysics: Introduction to its Perceptual, Neural, and Social Prospects. Stevens. Wiley, 1975.
- Graphical Perception: Theory, Experimentation, and Application to the Development of Graphical Methods. Cleveland and McGill. Journ. American Statistical Association 79:387 (1984), 531–554.
- Perception in Vision. Healey. <http://www.csc.ncsu.edu/faculty/healey/PP>
- Visual Thinking for Design. Ware. Morgan Kaufmann, 2008.
- Information Visualization: Perception for Design, 3rd edition. Ware. Morgan Kaufmann / Academic Press, 2004.

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How?

Encode

- Arrange
- Express
- Order
- Use

Manipulate

- Change
- Select
- Navigate

Facet

- Juxtapose
- Partition
- Superimpose

Reduce

- Filter
- Aggregate
- Embed

Map from categorical and ordered attributes

- Color
- Hue
- Saturation
- Luminance
- Size, Angle, Curvature, ...
- Shape
- Motion
- Direction, Rate, Frequency, ...

What? Why? How?

Encode tables: Arrange space

Encode

Arrange



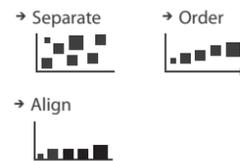
49

Arrange tables

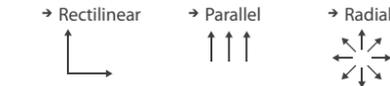
Express Values



Separate, Order, Align Regions



Axis Orientation



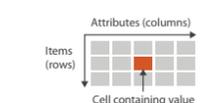
Layout Density



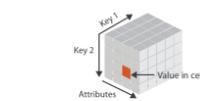
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Keys and values

Tables



Multidimensional Table

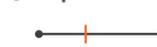


- **key**
 - independent attribute
 - used as unique index to look up items
 - simple tables: 1 key
 - multidimensional tables: multiple keys
- **value**
 - dependent attribute, value of cell
- **classify arrangements by key count**
 - 0, 1, 2, many...



Idiom: scatterplot

Express Values

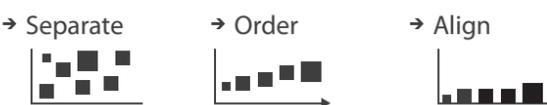


- **express values**
 - quantitative attributes
- **no keys, only values**
 - data
 - 2 quant attribs
 - mark: points
 - horiz + vert position
 - tasks
 - find trends, outliers, distribution, correlation, clusters
 - scalability
 - hundreds of items

[A layered grammar of graphics. Wickham. Journ. Computational and Graphical Statistics 19:1 (2010), 3–28.]

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Some keys: Categorical regions



- **regions: contiguous bounded areas distinct from each other**
 - using space to *separate* (proximity)
 - following expressiveness principle for categorical attributes
- **use ordered attribute to *order* and *align* regions**

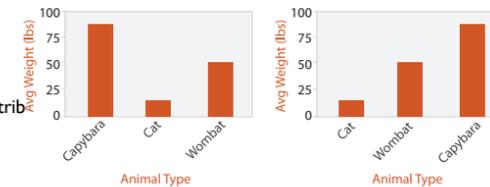


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Idiom: bar chart

one key, one value

- data
 - 1 categ attrib, 1 quant attrib
- mark: lines
- channels
 - length to express quant value
 - spatial regions: one per mark
 - separated horizontally, aligned vertically
 - ordered by quant attrib
 - » by label (alphabetical), by length attrib (data-driven)
- task
 - compare, lookup values
- scalability
 - dozens to hundreds of levels for key attrib

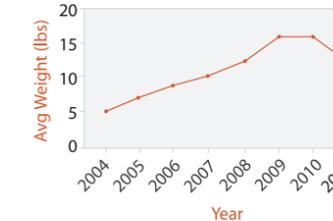


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Idiom: line chart

one key, one value

- data
 - 2 quant attribs
- mark: points
 - line connection marks between them
- channels
 - aligned lengths to express quant value
 - separated and ordered by key attrib into horizontal regions
- task
 - find trend
 - connection marks emphasize ordering of items along key axis by explicitly showing relationship between one item and the next

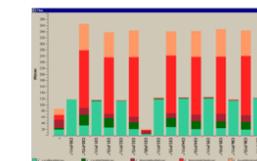


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Idiom: stacked bar chart

one more key

- data
 - 2 categ attrib, 1 quant attrib
- mark: vertical stack of line marks
 - **glyph: composite object, internal structure from multiple marks**
- channels
 - length and color hue
 - spatial regions: one per glyph
 - aligned: full glyph, lowest bar component
 - unaligned: other bar components
- task
 - part-to-whole relationship
- scalability
 - several to one dozen levels for stacked attrib



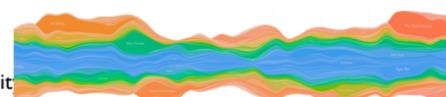
[Using Visualization to Understand the Behavior of Computer Systems. Bosch, Ph.D. thesis, Stanford Computer Science, 2001.]

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Idiom: streamgraph

generalized stacked graph

- emphasizing horizontal continuity
 - vs vertical items
- data
 - 1 categ key attrib (artist)
 - 1 ordered key attrib (time)
 - 1 quant value attrib (counts)
- derived data
 - geometry: layers, where height encodes counts
 - 1 quant attrib (layer ordering)
- scalability
 - hundreds of time keys
 - dozens to hundreds of artist keys
 - more than stacked bars, since most layers don't extend across whole chart



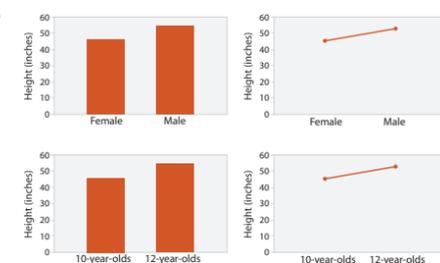
[Stacked Graphs Geometry & Aesthetics. Byron and Wattenberg. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2008) 14(6): 1245–1252, (2008).]

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Choosing bar vs line charts

depends on type of key attrib

- bar charts if categorical
- line charts if ordered
- **do not use line charts for categorical key attribs**
 - violates expressiveness principle
 - implication of trend so strong that it overrides semantics!
 - “The more male a person is, the taller he/she is”



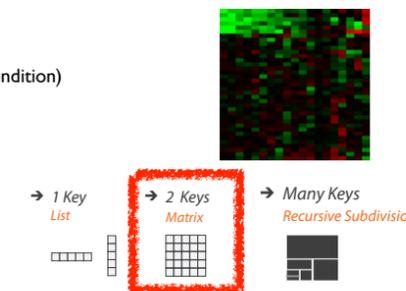
after [Bars and Lines: A Study of Graphic Communication. Zacks and Tversky. Memory and Cognition 27:6 (1999), 1073–1079.]

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Idiom: heatmap

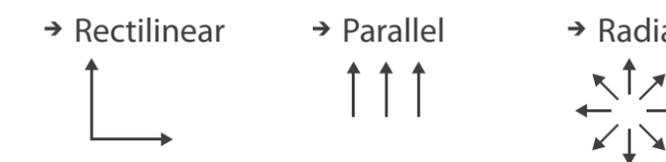
two keys, one value

- data
 - 2 categ attribs (gene, experimental condition)
 - 1 quant attrib (expression levels)
- marks: area
 - separate and align in 2D matrix
 - indexed by 2 categorical attributes
- channels
 - color by quant attrib
 - (ordered diverging colormap)
- task
 - find clusters, outliers
- scalability
 - 1M items, 100s of categ levels, ~10 quant attrib levels



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Axis Orientation

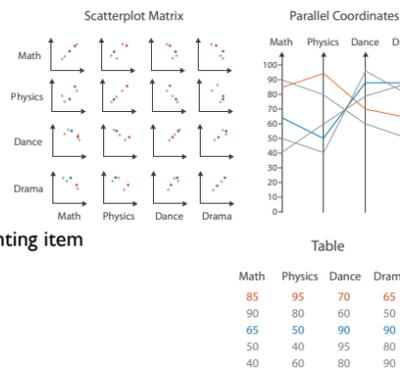


60

Idioms: scatterplot matrix, parallel coordinates

scatterplot matrix (SPLOM)

- rectilinear axes, point mark
- all possible pairs of axes
- scalability
 - one dozen attribs
 - dozens to hundreds of items
- **parallel coordinates**
 - parallel axes, jagged line representing item
 - rectilinear axes, item as point
 - axis ordering is major challenge
 - scalability
 - dozens of attribs
 - hundreds of items



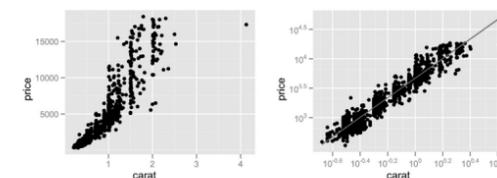
offer [Visualization Course Figures. McGuffin, 2014. <http://www.michaelmcguffin.com/courses/vis/>]

61

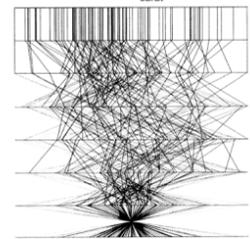
Task: Correlation

scatterplot matrix

- positive correlation
 - diagonal low-to-high
- negative correlation
 - diagonal high-to-low
- uncorrelated
- **parallel coordinates**
 - positive correlation
 - parallel line segments
 - negative correlation
 - all segments cross at halfway point
 - uncorrelated
 - scattered crossings



[A layered grammar of graphics. Wickham. Journ. Computational and Graphical Statistics 19:1 (2010), 3–28.]

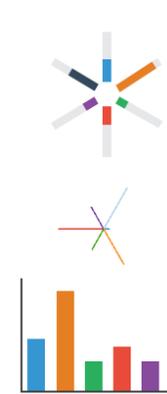


[Hyperdimensional Data Analysis Using Parallel Coordinates. Wigman. Journ. American Statistical Association 85:411 (1990), 664–675.]

62

Idioms: radial bar chart, star plot

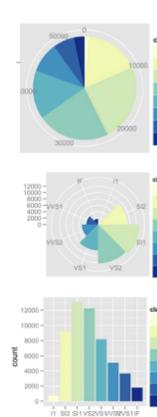
- **radial bar chart**
 - radial axes meet at central ring, line mark
- **star plot**
 - radial axes, meet at central point, line mark
- **bar chart**
 - rectilinear axes, aligned vertically
- **accuracy**
 - length unaligned with radial
 - less accurate than aligned with rectilinear



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Idioms: pie chart, polar area chart

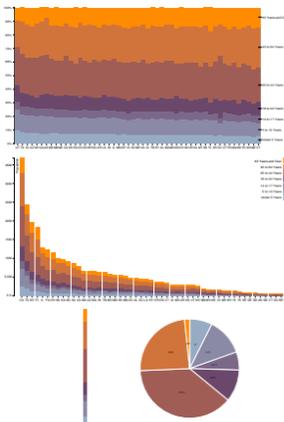
- **pie chart**
 - area marks with angle channel
 - accuracy: angle/area less accurate than line length
 - arclength also less accurate than line length
- **polar area chart**
 - area marks with length channel
 - more direct analog to bar charts
- **data**
 - 1 categ key attrib, 1 quant value attrib
- **task**
 - part-to-whole judgements



[A layered grammar of graphics. Wickham. Journ. Computational and Graphical Statistics 19:1 (2010), 3–28.]

Idioms: **normalized stacked bar chart**

- task
 - part-to-whole judgements
- normalized stacked bar chart
 - stacked bar chart, normalized to full vert height
 - single stacked bar equivalent to full pie
 - high information density: requires narrow rectangle
- pie chart
 - information density: requires large circle

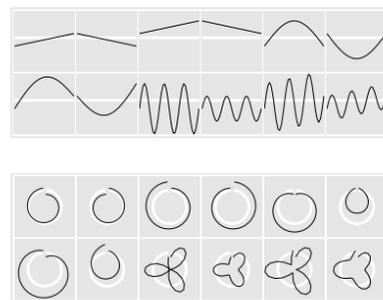
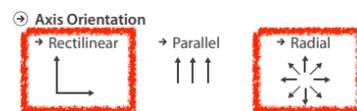


<http://bl.ocks.org/mbostock/3887235>
<http://bl.ocks.org/mbostock/3886208>
<http://bl.ocks.org/mbostock/3886394>

65

Idiom: **glyphmaps**

- rectilinear good for linear vs nonlinear trends
- radial good for cyclic patterns



[Glyph-maps for Visually Exploring Temporal Patterns in Climate Data and Models. Wickham, Heymann, Wickham, and Cook. *Environmetrics* 23:5 (2012), 382–393.]

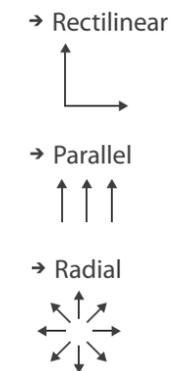
66

Orientation limitations

- rectilinear: scalability wrt #axes
 - 2 axes best
 - 3 problematic
 - more in afternoon
 - 4+ impossible
- parallel: unfamiliarity, training time
- radial: perceptual limits
 - angles lower precision than lengths
 - asymmetry between angle and length
 - can be exploited!

[Uncovering Strengths and Weaknesses of Radial Visualizations - an Empirical Approach. Diehl, Beck and Burch. *IEEE TVCG (Proc. InfoVis)* 16(6):935–942, 2010.]

⊕ Axis Orientation



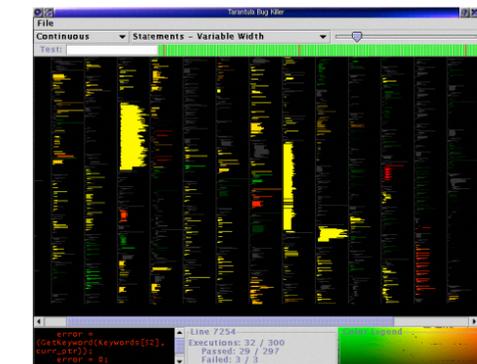
67

⊕ Layout Density

→ Dense



dense software overviews



[Visualization of test information to assist fault localization. Jones, Harrold, Staska. *Proc. ICSE 2002*, p. 467-477.]

68

Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
 - Chap 7: Arrange Tables
- Visualizing Data. Cleveland. Hobart Press, 1993.
- A Brief History of Data Visualization. Friendly. 2008. <http://www.datavis.ca/milestones>

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<http://www.cs.ubc.ca/~tmm/talks.html#vad16act>

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Arrange spatial data

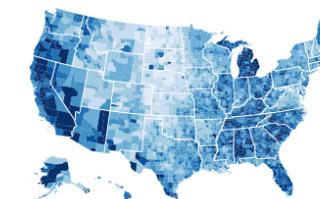
⊕ Use Given

- Geometry
 - Geographic
 - Other Derived
- Spatial Fields
 - Scalar Fields (one value per cell)
 - Isocontours
 - Direct Volume Rendering
 - Vector and Tensor Fields (many values per cell)
 - Flow Glyphs (local)
 - Geometric (sparse seeds)
 - Textures (dense seeds)
 - Features (globally derived)



Idiom: **choropleth map**

- use given spatial data
 - when central task is understanding spatial relationships
- data
 - geographic geometry
 - table with 1 quant attribute per region
- encoding
 - use given geometry for area mark boundaries
 - sequential segmented colormap [more later]

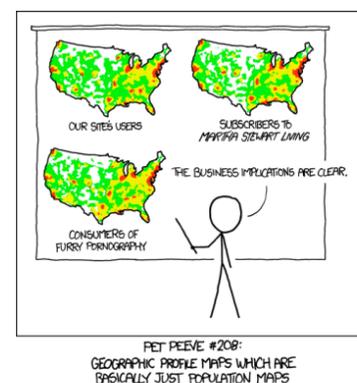


<http://bl.ocks.org/mbostock/4060606>

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Beware: Population maps trickiness!

[<https://xkcd.com/1138>]



Idiom: **topographic map**

- data
 - geographic geometry
 - scalar spatial field
 - 1 quant attribute per grid cell
- derived data
 - isoline geometry
 - isocontours computed for specific levels of scalar values

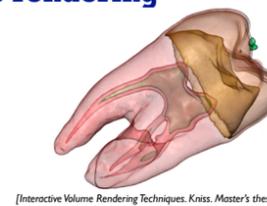


Land Information New Zealand Data Service

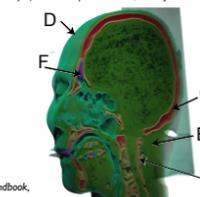
74

Idioms: **isosurfaces, direct volume rendering**

- data
 - scalar spatial field
 - 1 quant attribute per grid cell
- task
 - shape understanding, spatial relationships
- isosurface
 - derived data: isocontours computed for specific levels of scalar values
- direct volume rendering
 - transfer function maps scalar values to color, opacity



[Interactive Volume Rendering Techniques. Kniss. Master's thesis, University of Utah Computer Science, 2002.]

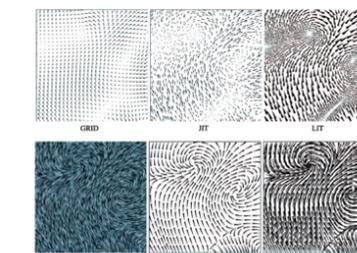


[Multidimensional Transfer Functions for Volume Rendering. Kniss, Kindmann, and Hansen. In The Visualization Handbook, edited by Charles Hansen and Christopher Johnson, pp. 189–210. Elsevier, 2005.]

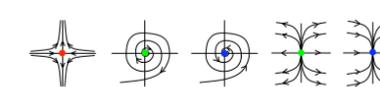
75

Vector and tensor fields

- data
 - many attribs per cell
- idiom families
 - flow glyphs
 - purely local
 - geometric flow
 - derived data from tracing particle trajectories
 - sparse set of seed points
 - texture flow
 - derived data, dense seeds
 - feature flow
 - global computation to detect features
 - encoded with one of methods above



[Comparing 2D vector field visualization methods: A user study. Laidlaw et al. *IEEE Trans. Visualization and Computer Graphics (TVCG)* 11:1 (2005), 59–70.]

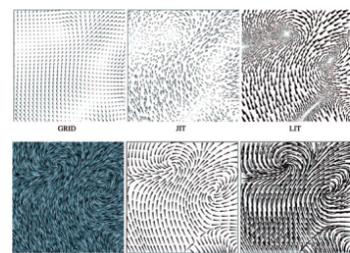


[Topology tracking for the visualization of time-dependent two-dimensional flows. Tricoche, Wächhoff, Scheuermann, and Hagen. *Computers & Graphics* 26:2 (2002), 249–257.]

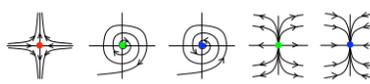
76

Vector fields

- empirical study tasks
 - finding critical points, identifying their types
 - identifying what type of critical point is at a specific location
 - predicting where a particle starting at a specified point will end up (advection)



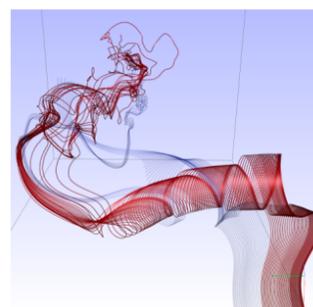
[Comparing 2D vector field visualization methods: A user study. Laidlaw et al. *IEEE Trans. Visualization and Computer Graphics (TVCG)* 11:1 (2005), 59–70.]



[Topology tracking for the visualization of time-dependent two-dimensional flows. Tricoche, Wächhoff, Scheuermann, and Hagen. *Computers & Graphics* 26:2 (2002), 249–257.]

Idiom: **similarity-clustered streamlines**

- data
 - 3D vector field
- derived data (from field)
 - streamlines: trajectory particle will follow
- derived data (per streamline)
 - curvature, torsion, tortuosity
 - signature: complex weighted combination
 - compute cluster hierarchy across all signatures
 - encode: color and opacity by cluster
- tasks
 - find features, query shape
- scalability
 - millions of samples, hundreds of streamlines



[Similarity Measures for Enhancing Interactive Streamline Seeding. McLoughlin, Jones, Laramee, Malki, Masters, and Hansen. *IEEE Trans. Visualization and Computer Graphics* 19:8 (2013), 1342–1353.]

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Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
 - Chap 8: Arrange Spatial Data
- How Maps Work: Representation, Visualization, and Design. MacEachren. Guilford Press, 1995.
- Overview of visualization. Schroeder and. Martin. In The Visualization Handbook, edited by Charles Hansen and Christopher Johnson, pp. 3–39. Elsevier, 2005.
- Real-Time Volume Graphics. Engel, Hadwiger, Kniss, Reza-Salama, and Weiskopf. AK Peters, 2006.
- Overview of flow visualization. Weiskopf and Erlebacher. In The Visualization Handbook, edited by Charles Hansen and Christopher Johnson, pp. 261–278. Elsevier, 2005.

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<http://www.cs.ubc.ca/~tmm/talks.html#vad16act>

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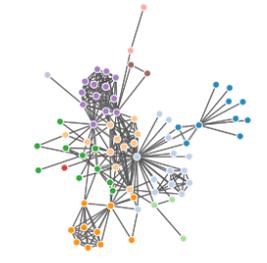
80

Arrange networks and trees

- Node-Link Diagrams**
Connection Marks
✓ NETWORKS ✓ TREES
- Adjacency Matrix**
Derived Table
✓ NETWORKS ✓ TREES
- Enclosure**
Containment Marks
✗ NETWORKS ✓ TREES

Idiom: force-directed placement

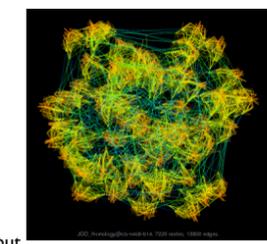
- visual encoding
 - link connection marks, node point marks
- considerations
 - spatial position: no meaning directly encoded
 - left free to minimize crossings
 - proximity semantics?
 - sometimes meaningful
 - sometimes arbitrary, artifact of layout algorithm
 - tension with length
 - long edges more visually salient than short
- tasks
 - explore topology; locate paths, clusters
- scalability
 - node/edge density $E < 4N$



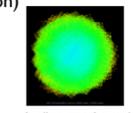
<http://mbostock.github.com/d3/lex/force.html>

Idiom: sfdp (multi-level force-directed placement)

- data
 - original: network
 - derived: cluster hierarchy atop it
- considerations
 - better algorithm for same encoding technique
 - same: fundamental use of space
 - hierarchy used for algorithm speed/quality but not shown explicitly
 - (more on algorithm vs encoding in afternoon)
- scalability
 - nodes, edges: 1K-10K
 - hairball problem eventually hits



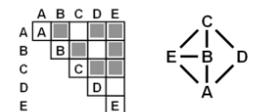
[Efficient and high quality force-directed graph drawing. Hu. The Mathematica Journal 10:37-71, 2005.]



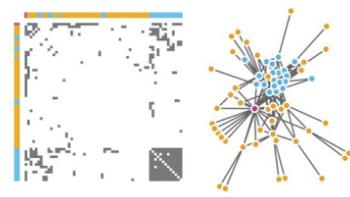
<http://www.research.att.com/yfoshu/GALLERY/GRAPHS/index1.html>

Idiom: adjacency matrix view

- data: network
 - transform into same data/encoding as heatmap
- derived data: table from network
 - 1 quant attrib
 - weighted edge between nodes
 - 2 categ attribs: node list x 2
- visual encoding
 - cell shows presence/absence of edge
- scalability
 - 1K nodes, 1M edges



[NodeTrix: a Hybrid Visualization of Social Networks. Henry, Fekete, and McGuffin. IEEE TVCG (Proc. InfoVis) 13(6):1302-1309, 2007.]

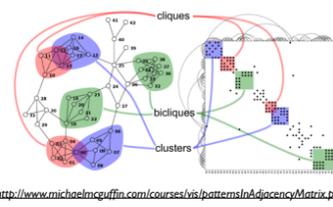


[Points of view: Networks. Gehlenborg and Wang. Nature Methods 9:115.]

Connection vs. adjacency comparison

- adjacency matrix strengths
 - predictability, scalability, supports reordering
 - some topology tasks trainable
- node-link diagram strengths
 - topology understanding, path tracing
 - intuitive, no training needed
- empirical study
 - node-link best for small networks
 - matrix best for large networks
 - if tasks don't involve topological structure!

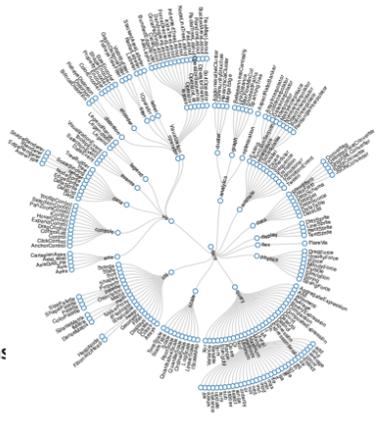
[On the readability of graphs using node-link and matrix-based representations: a controlled experiment and statistical analysis. Ghoniem, Fekete, and Castagliola. Information Visualization 4:2 (2005), 114-135.]



<http://www.michaelmcguffin.com/courses/vis/patterns/AdjacencyMatrix.png>

Idiom: radial node-link tree

- data
 - tree
- encoding
 - link connection marks
 - point node marks
 - radial axis orientation
 - angular proximity: siblings
 - distance from center: depth in tree
- tasks
 - understanding topology, following paths
- scalability
 - 1K - 10K nodes



<http://mbostock.github.com/d3/lex/tree.html>

Idiom: treemap

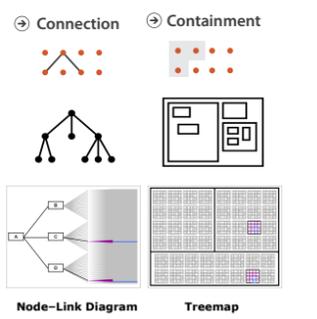
- data
 - tree
 - 1 quant attrib at leaf nodes
- encoding
 - area containment marks for hierarchical structure
 - rectilinear orientation
 - size encodes quant attrib
- tasks
 - query attribute at leaf nodes
- scalability
 - 1M leaf nodes



http://tulip.lafri.fr/Documentation/3_7userHandbook/html/ch06.html

Link marks: Connection and containment

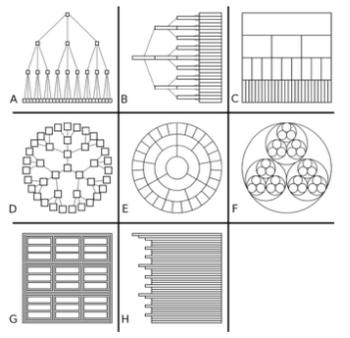
- marks as links (vs. nodes)
 - common case in network drawing
 - 1D case: connection
 - ex: all node-link diagrams
 - emphasizes topology, path tracing
 - networks and trees
 - 2D case: containment
 - ex: all treemap variants
 - emphasizes attribute values at leaves (size coding)
 - only trees



[Elastic Hierarchies: Combining Treemaps and Node-Link Diagrams. Dong, McGuffin, and Chignell. Proc. InfoVis 2005, p. 57-64.]

Tree drawing idioms comparison

- data shown
 - link relationships
 - tree depth
 - sibling order
- design choices
 - connection vs containment link marks
 - rectilinear vs radial layout
 - spatial position channels
- considerations
 - redundant? arbitrary?
 - information density?
 - avoid wasting space



[Quantifying the Space-Efficiency of 2D Graphical Representations of Trees. McGuffin and Robert. Information Visualization 9:2 (2010), 115-140.]

Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
 - Chap 9: Arrange Networks and Trees
- Visual Analysis of Large Graphs: State-of-the-Art and Future Research Challenges. von Landesberger et al. Computer Graphics Forum 30:6 (2011), 1719-1749.
- Simple Algorithms for Network Visualization: A Tutorial. McGuffin. Tsinghua Science and Technology (Special Issue on Visualization and Computer Graphics) 17:4 (2012), 383-398.
- Drawing on Physical Analogies. Brandes. In Drawing Graphs: Methods and Models, LNCS Tutorial, 2025, edited by M. Kaufmann and D. Wagner, LNCS Tutorial, 2025, pp. 71-86. Springer-Verlag, 2001.
- <http://www.treevis.net> Treevis.net: A Tree Visualization Reference. Schulz. IEEE Computer Graphics and Applications 31:6 (2011), 11-15.
- Perceptual Guidelines for Creating Rectangular Treemaps. Kong, Heer, and Agrawala. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis) 16:6 (2010), 990-998.

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<http://www.cs.ubc.ca/~tmm/talks.html#vad16act> @tamaramunzner

Idiom design choices: Encode

Encode

- Arrange
 - Express
 - Order
 - Use
- Separate
- Align
- Map from categorical and ordered attributes
 - Color
 - Hue → Saturation → Luminance
 - Size, Angle, Curvature, ...
 - Shape
 - Motion

Categorical vs ordered color

[Seriously Colorful: Advanced Color Principles & Practices. Stone. Tableau Customer Conference 2014.]

Color: Luminance, saturation, hue

- 3 channels
 - identity for categorical
 - hue
 - magnitude for ordered
 - luminance
 - saturation
- RGB: poor for encoding
- HSL: better, but beware
 - lightness ≠ luminance

Spectral sensitivity

Opponent color and color deficiency

- perceptual processing before optic nerve
 - one achromatic luminance channel L
 - edge detection through luminance contrast
 - two chroma channels, R-G and Y-B axis
- "color blind" if one axis has degraded acuity
 - 8% of men are red/green color deficient
 - blue/yellow is rare

[Seriously Colorful: Advanced Color Principles & Practices. Stone. Tableau Customer Conference 2014.]

Designing for color deficiency: Check with simulator

Normal vision **Deuteranope** **Protanope** **Tritanope**

<http://rehue.net>

[Seriously Colorful: Advanced Color Principles & Practices. Stone, Tableau Customer Conference 2014.]

Designing for color deficiency: Avoid encoding by hue alone

- redundantly encode
 - vary luminance
 - change shape

Change the shape

Vary luminance

[Seriously Colorful: Advanced Color Principles & Practices. Stone, Tableau Customer Conference 2014.]

Color deficiency: Reduces color to 2 dimensions

Normal **Protanope**

Deuteranope **Tritanope**

[Seriously Colorful: Advanced Color Principles & Practices. Stone, Tableau Customer Conference 2014.]

Designing for color deficiency: Blue-Orange is safe

[Seriously Colorful: Advanced Color Principles & Practices. Stone, Tableau Customer Conference 2014.]

Bezold Effect: Outlines matter

- color constancy: simultaneous contrast effect

[Seriously Colorful: Advanced Color Principles & Practices. Stone, Tableau Customer Conference 2014.]

Color/Lightness constancy: Illumination conditions

Do they match?

Image courtesy of John McCann

Color/Lightness constancy: Illumination conditions

Do they match?

Image courtesy of John McCann

Colormaps

- Categorical
 - Binary
 - Ordered
 - Diverging
- Sequential

after [Color Use Guidelines for Mapping and Visualization. Brewer, 1994. <http://www.personal.psu.edu/faculty/c/a/cab38/ColorSch/Schemes.html>]

Colormaps

- Categorical
- Ordered
- Bivariate

after [Color Use Guidelines for Mapping and Visualization. Brewer, 1994. <http://www.personal.psu.edu/faculty/c/a/cab38/ColorSch/Schemes.html>]

Colormaps

- Categorical
- Ordered
- Bivariate

use with care!

after [Color Use Guidelines for Mapping and Visualization. Brewer, 1994. <http://www.personal.psu.edu/faculty/c/a/cab38/ColorSch/Schemes.html>]

Colormaps

- Categorical
- Ordered
- Bivariate

- color channel interactions
 - size heavily affects salience
 - small regions need high saturation
 - large need low saturation
 - saturation & luminance: 3-4 bins max
 - also not separable from transparency

after [Color Use Guidelines for Mapping and Visualization. Brewer, 1994. <http://www.personal.psu.edu/faculty/c/a/cab38/ColorSch/Schemes.html>]

Categorical color: Discriminability constraints

- noncontiguous small regions of color: only 6-12 bins

[Cinteny: flexible analysis and visualization of synteny and genome rearrangements in multiple organisms. Sinha and Meller. BMC Bioinformatics, 8:82, 2007.]

ColorBrewer

- <http://www.colorbrewer2.org>
- saturation and area example: size affects salience!

[Why Should Engineers Be Worried About Color? Treish and Ragwitz, 1998. <http://www.research.ibm.com/people/treish/color/color.htm>]

Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable

[Why Should Engineers Be Worried About Color? Treish and Ragwitz, 1998. <http://www.research.ibm.com/people/treish/color/color.htm>]

Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable
- alternatives
 - large-scale structure: fewer hues

[Why Should Engineers Be Worried About Color? Treish and Ragwitz, 1998. <http://www.research.ibm.com/people/treish/color/color.htm>]

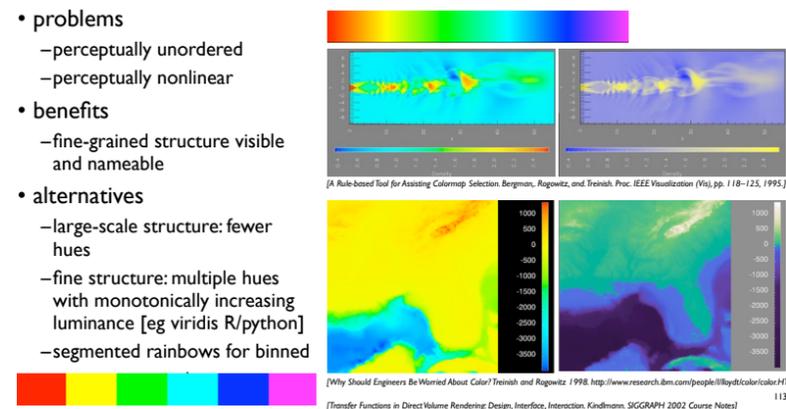
Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable
- alternatives
 - large-scale structure: fewer hues
 - fine structure: multiple hues with monotonically increasing luminance [eg viridis R/python]

[Why Should Engineers Be Worried About Color? Treish and Ragwitz, 1998. <http://www.research.ibm.com/people/treish/color/color.htm>]

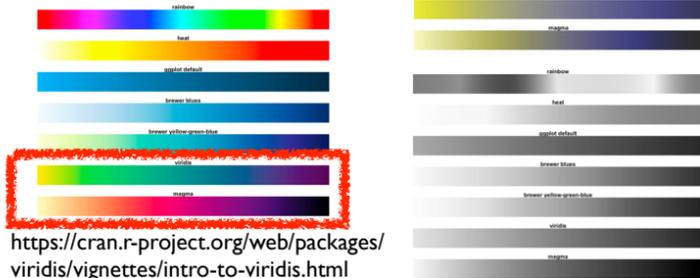
Ordered color: Rainbow is poor default

- problems
 - perceptually unordered
 - perceptually nonlinear
- benefits
 - fine-grained structure visible and nameable
- alternatives
 - large-scale structure: fewer hues
 - fine structure: multiple hues with monotonically increasing luminance [eg viridis R/python]
 - segmented rainbows for binned



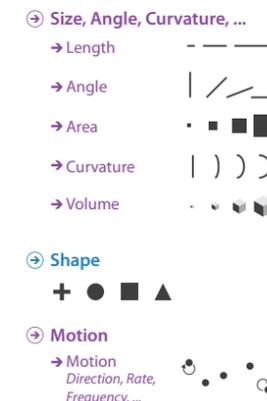
Viridis

- colorful, perceptually uniform, colorblind-safe, monotonically increasing luminance

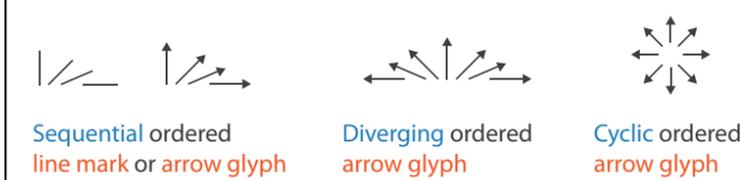


Map other channels

- size
 - length accurate, 2D area ok, 3D volume poor
- angle
 - nonlinear accuracy
 - horizontal, vertical, exact diagonal
- shape
 - complex combination of lower-level primitives
 - many bins
- motion
 - highly separable against static
 - binary: great for highlighting
 - use with care to avoid irritation



Angle



Further reading

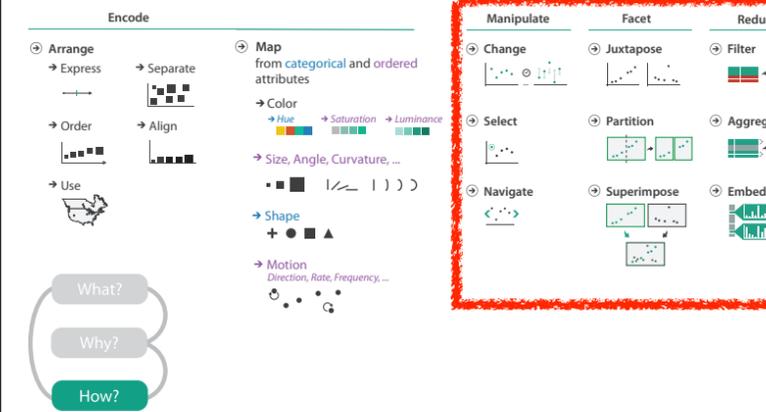
- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014
 - Chap 10: Map Color and Other Channels
- ColorBrewer, Brewer.
 - <http://www.colorbrewer2.org>
- Color In Information Display. Stone. IEEE Vis Course Notes, 2006.
 - <http://www.stonesc.com/Vis06>
- A Field Guide to Digital Color. Stone. AK Peters, 2003.
- Rainbow Color Map (Still) Considered Harmful. Borland and Taylor. IEEE Computer Graphics and Applications 27:2 (2007), 14–17.
- Visual Thinking for Design. Ware. Morgan Kaufmann, 2008.
- Information Visualization: Perception for Design, 3rd edition. Ware. Morgan Kaufmann / Academic Press, 2004.
- <https://cran.r-project.org/web/packages/viridis/vignettes/intro-to-viridis.html>

Outline

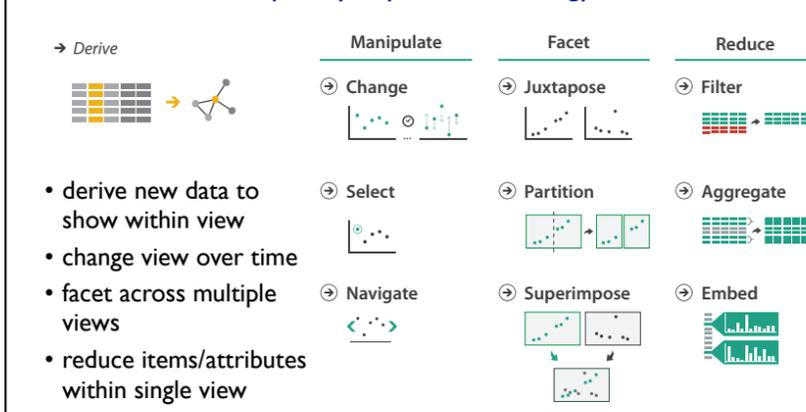
- Session 1 8:30-10:00am Visualization Analysis Framework**
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 - Marks and Channels
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 - Arrange Tables
 - Arrange Spatial Data
 - Arrange Networks and Trees
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 - Map Color
 - Manipulate: Change, Select, Navigate
 - Facet: Juxtapose, Partition, Superimpose
- Session 4 3:00-4:30pm Guidelines and Examples**
 - Reduce: Filter, Aggregate
 - Rules of Thumb
 - Q&A

<http://www.cs.ubc.ca/~tmm/talks.html#vad16act> @tamaramunzner

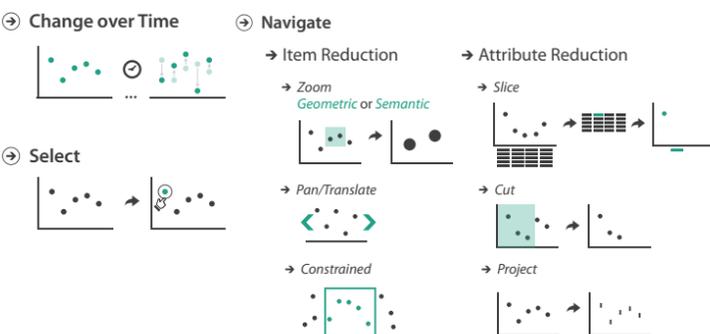
How?



How to handle complexity: 1 previous strategy + 3 more



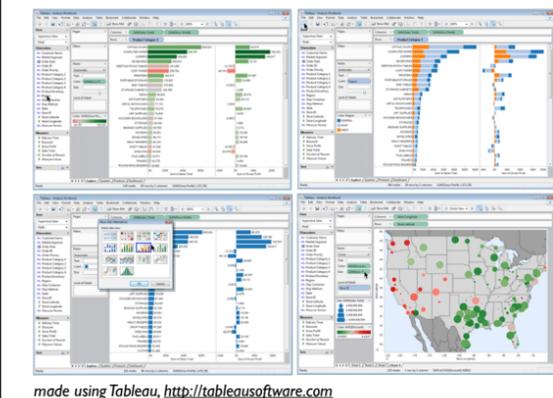
Manipulate



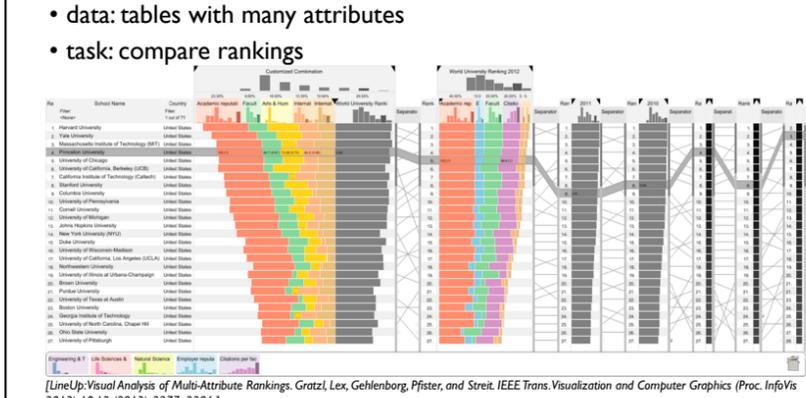
Change over time

- change any of the other choices
 - encoding itself
 - parameters
 - arrange: rearrange, reorder
 - aggregation level, what is filtered...
- interaction entails change

Idiom: Re-encode System: Tableau

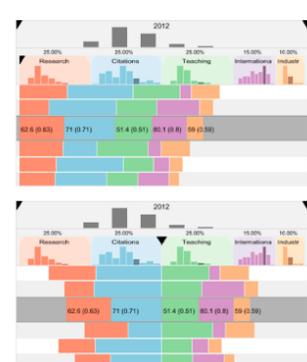


Idiom: Reorder System: LineUp



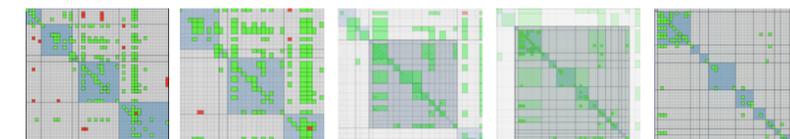
Idiom: Realign System: LineUp

- stacked bars
 - easy to compare
 - first segment
 - total bar
- align to different segment
 - supports flexible comparison



Idiom: Animated transitions

- smooth transition from one state to another
 - alternative to jump cuts
 - support for item tracking when amount of change is limited
- example: multilevel matrix views
- example: animated transitions in statistical data graphics
 - <https://vimeo.com/19278444>



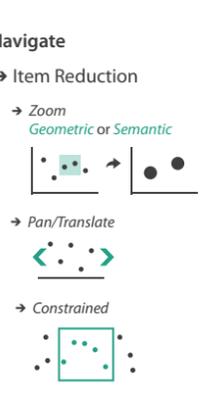
Select and highlight

- selection: basic operation for most interaction
- design choices
 - how many selection types?
 - click vs hover: heavyweight, lightweight
 - primary vs secondary: semantics (eg source/target)
- highlight: change visual encoding for selection targets
 - color
 - limitation: existing color coding hidden
 - other channels (eg motion)
 - add explicit connection marks between items



Navigate: Changing item visibility

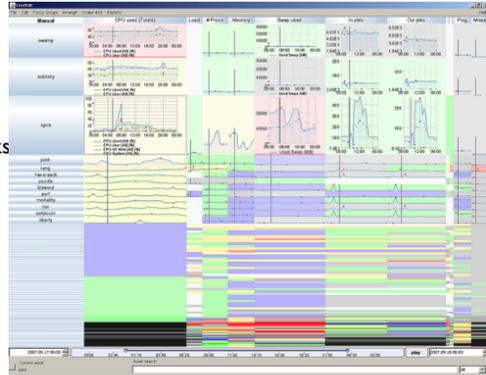
- change viewpoint
 - changes which items are visible within view
 - camera metaphor
 - zoom
 - geometric zoom: familiar semantics
 - semantic zoom: adapt object representation based on available pixels
 - » dramatic change, or more subtle one
 - pan/translate
 - rotate
 - especially in 3D
 - constrained navigation
 - often with animated transitions
 - often based on selection set



Idiom: Semantic zooming

System: LiveRAC

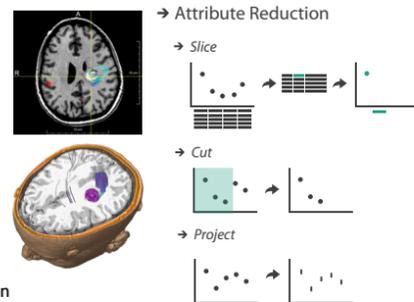
- visual encoding change
 - colored box
 - sparkline
 - simple line chart
 - full chart: axes and tickmarks



[LiveRAC - Interactive Visual Exploration of System Management Time-Series Data. McLachlan, Munzner, Koutsosios, and North. Proc. ACM Conf. Human Factors in Computing Systems (CHI), pp. 1483–1492, 2008.]

Navigate: Reducing attributes

- continuation of camera metaphor
 - slice
 - show only items matching specific value for given attribute: slicing plane
 - axis aligned, or arbitrary alignment
 - cut
 - show only items on far side of plane from camera
 - project
 - change mathematics of image creation
 - orthographic
 - perspective
 - many others: Mercator, cabinet, ...



[Interactive Visualization of Multimodal Volume Data for Neurosurgical Tumor Treatment. Rieder, Ritter, Raspe, and Peitgen. Computer Graphics Forum (Proc. EuroVis 2008) 27:3 (2008), 1055–1062.]

Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
 - Chap 11: Manipulate View
- Animated Transitions in Statistical Data Graphics. Heer and Robertson. IEEE Trans. on Visualization and Computer Graphics (Proc. InfoVis07) 13:6 (2007), 1240–1247.
- Selection: 524,288 Ways to Say “This is Interesting”. Wills. Proc. IEEE Symp. Information Visualization (InfoVis), pp. 54–61, 1996.
- Smooth and efficient zooming and panning. van Wijk and Nuij. Proc. IEEE Symp. Information Visualization (InfoVis), pp. 15–22, 2003.
- Starting Simple - adding value to static visualisation through simple interaction. Dix and Ellis. Proc. Advanced Visual Interfaces (AVI), pp. 124–134, 1998.

Outline

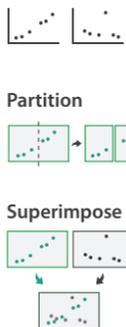
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<http://www.cs.ubc.ca/~tmm/talks.html#vad16act>

@tamaramunzner

Facet

- Juxtapose
- Partition
- Superimpose



Juxtapose and coordinate views

→ Share Encoding: Same/Different

→ Linked Highlighting



→ Share Data: All/Subset/None



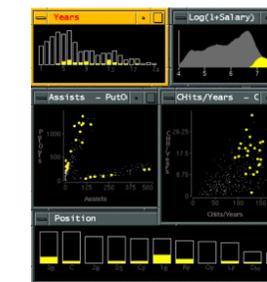
→ Share Navigation



Idiom: Linked highlighting

System: EDV

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



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

Idiom: bird's-eye maps

System: Google Maps

- encoding: same
- data: subset shared
- navigation: shared
 - bidirectional linking
- differences
 - viewpoint
 - (size)
- overview-detail

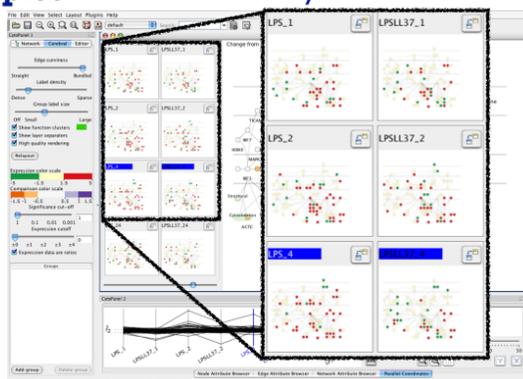


[A Review of Overview+Detail, Zooming, and Focus+Context Interfaces. Cockburn, Karlson, and Bederson. ACM Computing Surveys 41:1 (2008), 1–31.]

Idiom: Small multiples

System: Cerebral

- 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, Gardy, and Kincaid. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2008) 14:6 (2008), 1253–1260.]

Coordinate views: Design choice interaction

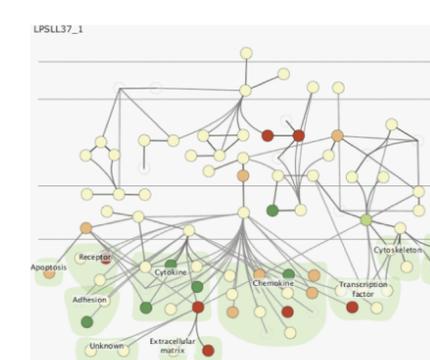
		Data		
		All	Subset	None
Encoding	Same	Redundant	Overview/Detail	Small Multiples
	Different	Multiform	Multiform, Overview/Detail	No Linkage

• 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

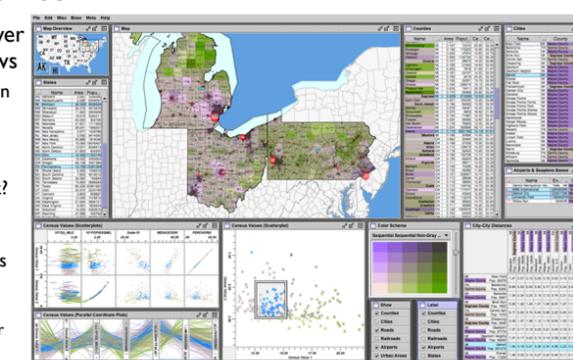
Why not animation?

- disparate frames and regions: comparison difficult
 - vs contiguous frames
 - vs small region
 - vs coherent motion of group
- safe special case
 - animated transitions



System: Improve

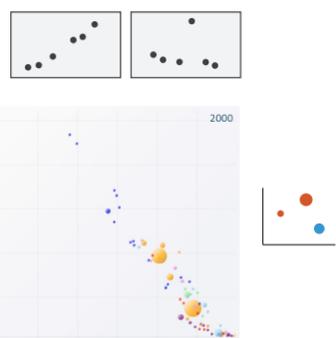
- investigate power of multiple views
 - pushing limits on view count, interaction complexity
 - how many is ok?
 - open research question
 - reorderable lists
 - easy lookup
 - useful when linked to other encodings



[Building Highly-Coordinated Visualizations In Improve. Weaver. Proc. IEEE Symp. Information Visualization (InfoVis), pp. 159–166, 2004.]

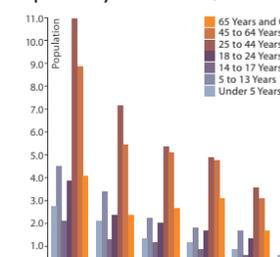
Partition into views

- how to divide data between views → Partition into Side-by-Side Views
 - split into regions by attributes
 - encodes association between items using spatial proximity
 - order of splits has major implications for what patterns are visible
- no strict dividing line
 - view: big/detailed
 - contiguous region in which visually encoded data is shown on the display
 - glyph: small/iconic
 - object with internal structure that arises from multiple marks

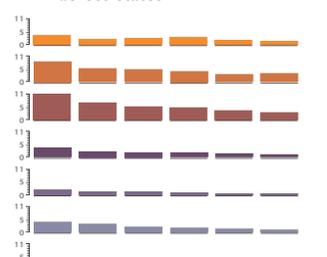


Partitioning: List alignment

- 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



- small-multiple bar charts
 - split by age into regions
 - one chart per region
 - compare: easy within age, harder across states



Partitioning: Recursive subdivision

System: HIVE

- split by neighborhood
- then by type
 - years as rows
 - months as columns
- color by price
- neighborhood patterns
 - where it's expensive
 - where you pay much more for detached type

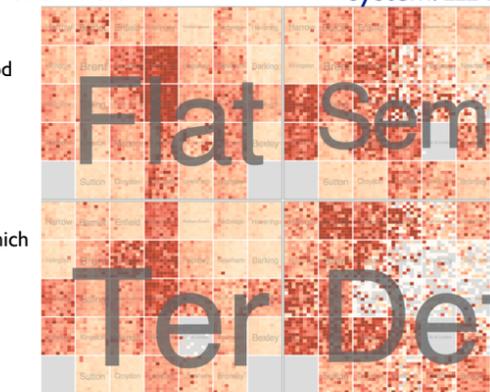


[Configuring Hierarchical Layouts to Address Research Questions. Slingsby, Dykes, and Wood. IEEE Transactions on Visualization and Computer Graphics (Proc. InfoVis 2009) 15:6 (2009), 977–984.]

Partitioning: Recursive subdivision

System: HIVE

- switch order of splits
 - type then neighborhood
- switch color
 - by price variation
- type patterns
 - within specific type, which neighborhoods inconsistent

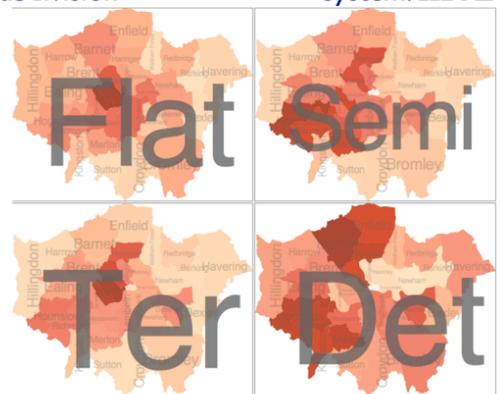


[Configuring Hierarchical Layouts to Address Research Questions. Slingsby, Dykes, and Wood. IEEE Transactions on Visualization and Computer Graphics (Proc. InfoVis 2009) 15:6 (2009), 977–984.]

Partitioning: Recursive subdivision

System: HIVE

- different encoding for second-level regions
- choropleth maps



[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

System: HIVE

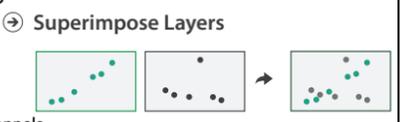
- size regions by sale counts
- not uniformly
- result: treemap



[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.]

Superimpose layers

- layer: set of objects spread out over region
- each set is visually distinguishable group
- extent: whole view
- design choices
- how many layers, how to distinguish?
- encode with different, nonoverlapping channels
- two layers achievable, three with careful design
- small static set, or dynamic from many possible?



Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
- Chap 12: Facet Into Multiple Views
- A Review of Overview+Detail, Zooming, and Focus+Context Interfaces. Cockburn, Karlson, and Bederson. ACM Computing Surveys 41:1 (2008), 1–31.
- A Guide to Visual Multi-Level Interface Design From Synthesis of Empirical Study Evidence. Lam and Munzner. Synthesis Lectures on Visualization Series, Morgan Claypool, 2010.
- Zooming versus multiple window interfaces: Cognitive costs of visual comparisons. Plumlee and Ware. ACM Trans. on Computer-Human Interaction (TOCHI) 13:2 (2006), 179–209.
- Exploring the Design Space of Composite Visualization. Javed and Elmquist. Proc. Pacific Visualization Symp. (PacificVis), pp. 1–9, 2012.
- Visual Comparison for Information Visualization. Gleicher, Albers, Walker, Jusufi, Hansen, and Roberts. Information Visualization 10:4 (2011), 289–309.
- Guidelines for Using Multiple Views in Information Visualizations. Baldonado, Woodruff, and Kuchinsky. In Proc. ACM Advanced Visual Interfaces (AVI), pp. 110–119, 2000.
- Cross-Filtered Views for Multidimensional Visual Analysis. Weaver. IEEE Trans. Visualization and Computer Graphics 16:2 (Proc. InfoVis 2010), 192–204, 2010.
- Linked Data Views. Willis. In Handbook of Data Visualization, Computational Statistics, edited by Unwin, Chen, and Härdle, pp. 216–241. Springer-Verlag, 2008.
- Glyph-based Visualization: Foundations, Design Guidelines, Techniques and Applications. Borgo, Kehrer, Chung, Maguire, Laramee, Hauser, Ward, and Chen. In Eurographics State of the Art Reports, pp. 39–63, 2013.

Static visual layering

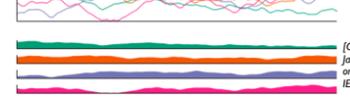
- foreground layer: roads
- hue, size distinguishing main from minor
- high luminance contrast from background
- background layer: regions
- desaturated colors for water, parks, land areas
- user can selectively focus attention
- “get it right in black and white”
- check luminance contrast with greyscale view



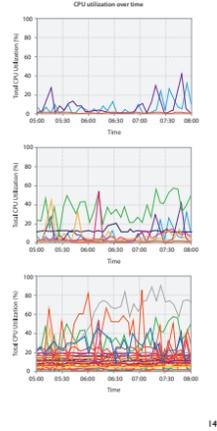
[Get it right in black and white. Stone. 2010. <http://www.stones.com/wordpress/2010/03/get-it-right-in-black-and-white/>]

Superimposing limits

- few layers, but many lines
- up to a few dozen
- but not hundreds
- superimpose vs juxtapose: empirical study
- superimposed for local, multiple for global
- tasks
- local: maximum, global: slope, discrimination
- same screen space for all multiples vs single superimposed



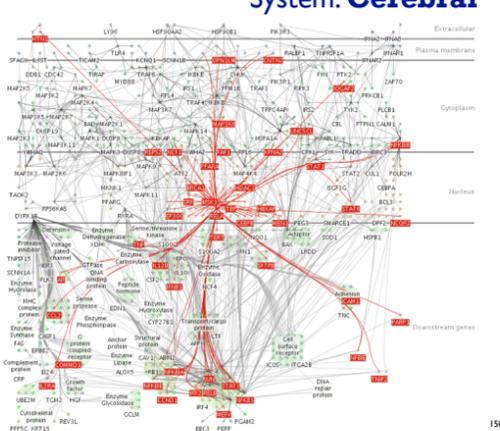
[Graphical Perception of Multiple Time Series. Javed, McDonnell, and Elmquist. IEEE Transactions on Visualization and Computer Graphics (Proc. InfoVis 2010) 16:6 (2010), 927–934.]



Dynamic visual layering

System: Cerebral

- interactive, from selection
- lightweight: click
- very lightweight: hover
- ex: 1-hop neighbors



[Cerebral: a Cytoscape plugin for layout of and interaction with biological networks using subcellular localization annotation. Barsky, Gardy, Hancock, and Munzner. Bioinformatics 23:8 (2007), 1040–1042.]

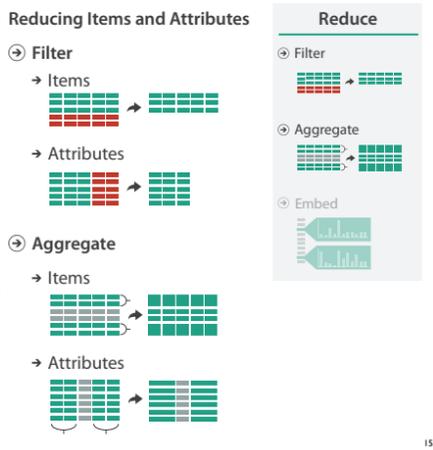
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<http://www.cs.ubc.ca/~tmm/talks.html#vad16act>

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Reduce items and attributes

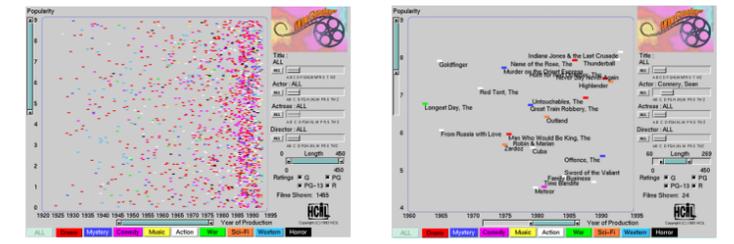


- reduce/increase: inverses
- filter
- pro: straightforward and intuitive
- to understand and compute
- con: out of sight, out of mind
- aggregation
- pro: inform about whole set
- con: difficult to avoid losing signal
- not mutually exclusive
- combine filter, aggregate
- combine reduce, change, facet

Idiom: dynamic filtering

System: FilmFinder

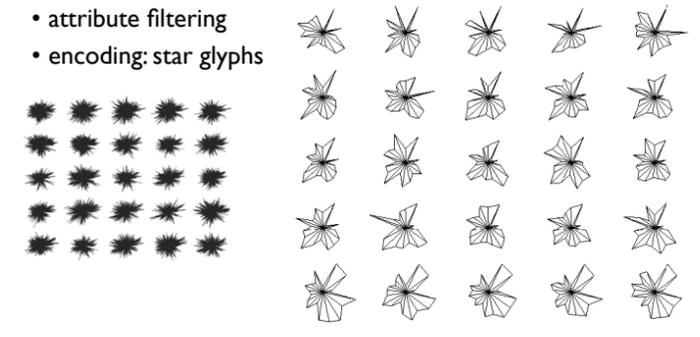
- item filtering
- browse through tightly coupled interaction
- alternative to queries that might return far too many or too few



[Visual information seeking: Tight coupling of dynamic query filters with starfield displays. Ahlberg and Shneiderman. Proc. ACM Conf. on Human Factors in Computing Systems (CHI), pp. 313–317, 1994.]

Idiom: DOSFA

- attribute filtering
- encoding: star glyphs



[Interactive Hierarchical Dimension Ordering, Spacing and Filtering for Exploration Of High Dimensional Datasets. Yang, Peng, Ward, and Rundensteiner. Proc. IEEE Symp. Information Visualization (InfoVis), pp. 105–112, 2003.]

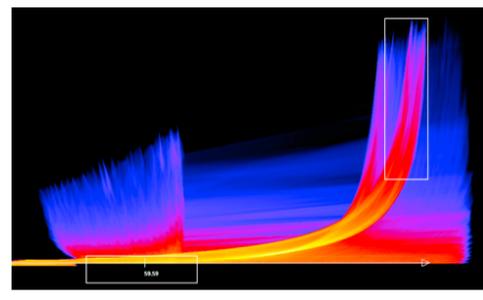
Idiom: histogram

- static item aggregation
- task: find distribution
- data: table
- derived data
- new table: keys are bins, values are counts
- bin size crucial
- pattern can change dramatically depending on discretization
- opportunity for interaction: control bin size on the fly



Continuous scatterplot

- static item aggregation
- data: table
- derived data: table
- key attribs x,y for pixels
- quant attrib: overplot density
- dense space-filling 2D matrix
- color: sequential categorical hue + ordered luminance



[Continuous Scatterplots. Bachthaler and Weiskopf. IEEE TVCG (Proc. Vis 08) 14:6 (2008), 1428–1435. 2008.]

Idiom: scented widgets

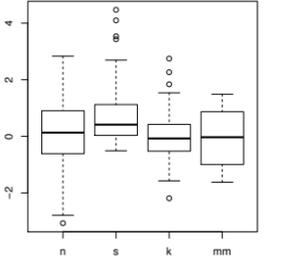
- augment widgets for filtering to show information scent
- cues to show whether value in drilling down further vs looking elsewhere
- concise, in part of screen normally considered control panel



[Scented Widgets: Improving Navigation Cues with Embedded Visualizations. Willett, Heer, and Agrawala. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2007) 13:6 (2007), 1129–1136.]

Idiom: boxplot

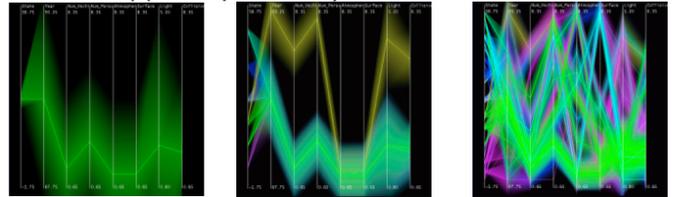
- static item aggregation
- task: find distribution
- data: table
- derived data
- 5 quant attribs
- median: central line
- lower and upper quartile: boxes
- lower upper fences: whiskers
- values beyond which items are outliers
- outliers beyond fence cutoffs explicitly shown



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

Idiom: Hierarchical parallel coordinates

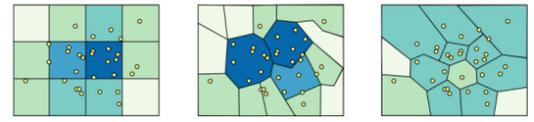
- dynamic item aggregation
- derived data: hierarchical clustering
- encoding:
- cluster band with variable transparency, line at mean, width by min/max values
- color by proximity in hierarchy



[Hierarchical Parallel Coordinates for Exploration of Large Datasets. Fua, Ward, and Rundensteiner. Proc. IEEE Visualization Conference (Vis '99), pp. 43–50, 1999.]

Spatial aggregation

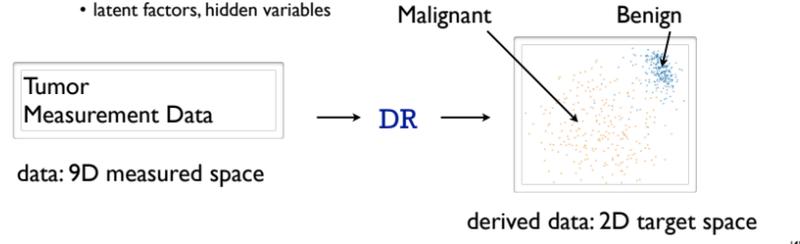
- MAUP: Modifiable Areal Unit Problem
 - gerrymandering (manipulating voting district boundaries) is one example!



[http://www.e-education.psu.edu/geog486/14_p7.html, Fig 4.cg.6]

Dimensionality reduction

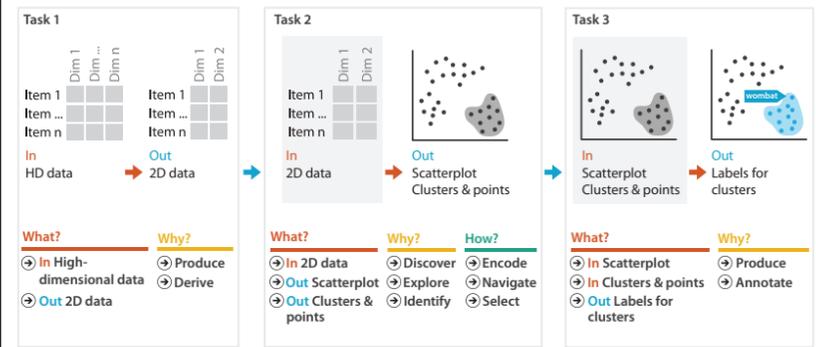
- attribute aggregation
 - derive low-dimensional target space from high-dimensional measured space
 - use when you can't directly measure what you care about
 - true dimensionality of dataset conjectured to be smaller than dimensionality of measurements
 - latent factors, hidden variables



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Idiom: Dimensionality reduction for documents



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Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014.
 - Chap 13: Reduce Items and Attributes
- Hierarchical Aggregation for Information Visualization: Overview, Techniques and Design Guidelines. Elmqvist and Fekete. IEEE Transactions on Visualization and Computer Graphics 16:3 (2010), 439–454.
- A Review of Overview+Detail, Zooming, and Focus+Context Interfaces. Cockburn, Karlson, and Bederson. ACM Computing Surveys 41:1 (2008), 1–31.
- A Guide to Visual Multi-Level Interface Design From Synthesis of Empirical Study Evidence. Lam and Munzner. Synthesis Lectures on Visualization Series, Morgan Claypool, 2010.

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Outline

- Session 1 8:30-10:00am Visualization Analysis Framework
 - Introduction: Definitions
 - Analysis: What, Why, How
 - Marks and Channels
- Session 2 10:30am-12:00pm Spatial Layout
 - Arrange Tables
 - Arrange Spatial Data
 - Arrange Networks and Trees
- Session 3 1:00-2:30pm Color & Interaction
 - Map Color
 - Manipulate: Change, Select, Navigate
 - Facet: Juxtapose, Partition, Superimpose
- Session 4 3:00-4:30pm Guidelines and Examples
 - Reduce: Filter, Aggregate
 - Rules of Thumb
 - Q&A

http://www.cs.ubc.ca/~tmm/talks.html#vad16act

@tamaramunzner

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Rules of Thumb

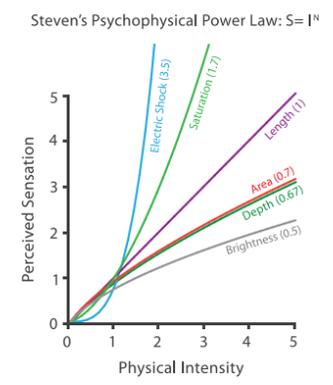
- No unjustified 3D
 - Power of the plane
 - Disparity of depth
 - Occlusion hides information
 - Perspective distortion dangers
 - Tilted text isn't legible
- No unjustified 2D
- Eyes beat memory
- Resolution over immersion
- Overview first, zoom and filter, details on demand
- Responsiveness is required
- Function first, form next

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No unjustified 3D: Power of the plane

- high-ranked spatial position channels: planar spatial position
 - not depth!

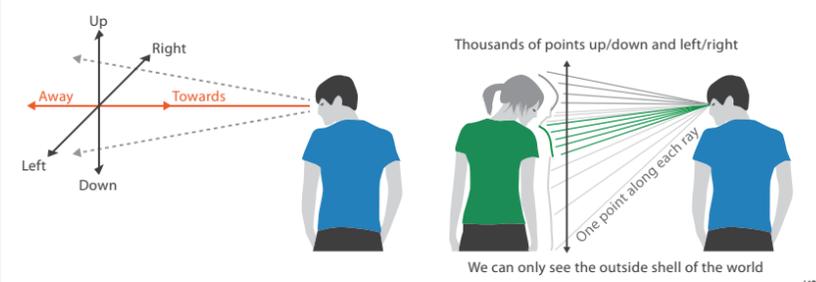
- Magnitude Channels: Ordered Attributes
 - Position on common scale
 - Position on unaligned scale
 - Length (1D size)
 - Tilt/angle
 - Area (2D size)
 - Depth (3D position)



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No unjustified 3D: Danger of depth

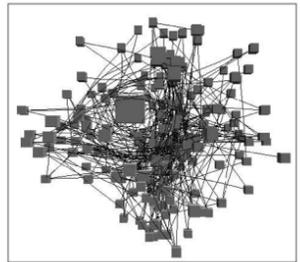
- we don't really live in 3D: we see in 2.05D
 - acquire more info on image plane quickly from eye movements
 - acquire more info for depth slower, from head/body motion



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Occlusion hides information

- occlusion
- interaction complexity



[Distortion Viewing Techniques for 3D Data. Carpendale et al. InfoVis 1996.]

Perspective distortion loses information

- perspective distortion
 - interferes with all size channel encodings
 - power of the plane is lost!

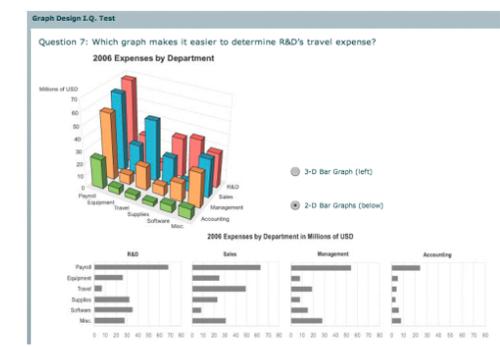


[Visualizing the Results of Multimedia Web Search Engines. Mukherjee, Hirata, and Hara. InfoVis 96]

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3D vs 2D bar charts

- 3D bars never a good idea!

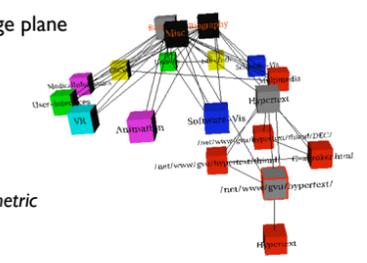


[http://perceptualedge.com/files/GraphDesignIQ.html]

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Tilted text isn't legible

- text legibility
 - far worse when tilted from image plane



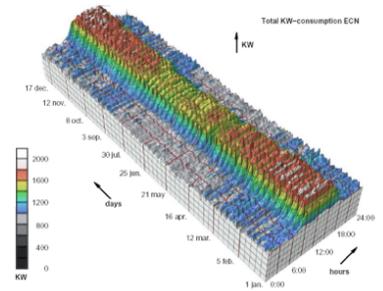
[Exploring and Reducing the Effects of Orientation on Text Readability in Volumetric Displays. Grossman et al. CHI 2007]

[Visualizing the World-Wide Web with the Navigational View Builder. Mukherjee and Foley. Computer Networks and ISDN Systems, 1995.]

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No unjustified 3D example: Time-series data

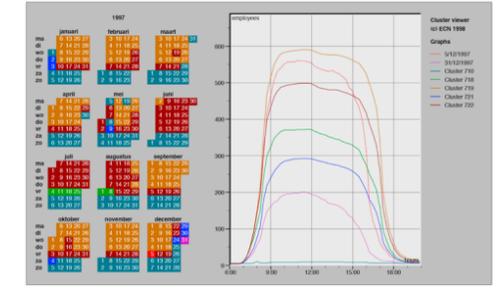
- extruded curves: detailed comparisons impossible



[Cluster and Calendar based Visualization of Time Series Data. van Wijk and van Selow, Proc. InfoVis 99.]

No unjustified 3D example: Transform for new data abstraction

- derived data: cluster hierarchy
- juxtapose multiple views: calendar, superimposed 2D curves

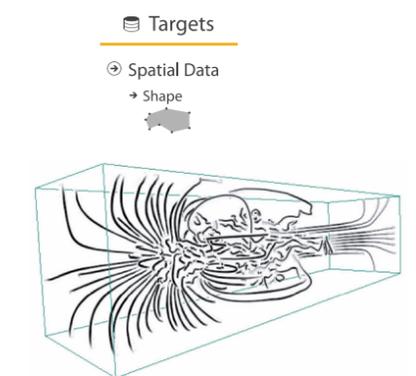


[Cluster and Calendar based Visualization of Time Series Data. van Wijk and van Selow, Proc. InfoVis 99.]

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Justified 3D: shape perception

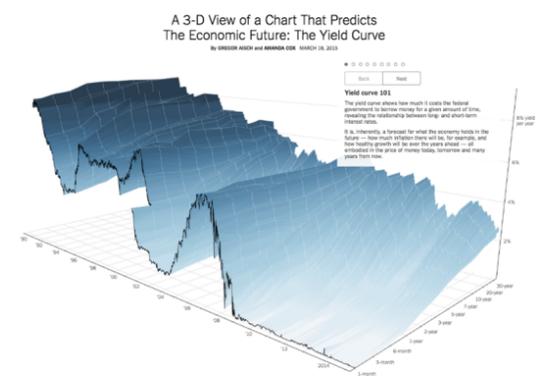
- benefits outweigh costs when task is shape perception for 3D spatial data
 - interactive navigation supports synthesis across many viewpoints



[Image-Based Streamline Generation and Rendering. Li and Shen. IEEE Trans. Visualization and Computer Graphics (TVCG) 13:3 (2007), 630–640.]

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Justified 3D: Economic growth curve

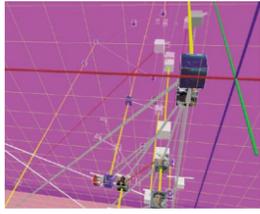


http://www.nytimes.com/interactive/2015/03/19/upshot/3d-yield-curve-economic-growth.html

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No unjustified 3D

- 3D legitimate for true 3D spatial data
- 3D needs very careful justification for abstract data
 - enthusiasm in 1990s, but now skepticism
 - be especially careful with 3D for point clouds or networks

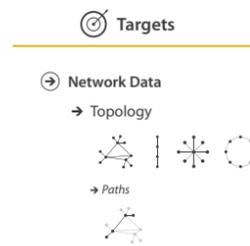


[WEBPATH—a three dimensional Web history. Frecon and Smith. Proc. InfoVis 1999]

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No unjustified 2D

- consider whether network data requires 2D spatial layout
 - especially if reading text is central to task!
 - arranging as network means lower information density and harder label lookup compared to text lists
- benefits outweigh costs when topological structure/context important for task
 - be especially careful for search results, document collections, ontologies



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Eyes beat memory

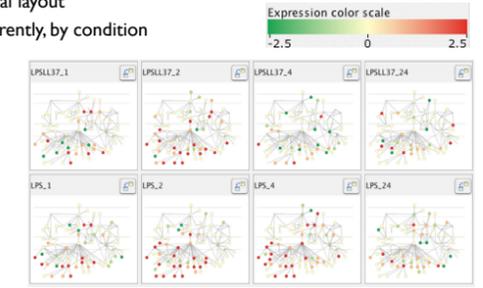
- principle: external cognition vs. internal memory
 - easy to compare by moving eyes between side-by-side views
 - harder to compare visible item to memory of what you saw
- implications for animation
 - great for choreographed storytelling
 - great for transitions between two states
 - poor for many states with changes everywhere
 - consider small multiples instead



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Eyes beat memory example: Cerebral

- small multiples: one graph instance per experimental condition
 - same spatial layout
 - color differently, by condition

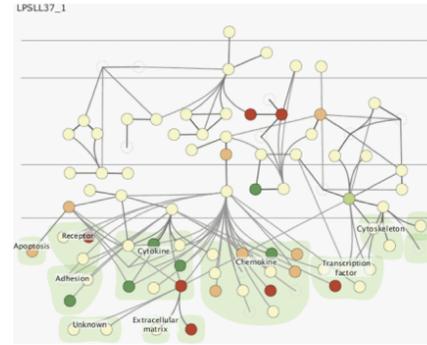


[Cerebral: Visualizing Multiple Experimental Conditions on a Graph with Biological Context. Barsky, Munzner, Gardy, and Kincaid. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2008) 14:6 (2008), 1253–1260.]

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Why not animation?

- disparate frames and regions: comparison difficult
 - vs contiguous frames
 - vs small region
 - vs coherent motion of group
- safe special case
 - animated transitions



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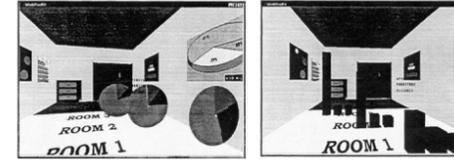
Change blindness

- if attention is directed elsewhere, even drastic changes not noticeable
 - door experiment
- change blindness demos
 - mask in between images

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Resolution beats immersion

- immersion typically not helpful for abstract data
 - do not need sense of presence or stereoscopic 3D
- resolution much more important
 - pixels are the scarcest resource
 - desktop also better for workflow integration
- virtual reality for abstract data very difficult to justify



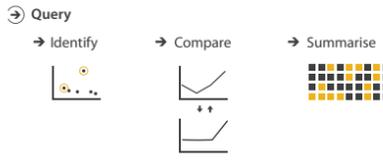
[Development of an information visualization tool using virtual reality. Kirner and Martins. Proc. Symp. Applied Computing 2000]

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Overview first, zoom and filter, details on demand

- influential mantra from Shneiderman
 - The Eyes Have It: A Task by Data Type Taxonomy for Information Visualizations. Shneiderman. Proc. IEEE Visual Languages, pp. 336–343, 1996.]

- overview = summary
 - microcosm of full vis design problem



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Responsiveness is required

- three major categories
 - 0.1 seconds: perceptual processing
 - 1 second: immediate response
 - 10 seconds: brief tasks
- importance of visual feedback

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Function first, form next

- start with focus on functionality
 - straightforward to improve aesthetics later on, as refinement
 - if no expertise in-house, find good graphic designer to work with
- dangerous to start with aesthetics
 - usually impossible to add function retroactively

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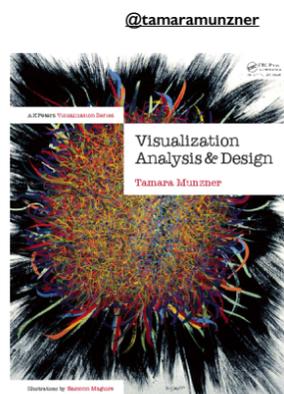
Further reading

- Visualization Analysis and Design. Tamara Munzner. CRC Press, 2014.
 - Chap 6: Rules of Thumb
- Designing with the Mind in Mind: Simple Guide to Understanding User Interface Design Rules. Jeff Johnson. Morgan Kaufmann, 2010.
 - Chap 12: We Have Time Requirements

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More Information

- this talk
 - <http://www.cs.ubc.ca/~tmm/talks.html#vad16act>
- 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, courses
 - <http://www.cs.ubc.ca/group/infovis>
 - <http://www.cs.ubc.ca/~tmm>



Visualization Analysis and Design. Munzner. A K Peters Visualization Series, CRC Press, Visualization Series, 2014.

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