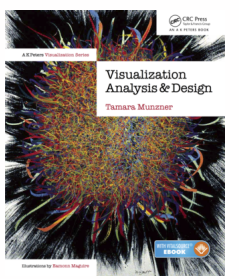


Visualization Analysis & Design



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Graphics Interface 2016 Invited Talk
June 2 2016, Victoria BC

<http://www.cs.ubc.ca/~tmm/talks.html#vad16gi>

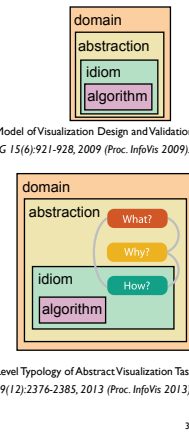
@tamaramunzner

Why talk about a textbook to a room of experts?

- many folks here in graphics or HCI, but few in visualization
 - my own roots in graphics, later added HCI quant methods, then HCI qual methods
- convince you of the value in thinking systematically about vis design
 - decompose into comprehensive framework of principles and design choices
 - situate specific examples within framework as concrete illustrations
- provide unified view that crosscuts entire field of visualization
 - infovis and scivis: addressing different kinds of data
 - visual analytics: interweave data analysis & transformation w/ interactive visual exploration
 - caveat: my own background in infovis shines through!

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



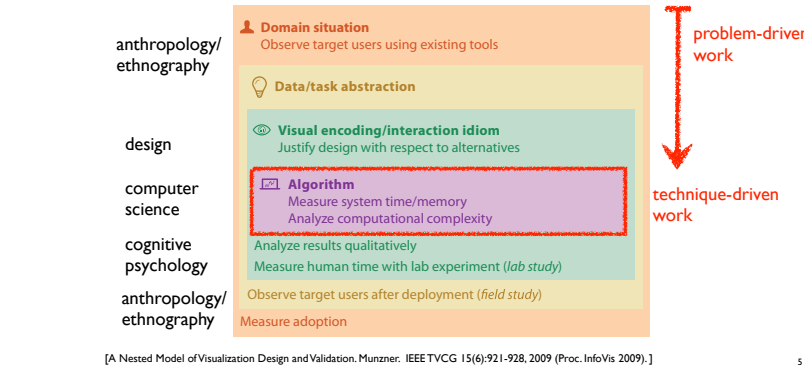
Why is validation difficult?

- different ways to get it wrong at each level



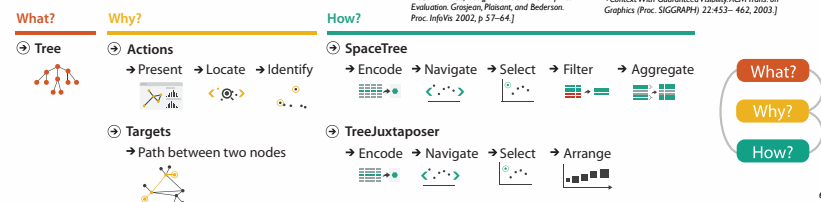
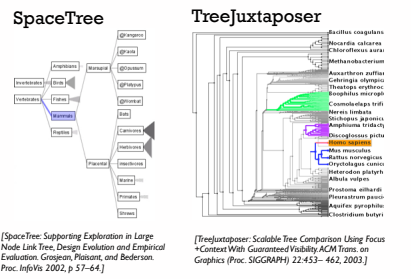
Why is validation difficult?

- solution: use methods from different fields at each level

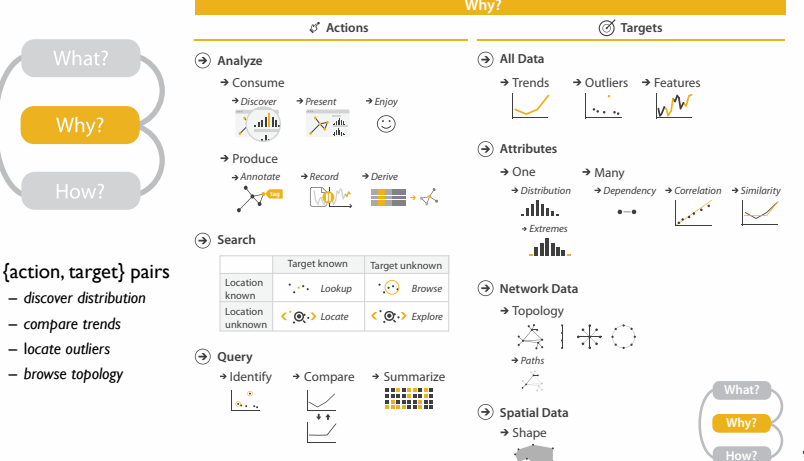
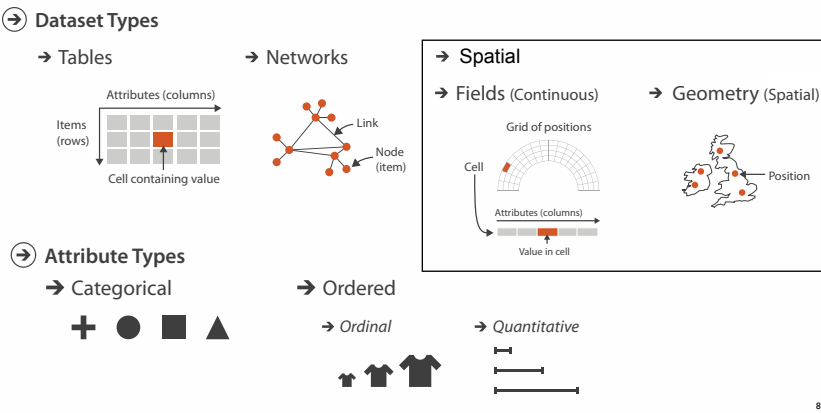


Why analyze?

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



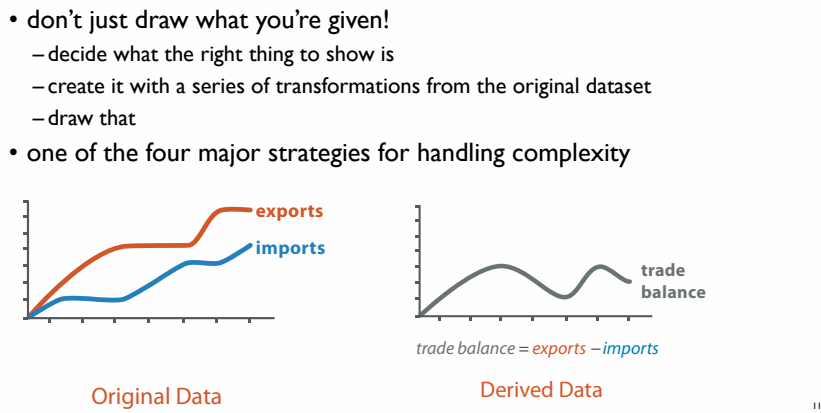
Types: Datasets and data



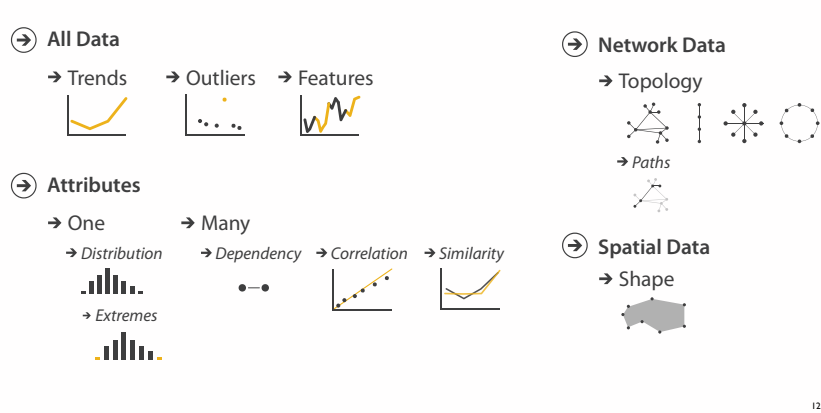
Actions: Analyze, Query



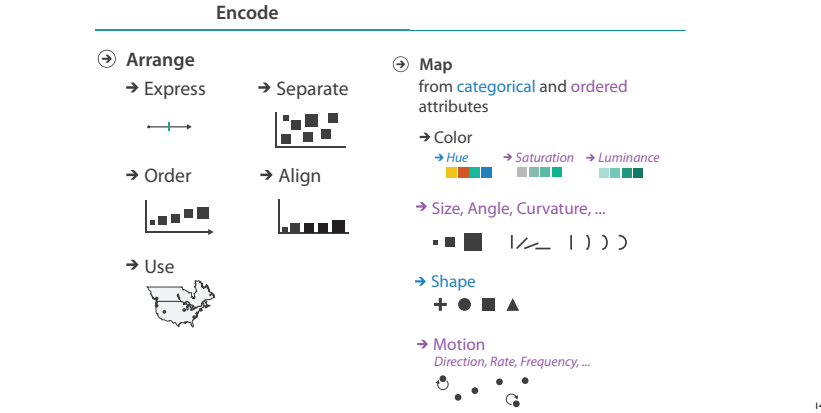
Derive: Crucial Design Choice



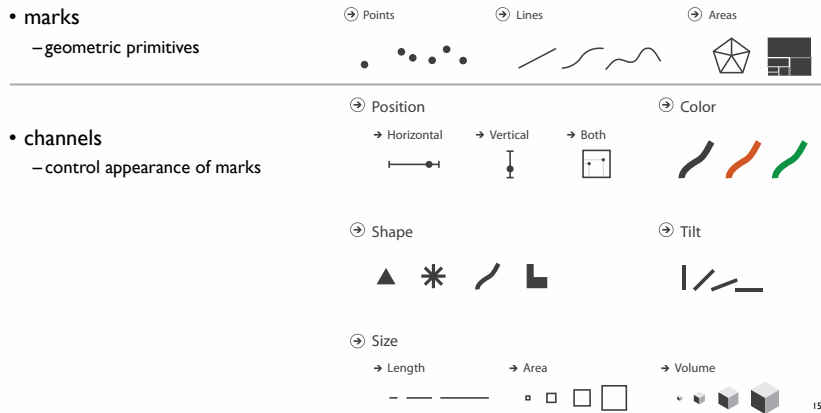
Targets



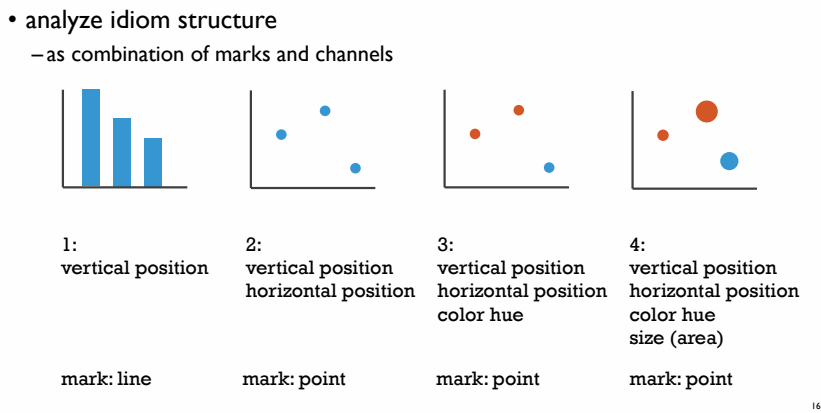
How to encode: Arrange space, map channels



Definitions: Marks and channels



Encoding visually with marks and channels



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)

Channels: Matching Types

⊕ **Magnitude Channels: Ordered Attributes**

⊕ **Identity Channels: Categorical Attributes**

• expressiveness principle
– match channel and data characteristics

Channels: Rankings

⊕ **Magnitude Channels: Ordered Attributes**

⊕ **Identity Channels: Categorical Attributes**

• expressiveness principle
– match channel and data characteristics

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

How?

Encode

Manipulate

Facet

Reduce

What?

Why?

How?

How to handle complexity: 3 more strategies + 1 previous

Manipulate

Facet

Reduce

Derive

Change

Juxtapose

Filter

Select

Partition

Aggregate

Navigate

Superimpose

Embed

• 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

Change

Juxtapose

Filter

Select

Partition

Aggregate

Navigate

Superimpose

Embed

• change over time

– most obvious & flexible of the 4 strategies

How to handle complexity: 3 more strategies + 1 previous

Manipulate

Facet

Reduce

Derive

Change

Juxtapose

Filter

Select

Partition

Aggregate

Navigate

Superimpose

Embed

• facet data across multiple views

Idiom: Linked highlighting System: EDV

• see how regions contiguous in one view are distributed within another

– powerful and pervasive interaction idiom

• encoding: different

• 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

How to handle complexity: 3 more strategies + 1 previous

Manipulate

Facet

Reduce

Derive

Change

Juxtapose

Filter

Select

Partition

Aggregate

Navigate

Superimpose

Embed

• reduce what is shown within single view

Reduce items and attributes

Reducing Items and Attributes

Filter

Aggregate

Embed

• 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

Idiom: boxplot

• static item aggregation

• task: find distribution

• data: table

• derived data

– 5 quant attrbs

• 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

Task 2

Task 3

What? Why? How?

What? Why? How? domain abstraction

What?

Why?

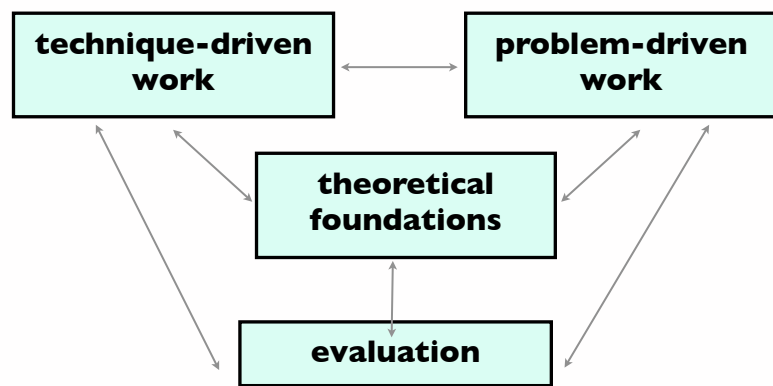
How?

domain abstraction

idiom

algorithm

A quick taste of my own work!



Technique-driven: Graph drawing

James Slack, Kristian Hildebrand, Daniel Archambault, David Auber (Bordeaux), TopoLayout, SPF, Grouse, GrouseFlocks, TugGraph

Evaluation: Graph drawing

Dmitry Nekrasovski, Adam Bodnar, Joanna McGrenere (UBC), Jessica Dawson, Joanna McGrenere (UBC), Stretch and squish navigation, Search set model of path tracing

Technique-driven: Dimensionality reduction

Stephen Ingram, Glimmer, DimStiller, GInt, QSNE

Evaluation: Dimensionality reduction

Melanie Tory, Michael Sedlmair, Melanie Tory (UVic), Taxonomy of cluster separation factors, Points vs landscapes for dimensionally reduced data, Guidance on DR & scatterplot choices

Problem-driven: Genomics

Aaron Barsky, Jenn Gardy (Microbio), Robert Kincaid (Agilent), Hanspeter Pfister (Harvard), Cerebral, Miriah Meyer, MizBee, MulteeSum, Pathline

Problem-driven: Genomics, fisheries

Joel Ferstay, Cydney Nielsen (BC Cancer), Variant View, Maryam Booshehrian, Torsten Moeller (SFU), Vismon

Problem-driven: Many domains

Heidi Lam, Diane Tang (Google), SessionViewer: web log analysis, Peter McLachlan, Stephen North (AT&T Research), LiveRAC: systems time-series

Evaluation: Focus+Context

Heidi Lam, Ron Rensink (UBC), Distortion impact on search/memory, Heidi Lam, Robert Kincaid (Agilent), Separate vs integrated views

Journalism

Matt Brehmer, Stephen Ingram, Jonathan Stray (Assoc Press), Overview, Johanna Fulda (Sud. Zeitung), Matt Brehmer, TimeLineCurator

Theoretical foundations

Visual Encoding Pitfalls, Strategy Pitfalls, Papers Process & Pitfalls, Design Study Methodology, Abstract Tasks, Nested Model, Michael Sedlmair, Miriah Meyer, Matt Brehmer

More Information

@tamaramunzner, this talk, book page (including tutorial lecture slides), illustrations: Eamonn Maguire, Visualization Analysis & Design, Tamara Munzner