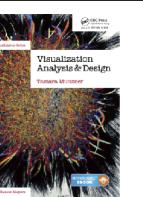


Visualization Analysis & Design



Tamara Munzner

Department of Computer Science
University of British Columbia

UBC Alumni/Industry Lecture
Feb 27 2020, Vancouver BC

<http://www.cs.ubc.ca/~tmm/talks.html#vad20alum>

Why is validation difficult?

- different ways to get it wrong at each level



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Visualization: definition & motivation

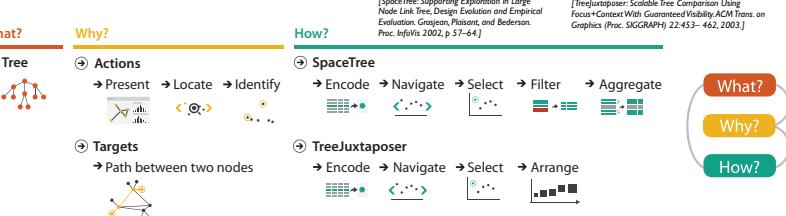
- Computer-based visualization systems provide visual representations of datasets designed to help people carry out tasks more effectively.
- Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.
- human in the loop needs the details & no trusted automatic solution exists
 - doesn't know exactly what questions to ask in advance
 - exploratory data analysis
 - speed up** through human-in-the-loop visual data analysis
 - present known results to others
 - stepping stone towards automation
 - before model creation to provide understanding
 - during algorithm creation to refine, debug, set parameters
 - before or during deployment to build trust and monitor

more at:
[Visualization Analysis and Design. Munzner. CRC Press, 2014.](http://www.cs.ubc.ca/~tmm/talks.html#vad20alum)

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Why analyze?

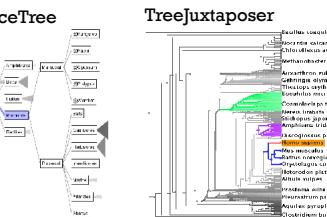
- imposes a structure on huge design space
- scaffold to help you think systematically about choices
- analyzing existing as stepping stone to designing new



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SpaceTree



[SpaceTree: Supporting Exploration in Large Node Link Trees. Design Evolution and Empirical Evaluation. Grasman, Plaisant, and Bederson. Proc. InfoVis 2009, p 57-64.]

[TreeJuxtaposer: Scalable Tree Comparison Using Focus+Context With Guaranteed Visibility. ACM Trans. on Graphics (Proc. SIGGRAPH) 22:453–462, 2003.]

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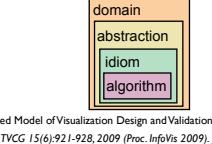
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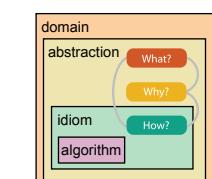
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Analysis framework: Four levels, three questions

- 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

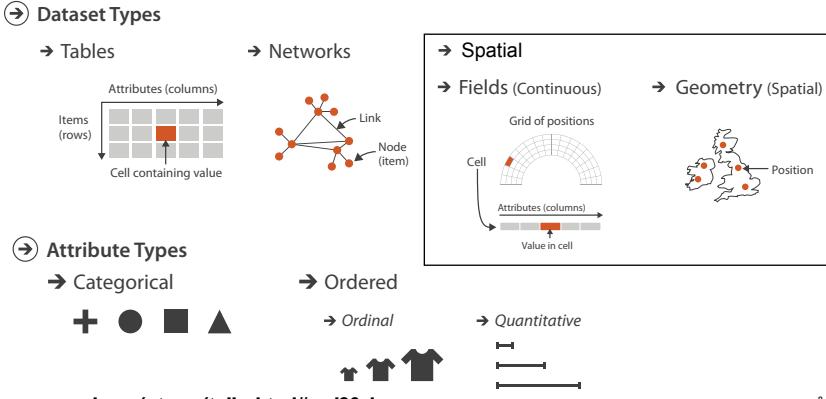


[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]



[A Multi-Level Typology of Abstract Visualization Tasks. Brehmer and Munzner. IEEE TVCG 19(12):2373-2385, 2013 (Proc. InfoVis 2013).]

Types: Datasets and data



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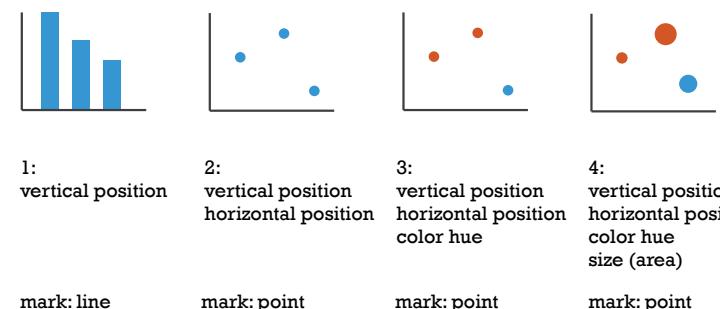
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Encoding visually with marks and channels

- analyze idiom structure
- as combination of marks and channels



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Channels

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

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Channels: Matching Types

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

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- expressiveness principle
- match channel and data characteristics

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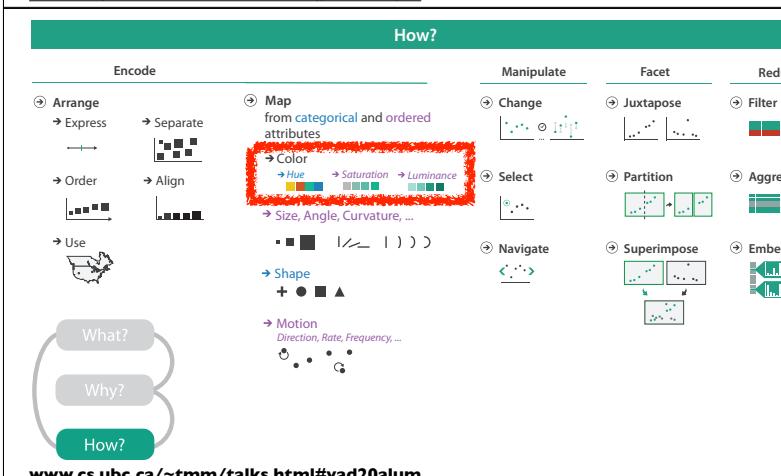
Channels: Rankings

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

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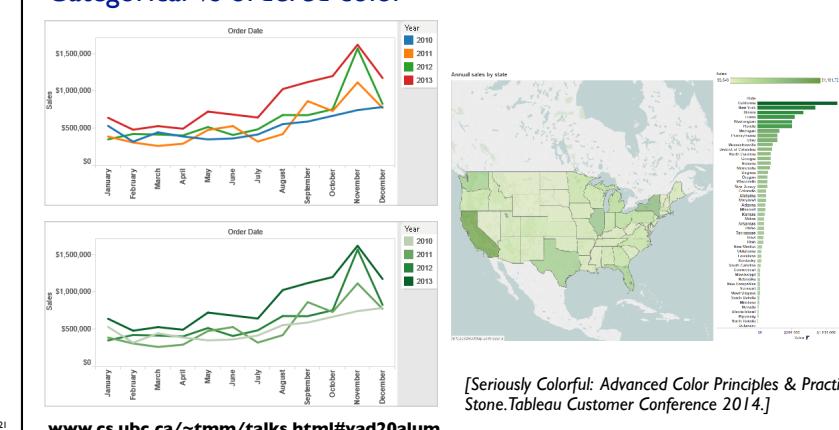
- expressiveness
- match channel and data characteristics
- effectiveness
- channels differ in accuracy of perception
- distinguishability
- match available levels in channel w/ data

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Categorical vs ordered color

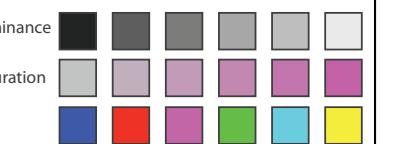


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Decomposing color

- first rule of color: do not talk about color!
 - color is confusing if treated as monolithic
- decompose into three channels
 - ordered can show magnitude
 - luminance: how bright
 - saturation: how colorful
 - categorical can show identity
 - hue: what color
 - caveat: not well supported by current tools
- channels have different properties
 - what they convey directly to perceptual system
 - how much they can convey: how many discriminable bins can we use?

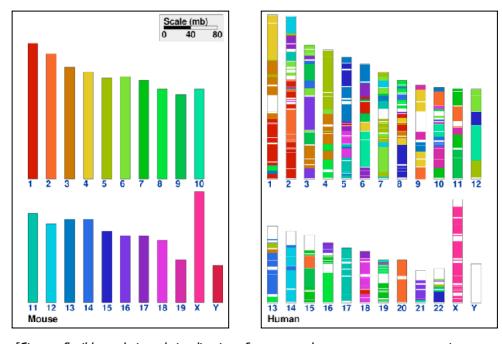
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Categorical color: limited number of discriminable bins

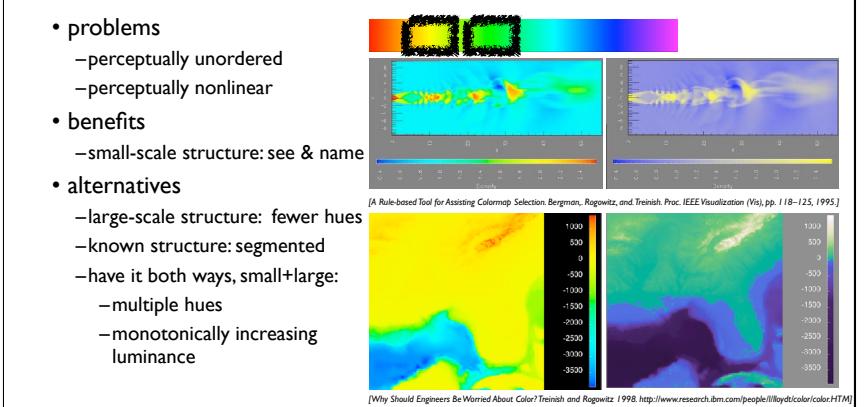
- human perception built on relative comparisons
 - great if color contiguous
 - surprisingly bad for absolute comparisons
- noncontiguous small regions of color
 - fewer bins than you want
 - rule of thumb: 6-12 bins, including background and highlights
- alternatives? other talks!

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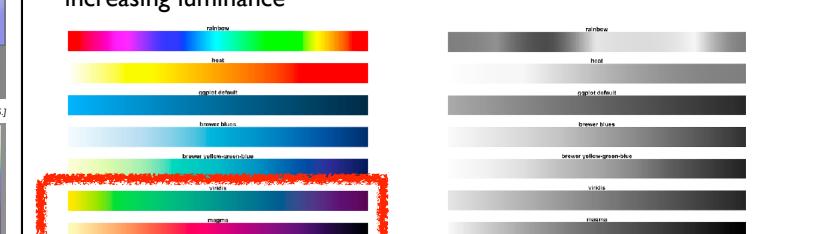
Ordered color: Rainbow is poor default



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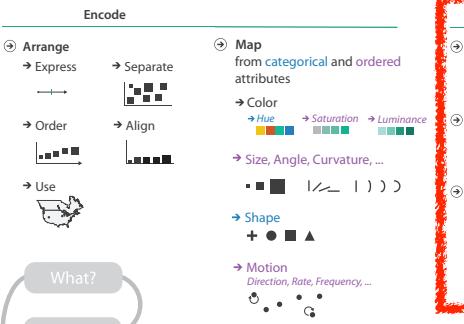
Viridis / Magma

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



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How?



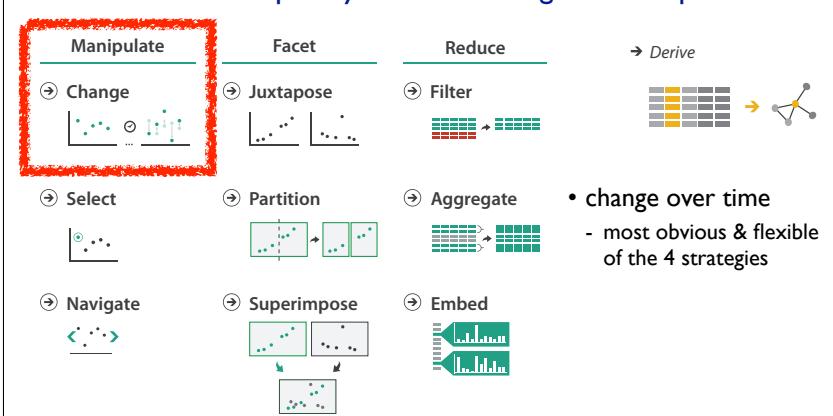
www.cs.ubc.ca/~tmm/talks.html#vad20alum

How to handle complexity: 3 more strategies + 1 previous

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

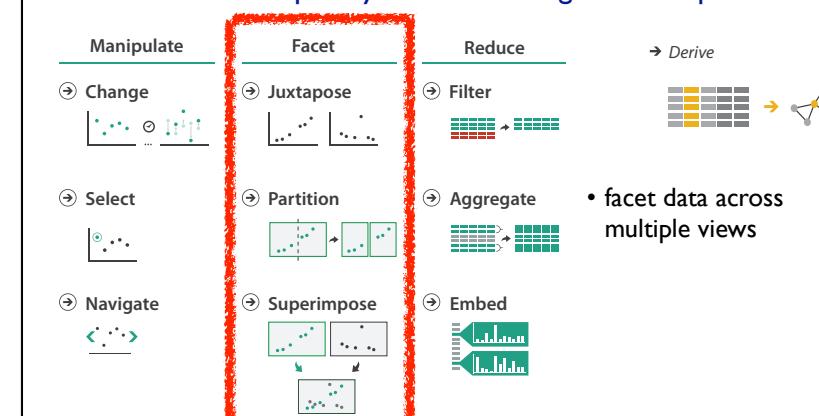
www.cs.ubc.ca/~tmm/talks.html#vad20alum

How to handle complexity: 3 more strategies + 1 previous



www.cs.ubc.ca/~tmm/talks.html#vad20alum

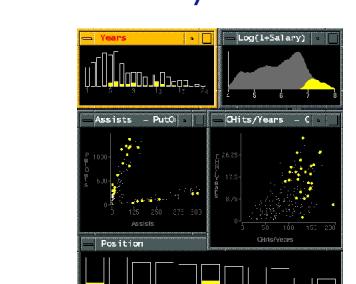
How to handle complexity: 3 more strategies + 1 previous



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Idiom: Linked highlighting

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



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Idiom: bird's-eye maps

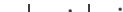
- encoding: same
- data: subset shared
- navigation: shared
 - bidirectional linking
- differences
 - viewpoint
 - (size)
- overview-detail



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		Data		
		All	Subset	None
Encoding	Same	Redundant		Overview/ Detail 
	Different			Multiform, Overview/ Detail 
<ul style="list-style-type: none"> why juxtapose views? 		<ul style="list-style-type: none"> –benefits: eyes vs memory <ul style="list-style-type: none"> lower cognitive load to move eyes between 2 views than remembering previous state with single changing view 		
<ul style="list-style-type: none"> –costs: display area, 2 views side by side each have only half the area of one view 		www.cs.ubc.ca/~tmml/talks.html#yad20alum		

- weaknesses
 - widespread changes
 - disparate frames
- strengths
 - choreographed storytelling
 - localized differences between contiguous frames
 - animated transitions between states

How to handle complexity: 3 more strategies + 1 previous			
Manipulate	Facet	Reduce	Derive
 Change	 Juxtapose	 Filter	 Derive
 Select	 Partition	 Aggregate	<ul style="list-style-type: none"> reduce what is shown within single view
 Navigate	 Superimpose	 Embed	

Reduce items and attributes

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

Reducing Items and Attributes

- ④ Filter
 - Items
 -
 - Attributes
 -
- ④ Aggregate
 - Items
 -
 - Attributes
 -
- ④ Embed

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Idiom: boxplot

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

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

Idiom: Dimensionality reduction for documents

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

The diagram illustrates the dimensionality reduction process for documents across three tasks:

- Task 1:** Shows a high-dimensional space (HD data) with items represented as grids of dimensions Dim 1, Dim 2, ..., Dim n. An arrow labeled "Out 2D data" points to a 2D representation where each item is a single point.
- Task 2:** Shows a 2D space with items represented as grids of dimensions Dim 1, Dim 2, ..., Dim n. An arrow labeled "Out Scatterplot Clusters & points" points to a scatterplot where items are grouped into clusters and individual points.
- Task 3:** Shows a scatterplot with items represented as clusters and points. An arrow labeled "Out Labels for clusters" points to a final state where labels are assigned to the clusters.

What? **Why?**

⊕ In High-dimensional data	⊕ Produce 2D data
⊕ Out 2D data	⊕ Derive

What?	Why?	How?
⊕ In 2D data	⊕ Discover	⊕ Encode
⊕ Out Scatterplot	⊕ Explore	⊕ Navigate
⊕ Out Clusters & points	⊕ Identify	⊕ Select

What?	Why?
⊕ In Scatterplot	⊕ Produce Labels for clusters
⊕ In Clusters & points	⊕ Annotate
⊕ Out Labels for clusters	

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The diagram illustrates the Tableau Data Science Toolkit, organized into several main sections:

- What?** (Top Bar)
 - Datasets**: Data Types → Items
 - Attributes**
- Why?** (Yellow Bar)
 - Actions**: Analyze (Discover, Present, Enjoy), Produce (Annotate)
 - Targets**: All Data (Trends, Outliers, Features)
- How?** (Green Bar)
 - Encode**: Arrange (Express, Separate, Align), Search (Order, Align), Use (Location known, Location unknown)
 - Manipulate**: Map (from categorical and ordered attributes, Color, Select, Size, Angle, Curvature...), Change (Hue, Saturation, Luminance), Navigate (Shape, Motion)
 - Facet**: Juxtapose, Partition
 - Reduce**: Filter, Aggregate, Superimpose, Embed
- Domain Abstraction** (Right Panel):
 - What? (Yellow box)
 - Why? (Blue box)
 - How? (Green box)
 - Idiom (Orange box)
 - Algorithm (Purple box)

```
graph TD; A[technique-driven work] <--> B[problem-driven work]; C[theoretical foundations] <--> D[evaluation]; A --> C; A --> D; C --> B; D --> B;
```

Technique-driven: Graph/network drawing

Daniel Archambault

David Auber (Bordeaux)

Benjamin Renoust

Guy Melançon (Bordeaux)

TopoLayout
SPF
Grouse
GrouseFlocks
TugGraph

Detangler

<https://youtu.be/AWXAe8zvkt8>

<https://youtu.be/QOInHSSuUV6k>

Technique-driven: Tree drawing

Zipeng Liu Shing Hei Chan



Aggregated Dendograms
<https://youtu.be/2SLcz7KNLJw>

TreeJuxtaposer
<https://youtu.be/GdaPj8a9QEO>

Evaluation experiments: Graph/tree drawing

T F P
E

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

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T
F
E

Technique-driven: Dimensionality reduction

Stephen Ingram

Glimmer

Dense Matrix DS

$\nabla_i \rightarrow \nabla_{i+1}$

Layout Points M

Check Convergence S

Glim Outer Loop

DimStiller

QSNE

Evaluation in the field: Dimensionality reduction

Dimensional Reduction: Dimensional Synthesis

- Input:**
- Output:**
- Process:**

Input: **Output:**

Name Synthesis: Dimensional Synthesis

- Input:**
- Output:**
- Process:**

Input: **Output:**

Map Synthesis: Dimensional to Original Dimensions

- Input:**
- Output:**
- Process:**

Input: **Output:**

DR in the Wild

Dimensional Reduction: Dimensional Synthesis

- Input:**
- Output:**
- Process:**

Input: **Output:**

Name Synthesis: Dimensional Synthesis

- Input:**
- Output:**
- Process:**

Input: **Output:**

Map Synthesis: Dimensional to Original Dimensions

- Input:**
- Output:**
- Process:**

Input: **Output:**

Match Clusters and Classes

- Input:**
- Output:**
- Process:**

Input: **Output:**

DR in the Wild

Willy Clusters

- Input:**
- Output:**
- Process:**

Input: **Output:**

Name Clusters

- Input:**
- Output:**
- Process:**

Input: **Output:**

Match Clusters and Classes

- Input:**
- Output:**
- Process:**

Input: **Output:**

Matt Brehmer Michael Sedlmair Melanie Tory Stephen Ingram

Problem-driven: Genomics, fisheries

T P F E

Variant View
https://youtu.be/AHDnv_qMXxQ

Vismon
<https://youtu.be/h0kHoS4VYmk>

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Problem-driven: Tech industry

T P F E

SessionViewer: web log analysis
<https://youtu.be/T4MaT2d56G4>

LiveRAC: systems time-series
<https://youtu.be/l0c3H0VSkw>

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Problem-driven: Building energy mgmt, journalism

T P F E

redesign success: industrial swdev resources committed

Energy Manager

Overview
<https://vimeo.com/71483614>

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Curation & Presentation: Timelines

T P F E

TimeLineCurator
<https://vimeo.com/123246662>

Timelines Revisited
<https://timelinesrevisited.github.io/>

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Matt Brehmer **Diane Tang** **Heidi Lam** **Peter McLachlan** **Stephen North** **Torsten Moeller** **Maryam Booshehri** **Stephen Ingram** **Jonathan Stray** **Kevin Tate** **Johanna Fulda**

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Problem-driven: Current data science

T P F E

Kimberly Dextras-Romagnino **recent work: Segmentifier (Mobify)**

recent work: Segmentifier (Mobify)

e-commerce clickstreams

build tools for human-in-the-loop visual data analysis

https://youtu.be/TobYDFeiSOg

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Michael Oppermann **recent work: Ocupado (Sensible Building Science)**

wifi proxy for real-time building occupancy

visual analytics for facilities management

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Theoretical foundations: Typologies

T P F E

Matt Brehmer

Anamaria Crisan

Abstract Tasks

GEViT: Genomic Epidemiology Visualization Typology

Regulatory & Organizational Constraints

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Theoretical foundations

T P F E

Visual Encoding Pitfalls

- Unjustified Visual Encoding
- Hammer In Search Of Nail
- 2D Good, 3D Better
- Color Cacophony
- Rainbows Just Like In The Sky

Strategy Pitfalls

- What I Did Over My Summer
- Least Publishable Unit
- Dense As Plutonium
- Bad Slice and Dice

Papers Process & Pitfalls

Design Study Methodology

Michael Sedlmaier **Miriah Meyer**

domain abstraction idiom algorithm

Nested Model

Visualization Analysis & Design

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More Information

@tamaramunzner

- this talk
<http://www.cs.ubc.ca/~tmm/talks.html#vad20alum>
- 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>

www.cs.ubc.ca/~tmm/talks.html#vad20alum

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