

Lecture 1: Intro, Data and Tasks Marks and Channels

Tamara Munzner
Department of Computer Science
University of British Columbia

DSCI 531: Data Visualization I
Lecture 1: 16 November 2016

https://github.ubc.ca/ubc-mds-2016/DSCI_531_viz-I_students

What's when

- 8 lectures in 4 weeks
 - Wed & Mon, 11am-12:20pm (80 min), Nov 16 - Dec 12, SPPH 143
- 4 labs
 - Thu, 2-4pm, Nov 17 - Dec 8, ESB 1042
 - start work Thu 2pm, due next Wed 9am, 12.5% each
- 2 quizzes: Week 3 (Dec 1) & week 5 (Dec 15)
 - Thu 2-2:30pm, 25% each
- my (optional) office hrs are in ICICS/CS X661
 - Thu Nov 17, 5-6pm
 - Thu Nov 24, 5-6pm
 - Thu Dec 1, 5-6pm
 - Wed Dec 7, 6-7pm (note outer building doors close at 6:30)

Reading

- core foundational material covered in lectures
- textbook as backup to lectures
 - Tamara Munzner: Visualization Analysis and Design. CRC Press, 2014.
 - library has multiple ebook copies for free
 - to buy yourself, see <http://www.cs.ubc.ca/~tmm/yadbook/>

Topics

- Lecture 1
 - Intro, Data and Tasks
 - Marks and Channels
- Lecture 2
 - In-Class Vis Design Exercise
- Lecture 3
 - Arrange Table Data, part I
- Lecture 4
 - Arrange Table Data, part I
- Lecture 5
 - Arrange Spatial Data
- Lecture 6
 - Color
- Lecture 7
 - Arrange Network Data
- Lecture 8
 - Rules of Thumb
 - Graphic Design Principles

Introduction: Defining visualization (vis)

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

Why?...

Why have a human in the loop?

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

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

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

Why use an external representation?

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

- external representation: replace cognition with perception



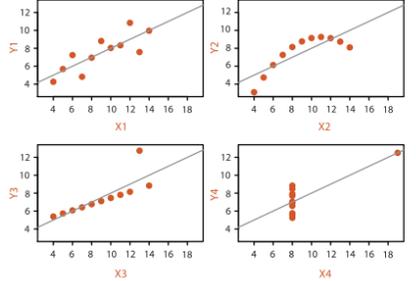
Why depend on vision?

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

Why show the data in detail?

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



Why focus on tasks and effectiveness?

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

- tasks serve as constraint on design (as does data)
 - idioms do not serve all tasks equally!
 - challenge: recast tasks from domain-specific vocabulary to abstract forms
- most possibilities ineffective
 - validation is necessary, but tricky
 - increases chance of finding good solutions if you understand full space of possibilities
- what counts as effective?
 - novel: enable entirely new kinds of analysis
 - faster: speed up existing workflows

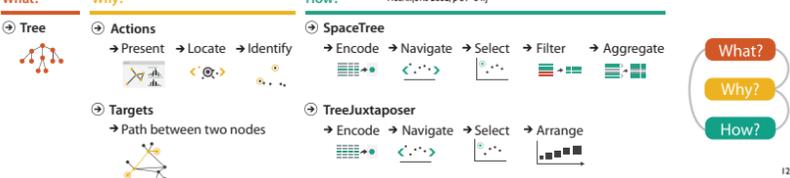
What resource limitations are we faced with?

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

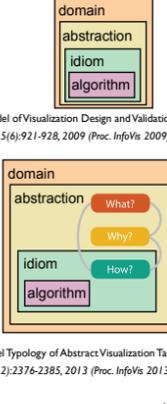
Why analyze?

- 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



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



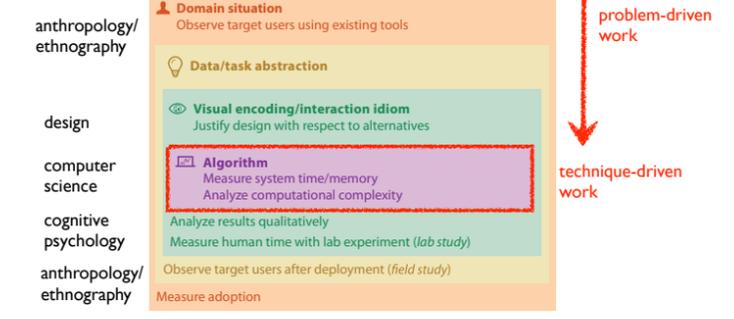
Why is validation difficult?

- different ways to get it wrong at each level

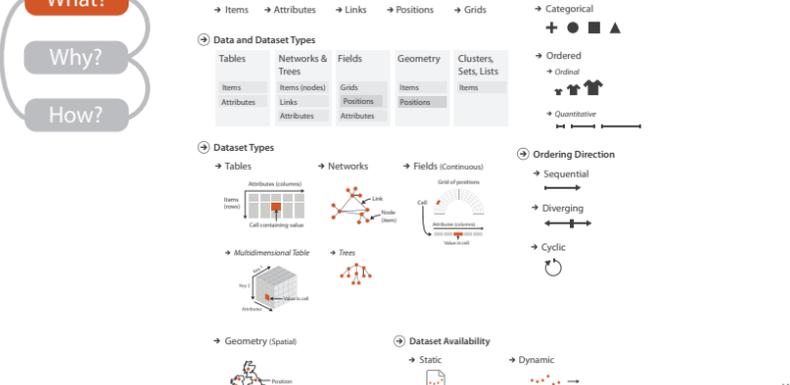


Why is validation difficult?

- solution: use methods from different fields at each level



What? Why? How?



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

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.]

Task 1: In Tree → Out Quantitative attribute on nodes

Task 2: In Tree + In Quantitative attribute on nodes → Out Filtered Tree (Removed unimportant parts)

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, Order, Use Map: Color (Hue, Saturation, Luminance), Size, Angle, Curvature, Shape, Motion (Direction, Rate, Frequency) 	<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

Viz-1, Viz-2

Encoding visually

- analyze idiom structure

Definitions: Marks and channels

- marks
 - geometric primitives
- channels
 - control appearance of marks

Encoding visually with marks and channels

- analyze idiom structure
- as combination of marks and channels

1: vertical position, mark: line
2: vertical position, horizontal position, mark: point
3: vertical position, horizontal position, color hue, mark: point
4: vertical position, horizontal position, color hue, size (area), 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	
Color saturation	
Curvature	
Volume (3D size)	

Channels: Rankings

- Magnitude Channels: Ordered Attributes
- Identity Channels: Categorical Attributes

Effectiveness: Best to Least

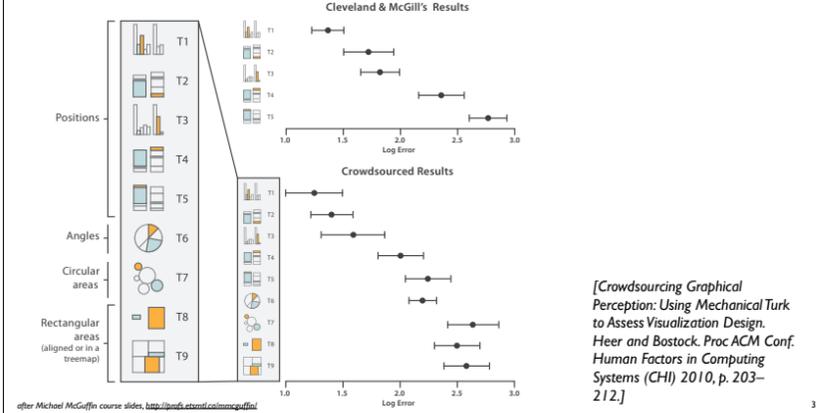
- effectiveness principle
 - encode most important attributes with highest ranked channels
- expressiveness principle
 - match channel and data characteristics

Accuracy: Fundamental Theory

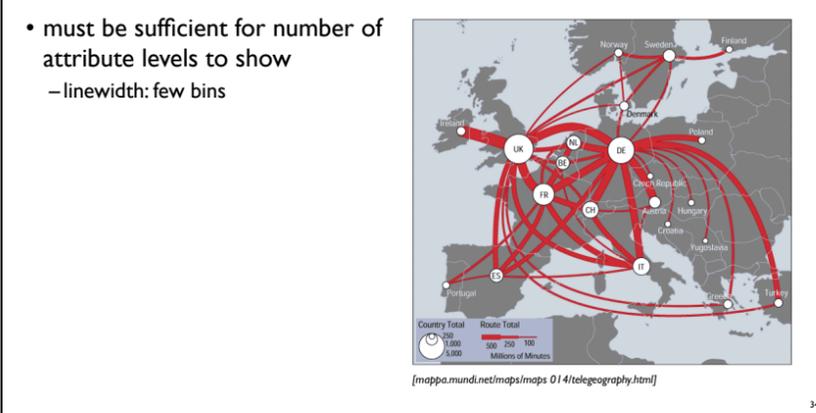
Steven's Psychophysical Power Law: $S = I^N$

Curves shown: Electric Shock (3.5), Saturation (1.7), Length (1), Area (0.7), Depth (0.67), Brightness (0.5)

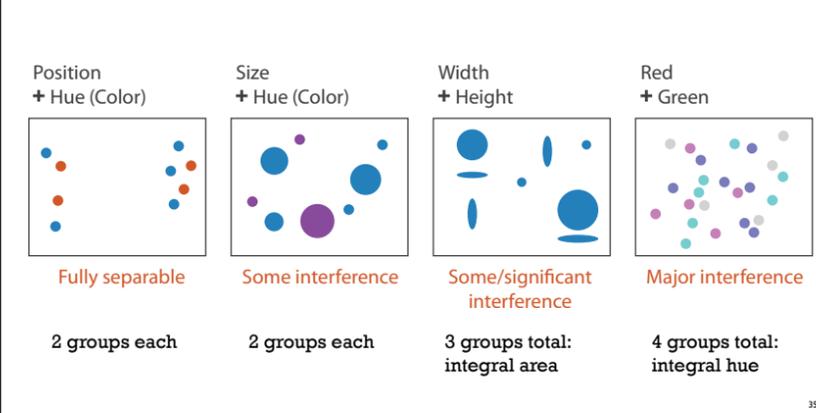
Accuracy: Vis experiments



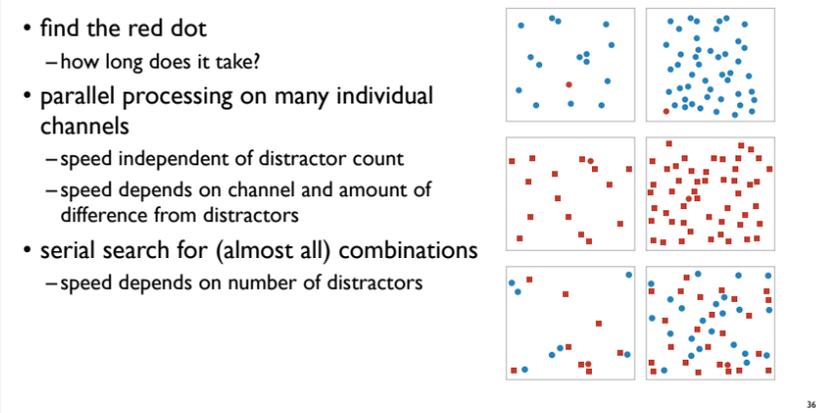
Discriminability: How many usable steps?



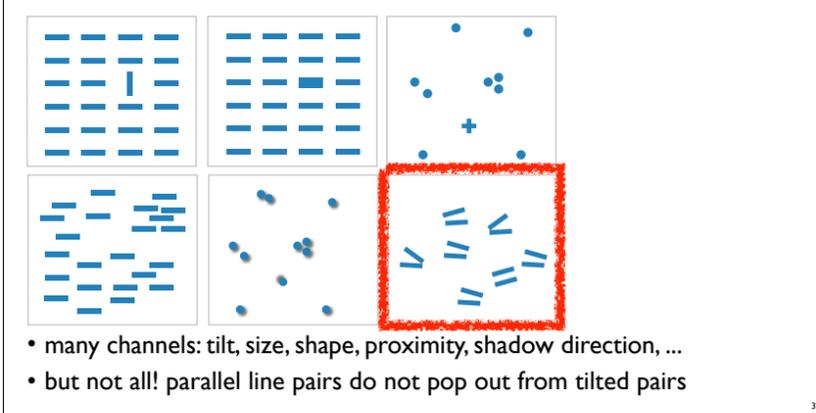
Separability vs. Integrality



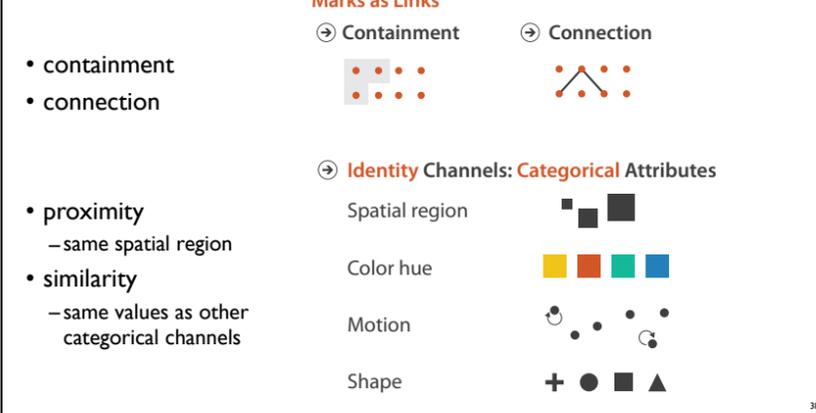
Popout



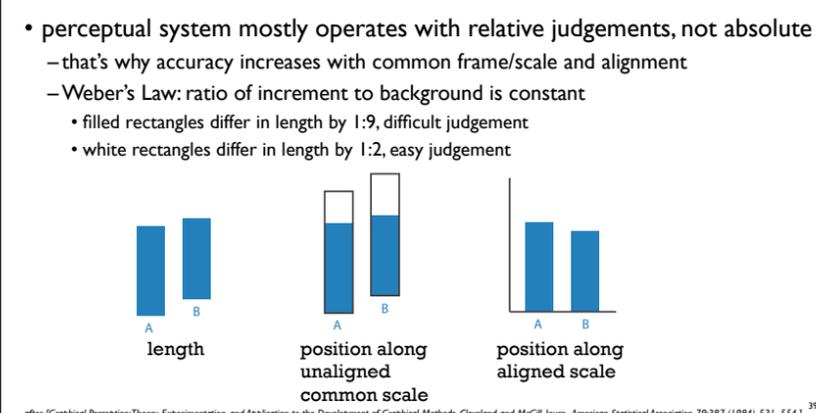
Popout



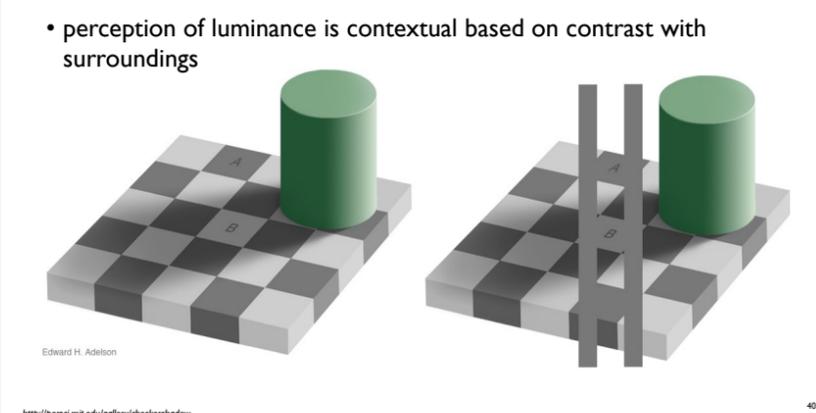
Grouping



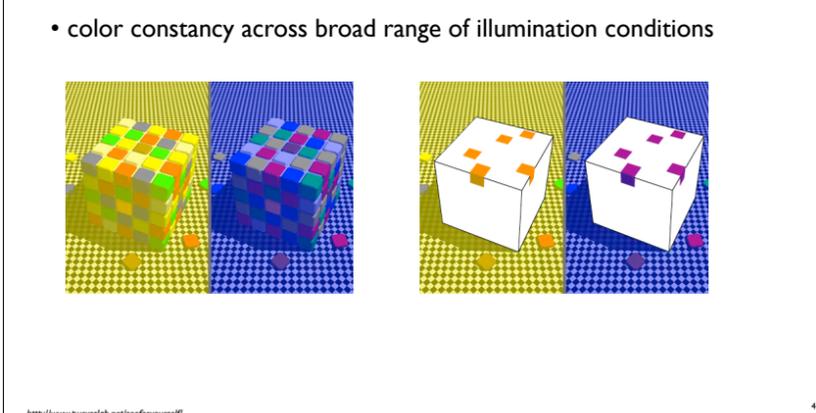
Relative vs. absolute judgements



Relative luminance judgements



Relative color judgements



- Visualization Analysis and Design. Tamara Munzner. CRC Press, 2014.
 - Chap 1, What's Vis, and Why Do It?
 - Chap 2, What: Data Abstraction
 - Chap 3, Why: Task Abstraction
 - Chap 4, Analysis: Four Levels for Validation
 - Chap 5, Marks and Channels
- Crowdsourcing Graphical Perception: Using Mechanical Turk to Assess Visualization Design. Jeffrey Heer and Michael Bostock. Proc. CHI 2010
- Perception in Vision web page with demos, Christopher Healey.
- Visual Thinking for Design. Colin Ware. Morgan Kaufmann, 2008.