Visualization Analysis & Design BMVC Tidbits

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UBC Biomedical Visualization and Communication Oct 2025, virtual

Visualization Analysis & Design Tamara Munzner





Defining visualization (vis)

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

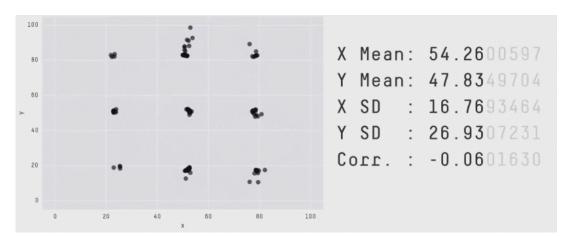
Why represent data visually?

Computer-based visualization systems provide visual representations of dataset 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 6 8 10 12 14 16 18 10 12 14 16 18 10 x variance 7.5 y mean 10 -3.75 y variance 0.816 x/y correlation 10 12 14 16 18 4 6 8 10 12 14 16 18 X3 X4

Datasaurus Dozen

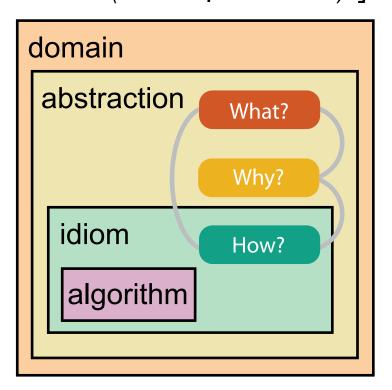


Same Stats, Different Graphs: Generating
Datasets with Varied Appearance and
Identical Statistics through Simulated
Annealing. CHI 2017. Matejka & Fitzmaurice

- domain situation
 - who are the target users?

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 - who are the target users?
- abstraction
 - translate from specifics of domain to vocabulary of vis
 - what is shown? data abstraction
 - why is the user looking at it? task abstraction

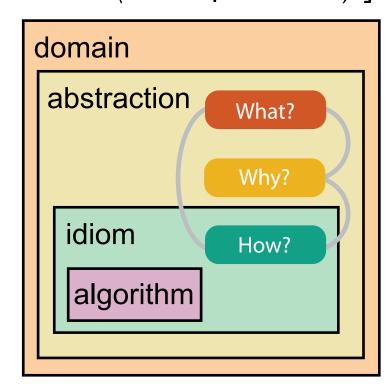
[A Nested Model of Visualization Design and Validation. Munzner. IEEETVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]



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

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

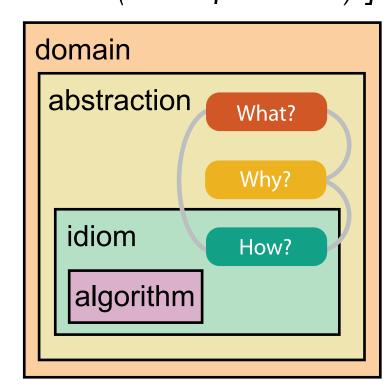
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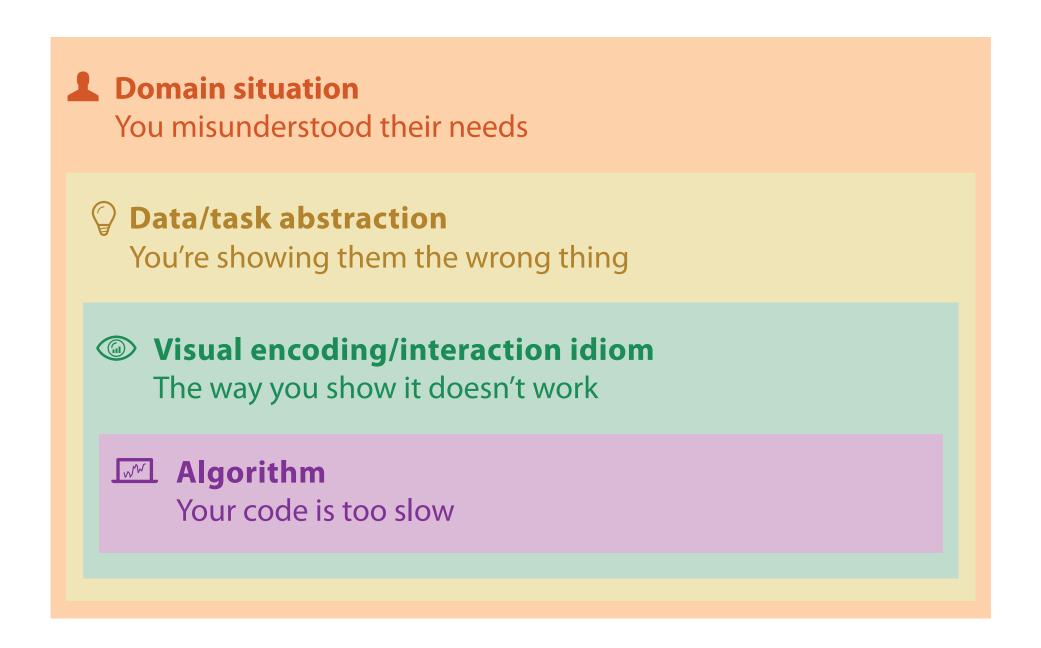
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- algorithm
 - efficient computation

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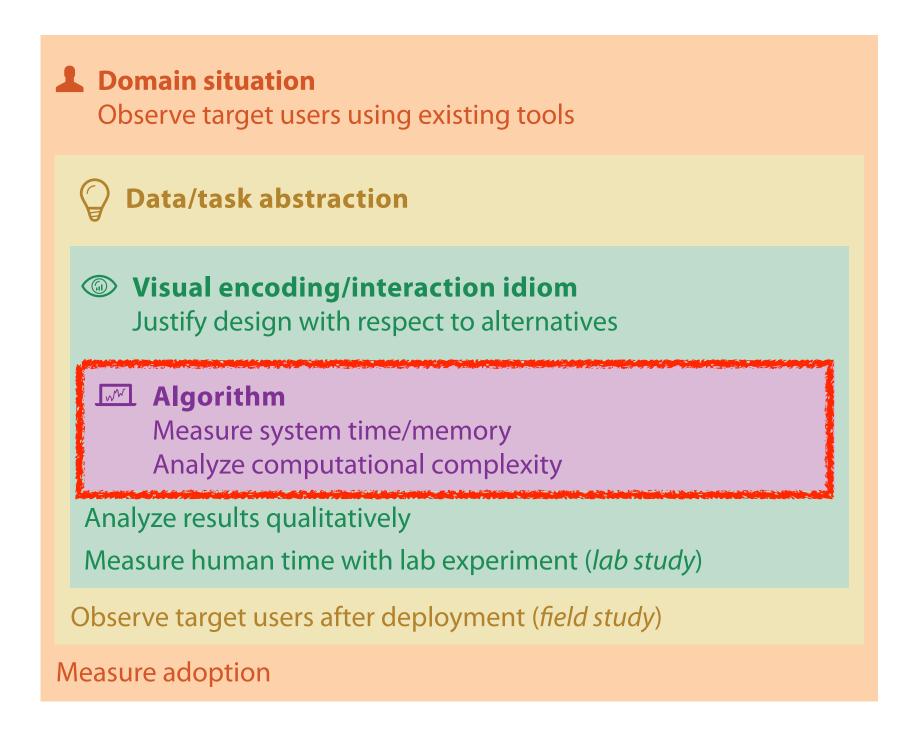


[A Multi-Level Typology of Abstract Visualization Tasks Brehmer and Munzner. IEEETVCG 19(12):2376-2385, 2013 (Proc. InfoVis 2013).]

different ways to get it wrong at each level



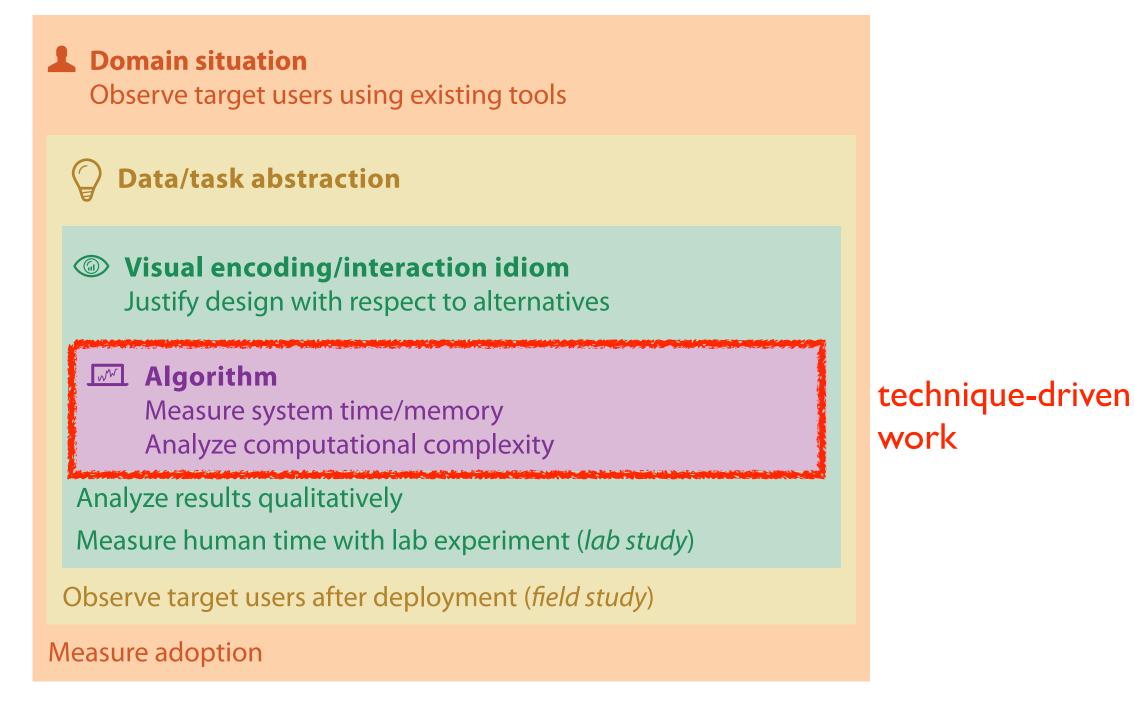
solution: use methods from different fields at each level



computer

science

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).]

design

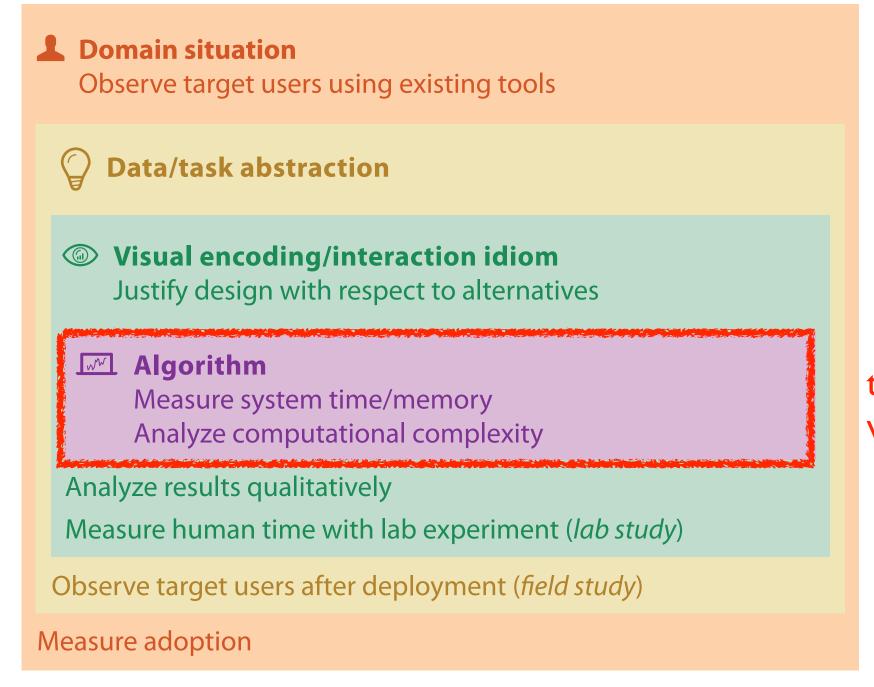
computer

science

cognitive

psychology

solution: use methods from different fields at each level



technique-driven work

solution: use methods from different fields at each level

Domain situation anthropology/ Observe target users using existing tools ethnography Data/task abstraction Wisual encoding/interaction idiom design Justify design with respect to alternatives **Algorithm** computer Measure system time/memory science Analyze computational complexity cognitive Analyze results qualitatively psychology Measure human time with lab experiment (*lab study*) Observe target users after deployment (*field study*) anthropology/ ethnography Measure adoption

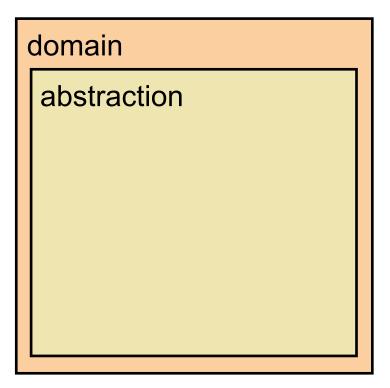
problem-driven work (design study)

technique-driven work

Abstractions

Abstraction: Data & task

- map what and why into generalized terms
 - identify tasks that users wish to perform, or already do
 - -find data types that will support those tasks
 - possibly transform /derive if need be



Example: Find good movies

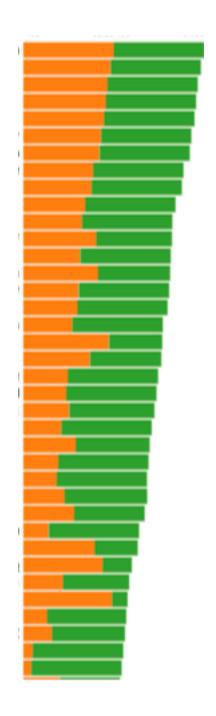
- identify good movies in genres I like
- domain:
 - -general population, movie enthusiasts
- task: what is a good movie for me?
 - highly rated by critics?
 - -highly rated by audiences?
 - -successful at the box office?
 - similar to movies I liked?
 - -matches specific genres?
- data: (is it available?)
 - -yes! data sources IMDB, Rotten Tomatoes...

Example: Find good movies

- one possible choice for data and tasks, in domain language
 - data: combine audience ratings and critic ratings
 - task: find high-scoring movies for specific genre
- abstractions?
 - -attribute: audience & critic ratings
 - ordinal
 - -levels: 3 or 5 or 10...
 - -attribute: genre
 - categorical
 - levels: < 20
 - items: movies
 - items: millions
 - -task: find extreme (high) values

one possible idiom

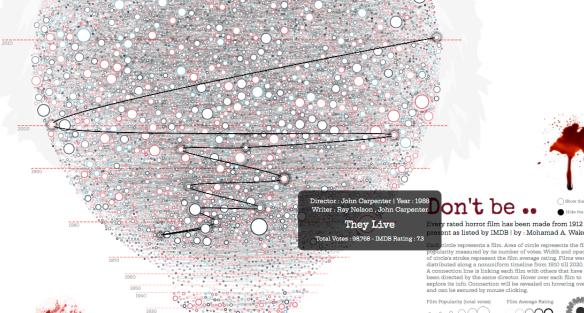
-stacked bar chart for ratings



Example: Horrified

- same task: high-score movies
- slightly different data
 - I4K rated horror movies from IMDB

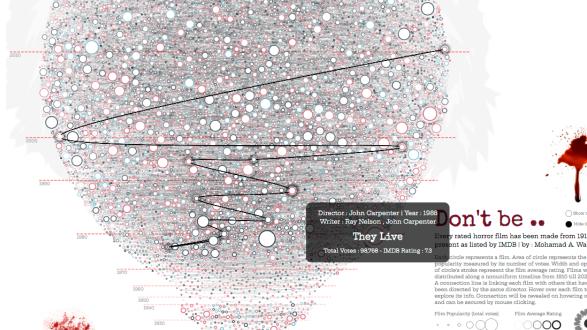




Example: Horrified

- same task: high-score movies
- slightly different data
 - 14K rated horror movies from IMDB
- very different visual encoding idiom
 - circle per item (movie)
 - circle area = popularity
 - -stroke width/opacity = avg rating
 - -year made = vertical position
- interaction idiom
 - lines connect movies w/ same director,on mouseover

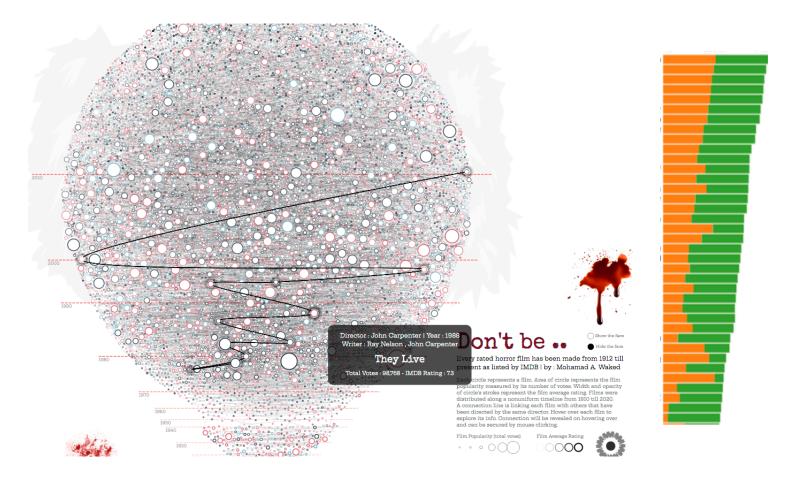




Example: Horrified vs stacked bars

- horrified: browse/explore
- stacked bars: locate/lookup

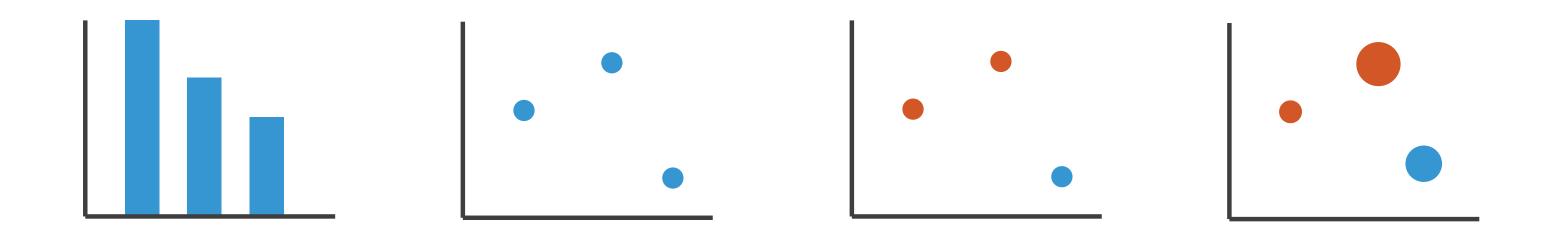
- which is better?
 - depends on goals / task
 - enjoy, social context, lots of time
 - find 2nd-best rated movie of all time
 - Jeopardy call, < 10 seconds to respond!</p>



http://alhadaqa.com/2019/10/horrified/

Visual Encoding Idioms

• analyze idiom structure



- marks
 - -geometric primitives



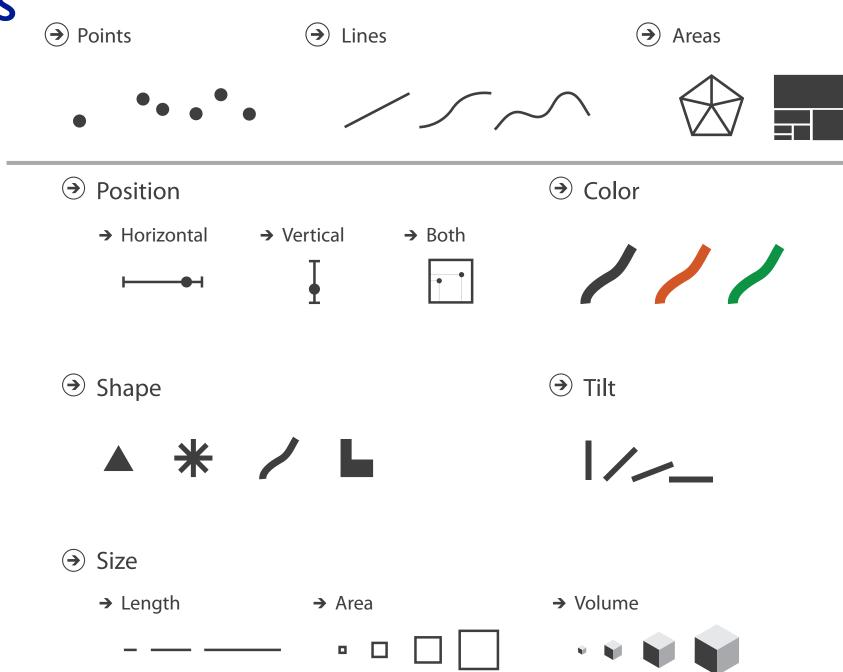




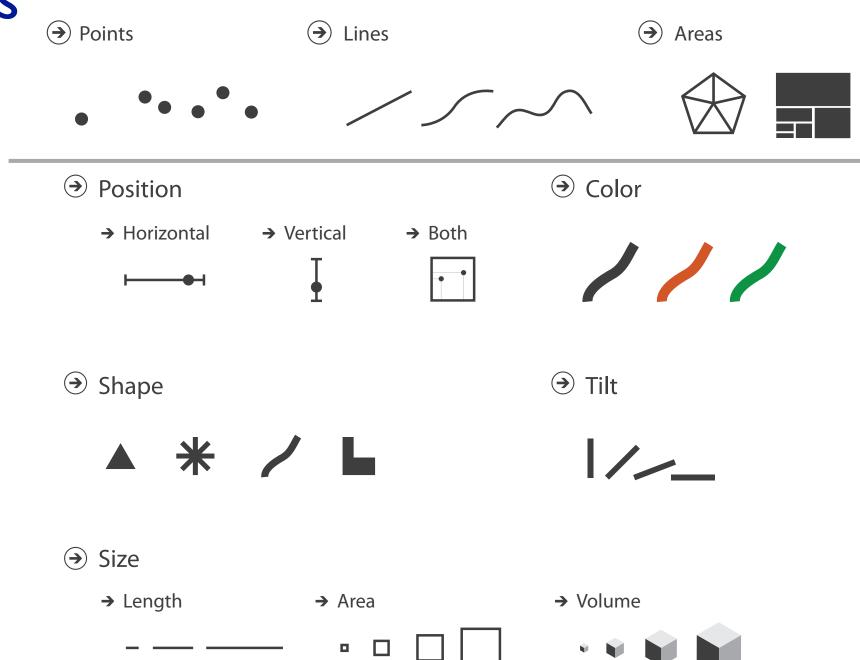




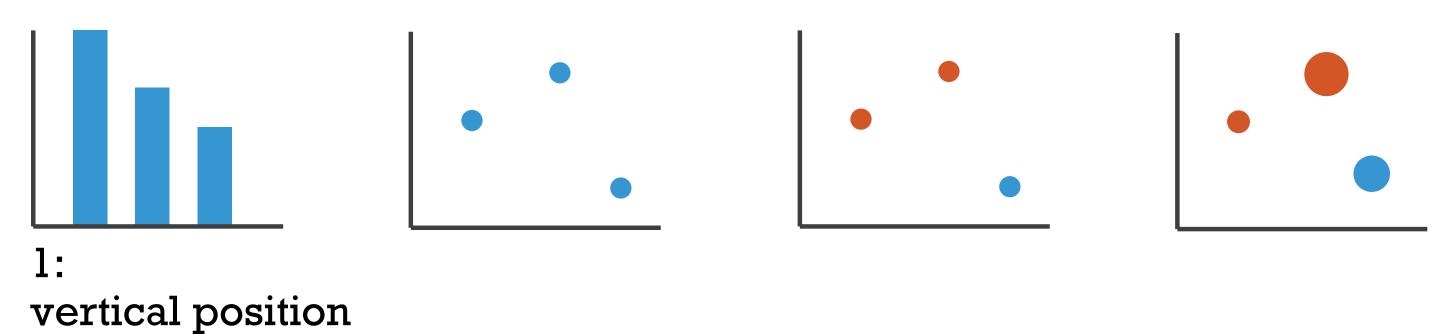
- marks
 - -geometric primitives
- channels
 - control appearance of marks



- marks
 - -geometric primitives
- channels
 - control appearance of marks
- channel properties differ
 - type & amount of information that can be conveyed to human perceptual system

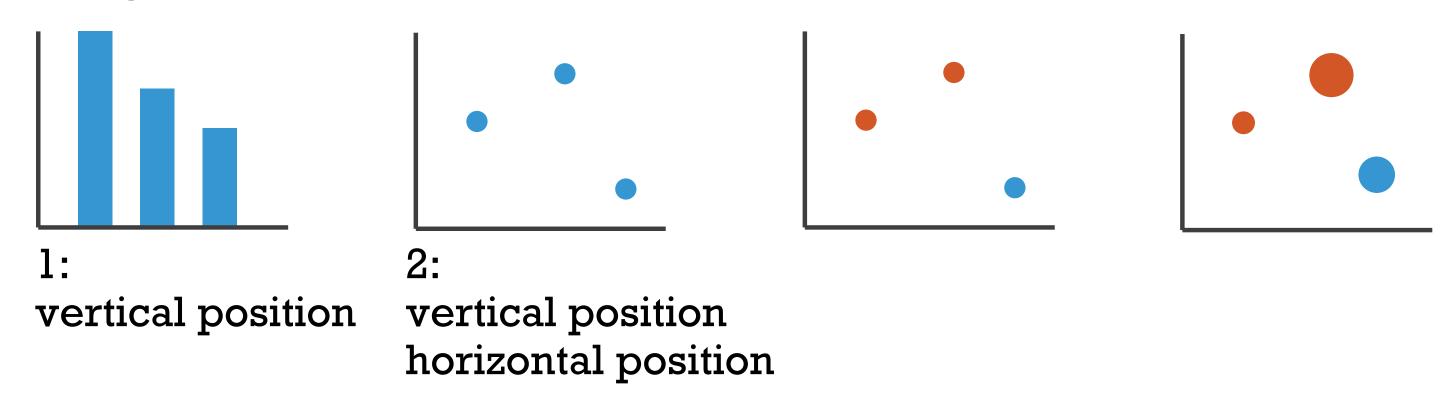


• analyze idiom structure as combination of marks and channels



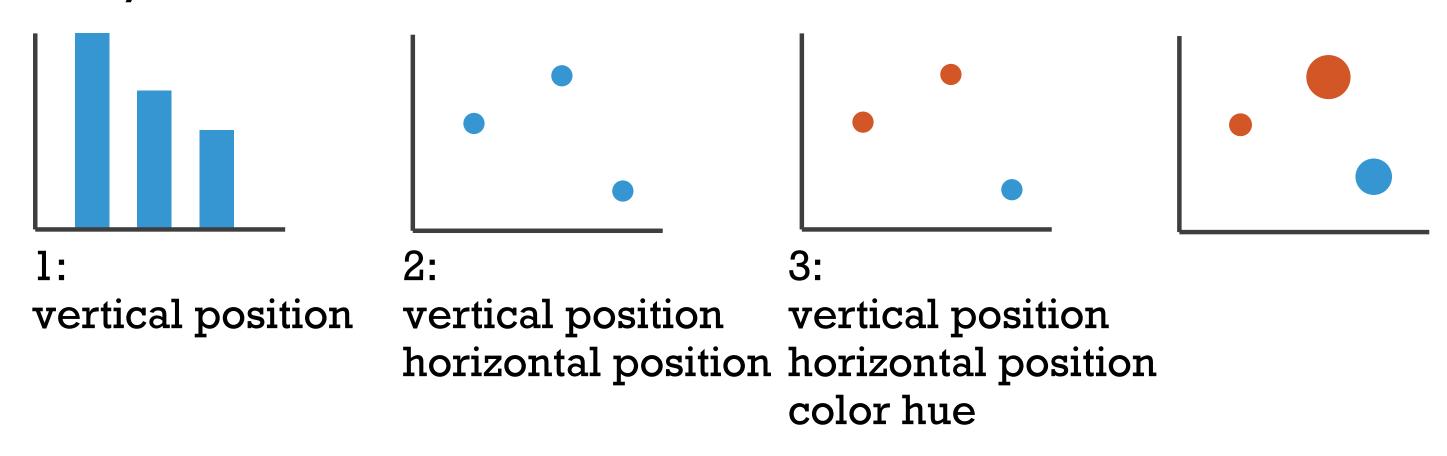
mark: line

analyze idiom structure as combination of marks and channels



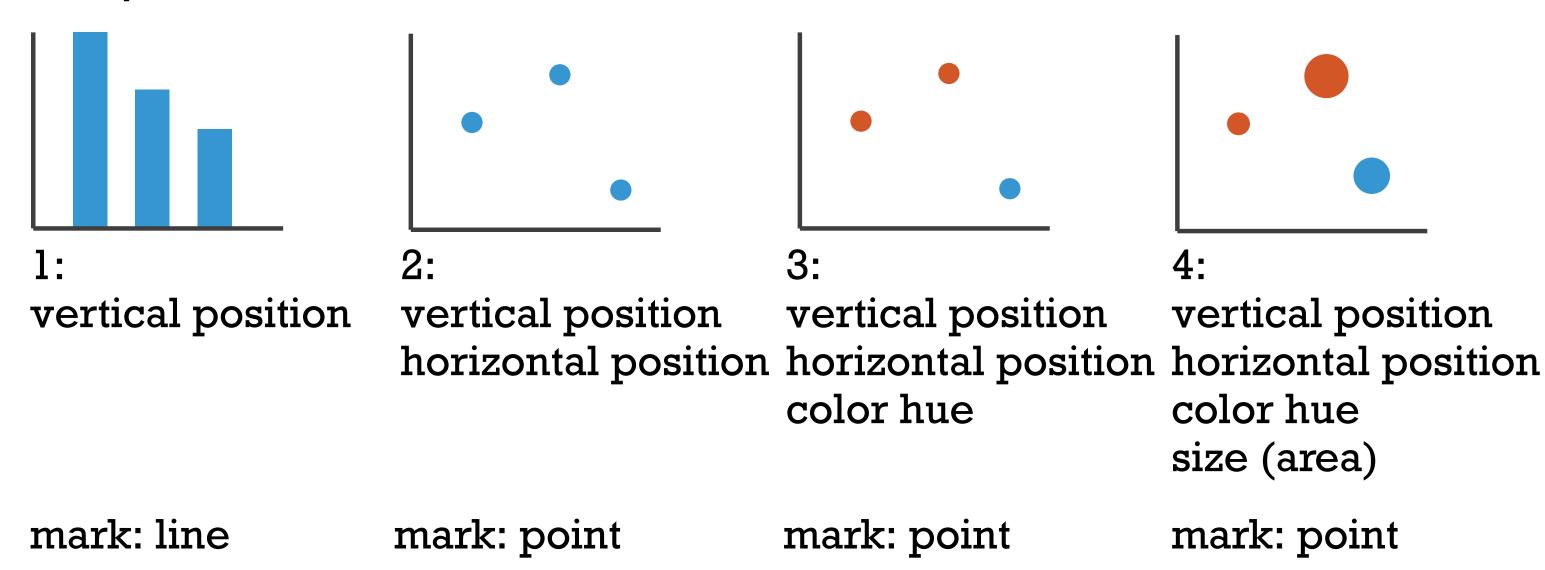
mark: line mark: point

analyze idiom structure as combination of marks and channels



mark: line mark: point mark: point

analyze idiom structure as combination of marks and channels

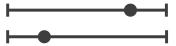


Position on common scale Position on unaligned scale Length (1D size) Tilt/angle Area (2D size) Depth (3D position) Color luminance Color saturation Curvature Volume (3D size)

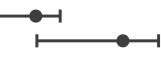


→ Magnitude Channels: Ordered Attributes

Position on common scale



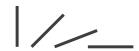
Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



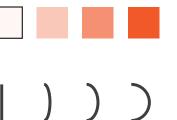
Depth (3D position)



Color luminance



Color saturation



Curvature



Volume (3D size)

→ Identity Channels: Categorical Attributes

Spatial region

Color hue

Motion

Shape

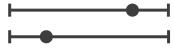
The state of th

expressiveness

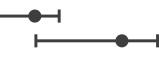
- match channel and data characteristics

→ Magnitude Channels: Ordered Attributes

Position on common scale



Position on unaligned scale



Length (1D size)



Tilt/angle



Area (2D size)



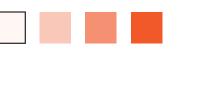
Depth (3D position)



Color luminance



Color saturation



Curvature



Volume (3D size)

Description Identity Channels: Categorical Attributes

Spatial region

Color hue

Motion

→ Categorical

→ Ordered

Ordered

→ Ordinal

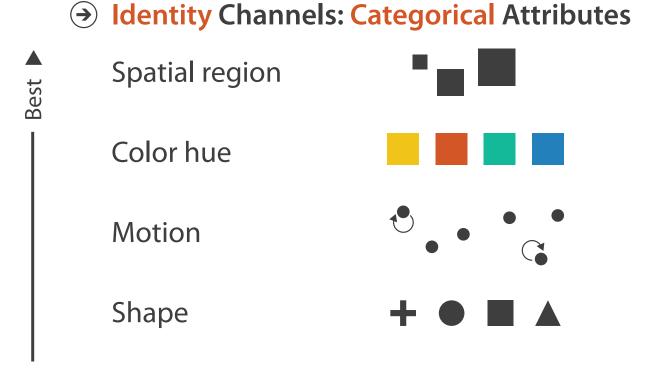
expressiveness

- match channel and data characteristics
- magnitude for ordered
 - how much? which rank?
- identity for categorical
 - what?

→ Ouantitative

→ 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)



expressiveness

Effectiveness

- match channel and data characteristics
- effectiveness
 - channels differ in accuracy of perception

→ 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)

- Jedentity Channels: Categorical Attributes
 Spatial region
 - Color hue
 - Motion
 - Shape + •

expressiveness

Best **▶**

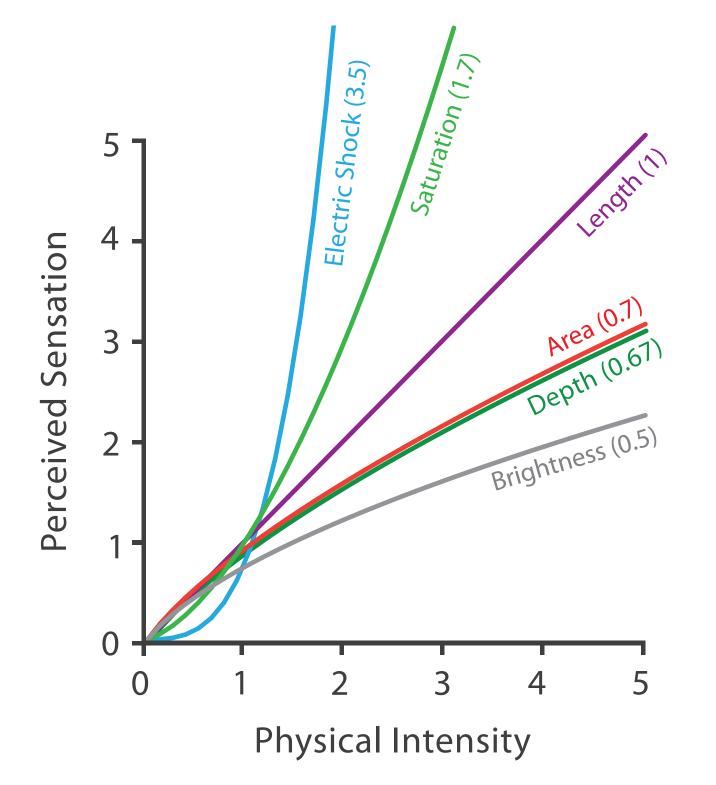
Effectiveness

- match channel and data characteristics
- effectiveness
 - channels differ in accuracy of perception
 - spatial position ranks high for both

Accuracy: Fundamental theory

- length is accurate: linear
- others magnified or compressed
 - –exponent characterizes

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

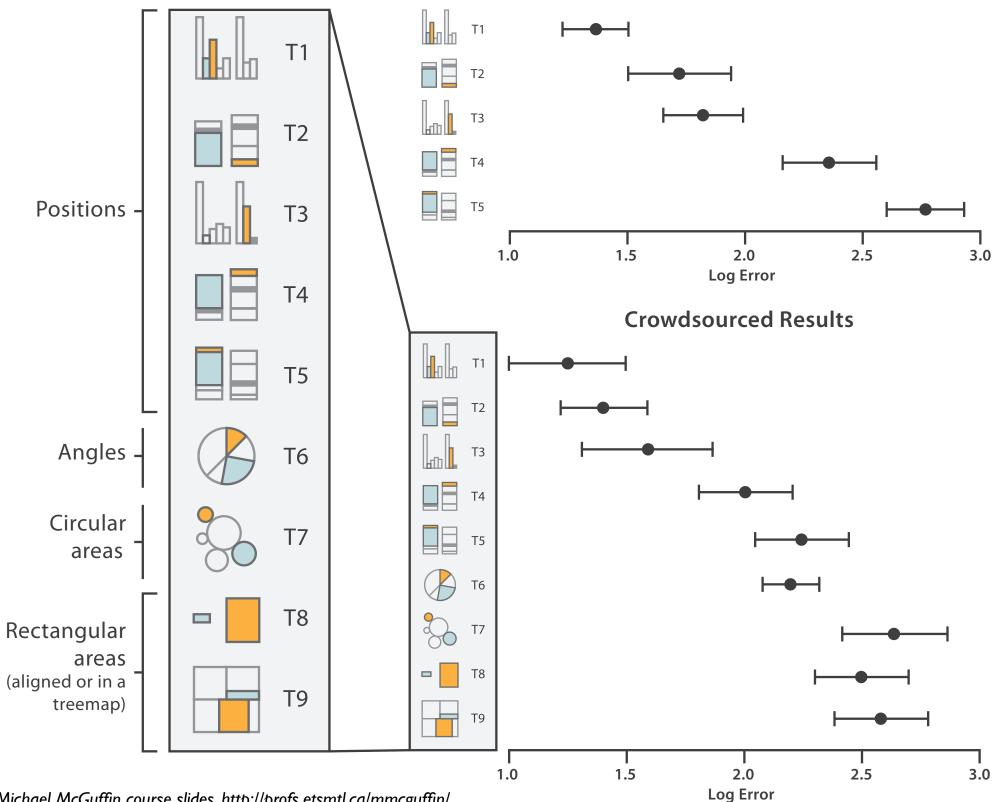


S = sensation

I = intensity

Accuracy: Vis experiments

Cleveland & McGill's Results

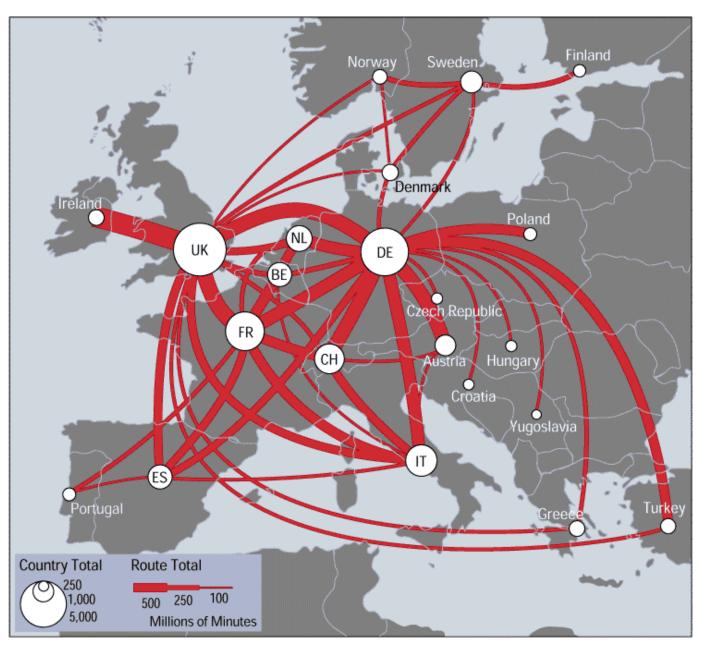


[Crowdsourcing Graphical Perception: Using Mechanical Turk to Assess Visualization Design. Heer and Bostock. Proc ACM Conf. **Human Factors in Computing** Systems (CHI) 2010, p. 203-212.]

Discriminability: How many usable steps?

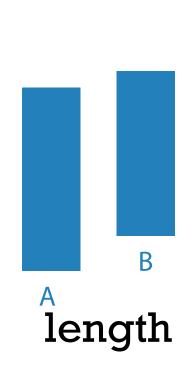
 must be sufficient for number of attribute levels to show

-linewidth: few bins

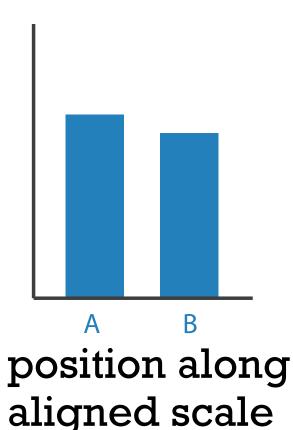


• perceptual system mostly operates with relative judgements, not absolute

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 - -that's why accuracy increases with common frame/scale and alignment

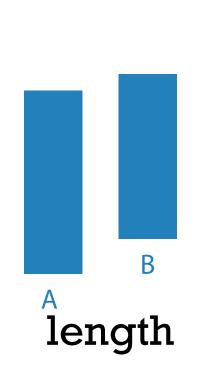


A position along unaligned common scale

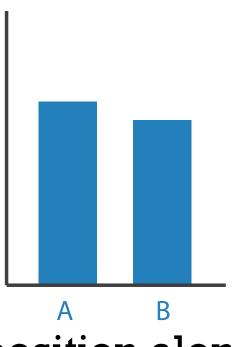


after [Graphical Perception:Theory, Experimentation, and Application to the Development of Graphical Methods. Cleveland and McGill. Journ. American Statistical Association 79:387 (1984), 531–554.]

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

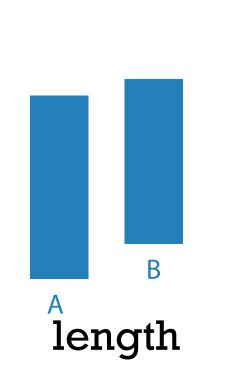


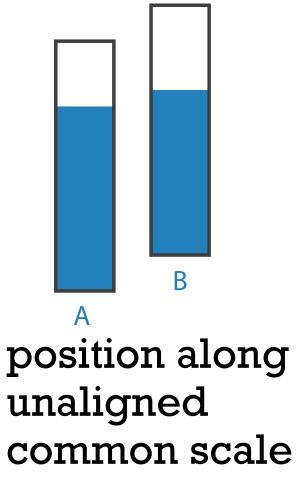
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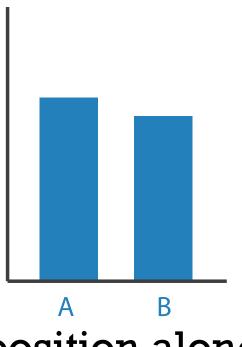


position along aligned scale

- 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



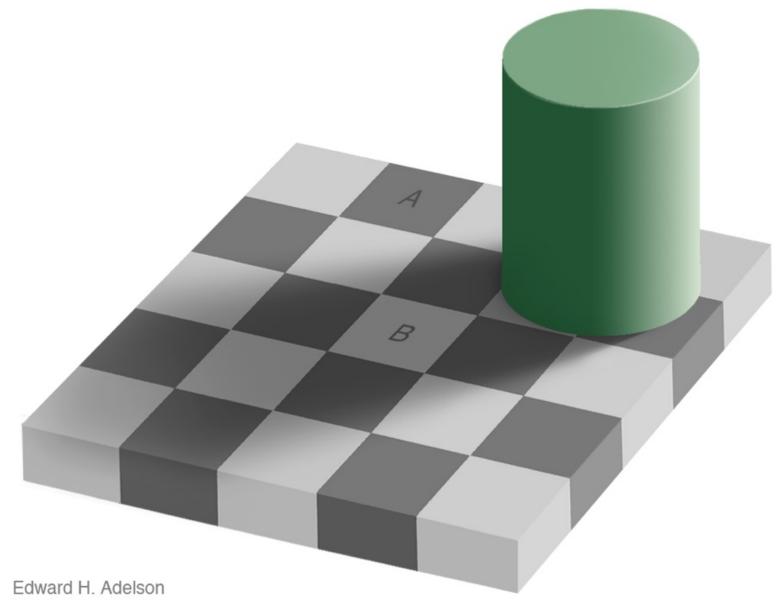




position along aligned scale

Relative luminance judgements

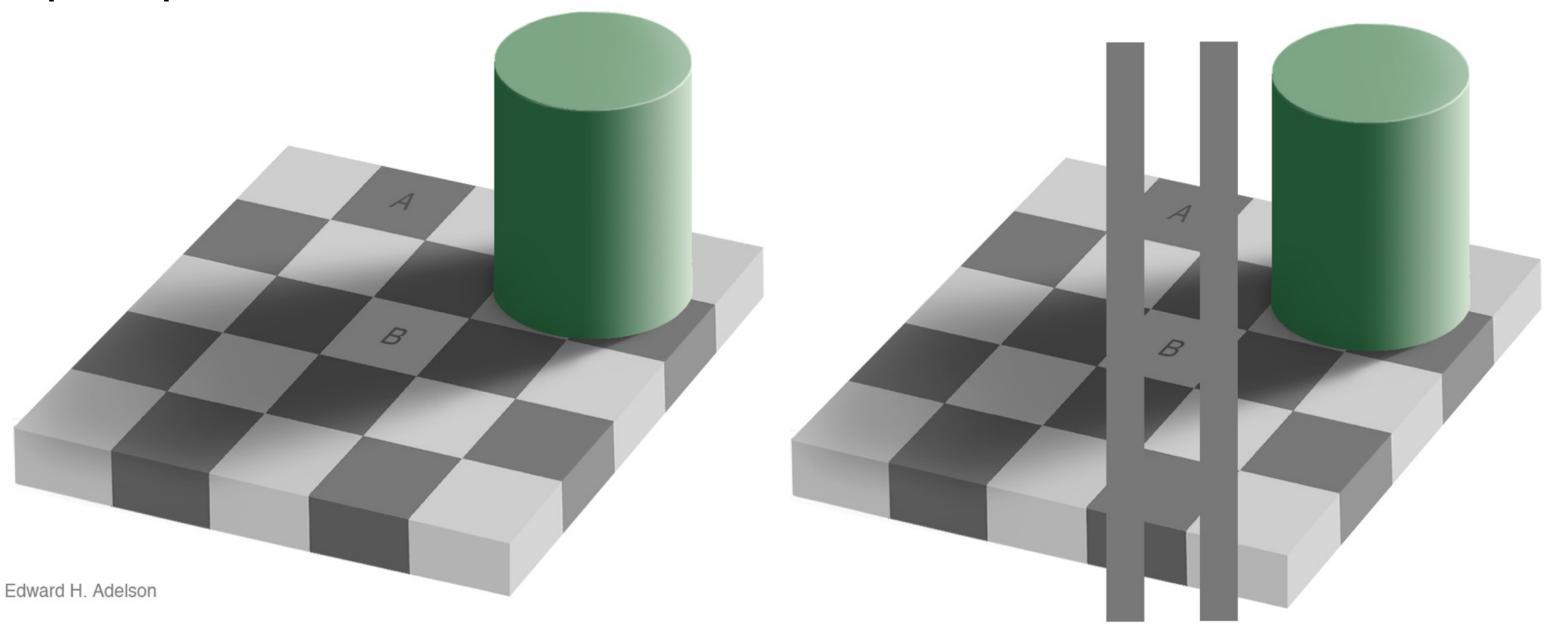
• perception of luminance is contextual based on contrast with surroundings



42

Relative luminance judgements

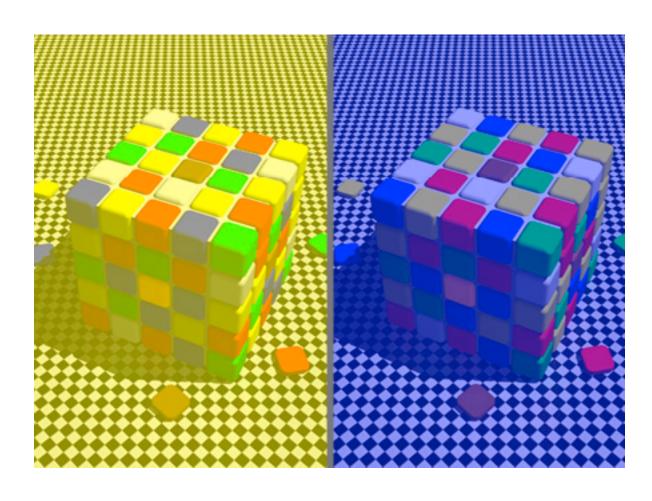
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43

Relative color judgements

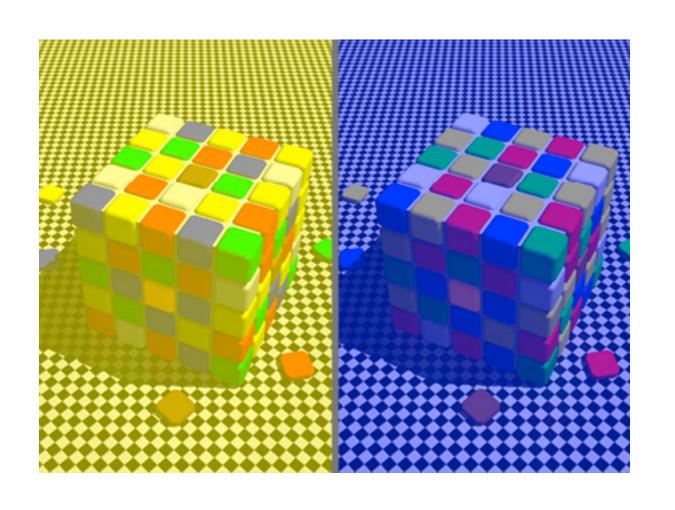
• color constancy across broad range of illumination conditions

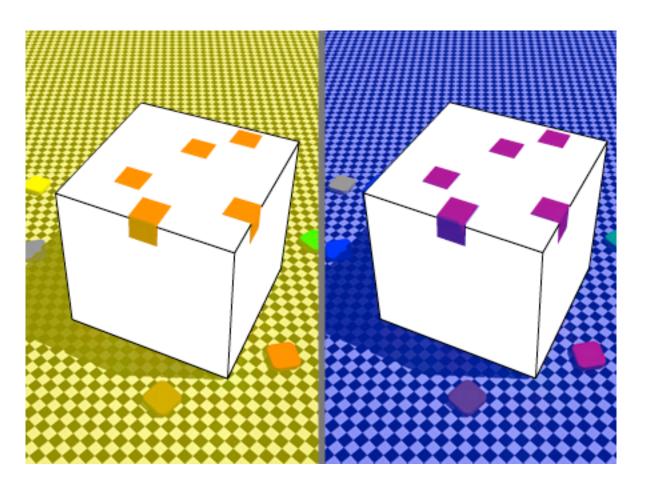


44

Relative color judgements

• color constancy across broad range of illumination conditions





Grouping

- containment
- connection

- proximity
 - -same spatial region
- similarity
 - same values as other categorical channels

Marks as Links

→ Containment







Identity Channels: Categorical Attributes

Spatial region



Color hue



Motion



Shape



Graphic Design for Non-Designers

What Goes Around

Comes Around

Lessons from hitchhiking

across the country

Robin Williams

January 1, 2005

- proximity
 - do group related items together
 - avoid equal whitespace between unrelated

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 - do unify by pushing existing consistencies

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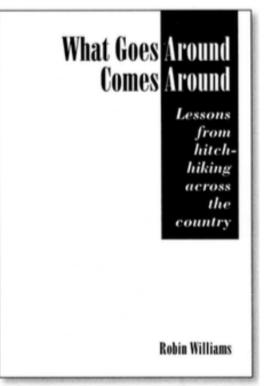
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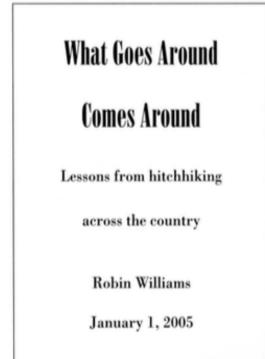
What Goes Around -Comes Around -

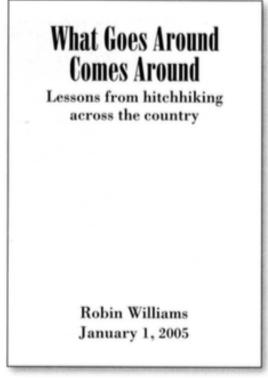
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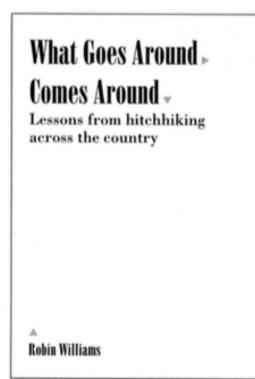


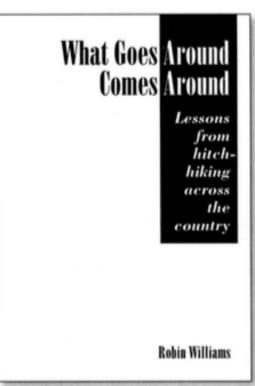
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- The Non-Designer's Design Book, 4th ed. Robin Williams, Peachpit Press, 2015.
 - fast read, very practical to work through whole thing

Function first, form next

- dangerous to start with aesthetics
 - usually impossible to add function retroactively
- start with focus on functionality
 - -possible to improve aesthetics later on, as refinement
 - if no expertise in-house, find good graphic designer to work with
 - -aesthetics do matter! another level of function
 - visual hierarchy, alignment, flow
 - Gestalt principles in action

Usability Testing

Guerilla/Discount Usability

- usability testing as formative evaluation: how could it improve?
 - vs summative evaluation: did it work?
- grab a few people and watch them use your interface
 - even 3-5 gives substantial coverage of major usability problems
 - -agile/lean qualitative, vs formal quantitative user studies
 - goal is not statistical significance!
- think-aloud protocol
 - -contextual inquiry (conversations back and forth) vs fly on the wall (you're silent)

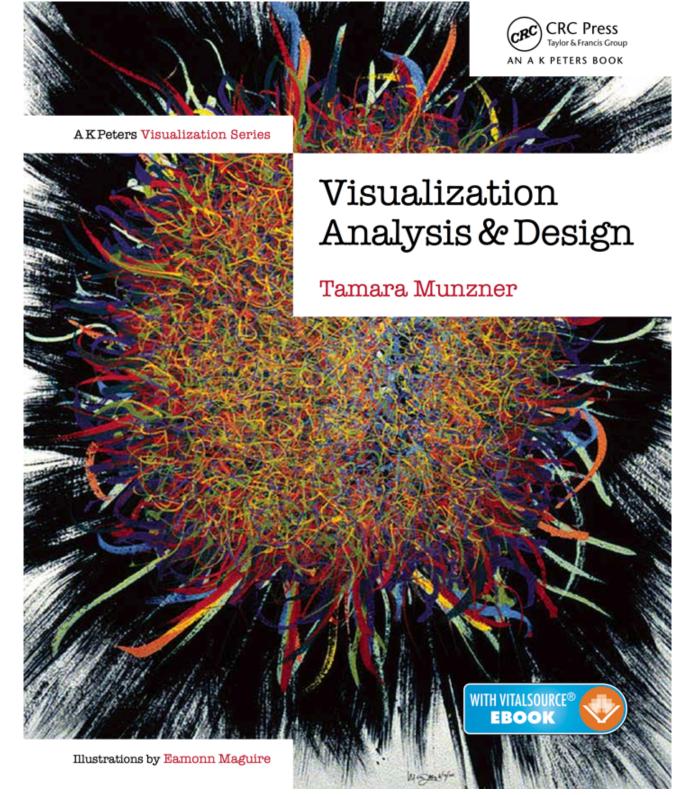
Further reading, usability

- 7 Step Guide to Guerrilla Usability Testing, Markus Piper
 - https://userbrain.net/blog/7-step-guide-guerrilla-usability-testing-diy-usability-testing-method
- Discount Usability: 20 Years, Jakob Nielsen
 - https://www.nngroup.com/articles/discount-usability-20-years/
- Interaction Design: Beyond Human-Computer Interaction
 - Preece, Sharp, Rogers. Wiley, 5th edition, 2019.
- About Face: The Essentials of Interaction Design
 - Cooper, Reimann, Cronin, Noessel. Wiley, 4th edition, 2014.
- Task-Centered User Interface Design. Lewis & Rieman, 1994
 - http://hcibib.org/tcuid/
- Designing with the Mind in Mind. Jeff Johnson. Morgan Kaufmann, 2nd, 2014.

More Resources: Books

Visualization Analysis and Design

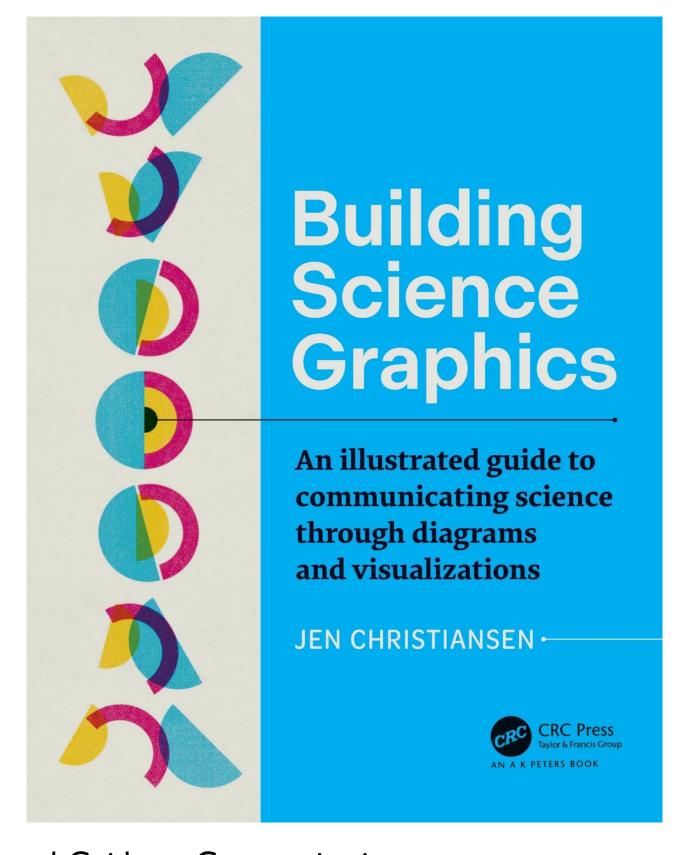
- book from which these tidbits were drawn
 - illustration acknowledgement: Eamonn Maguire
- full courses, papers, videos, software, talks http://www.cs.ubc.ca/group/infovis
 http://www.cs.ubc.ca/~tmm



https://www.cs.ubc.ca/~tmm/vadbook

Building Science Graphics

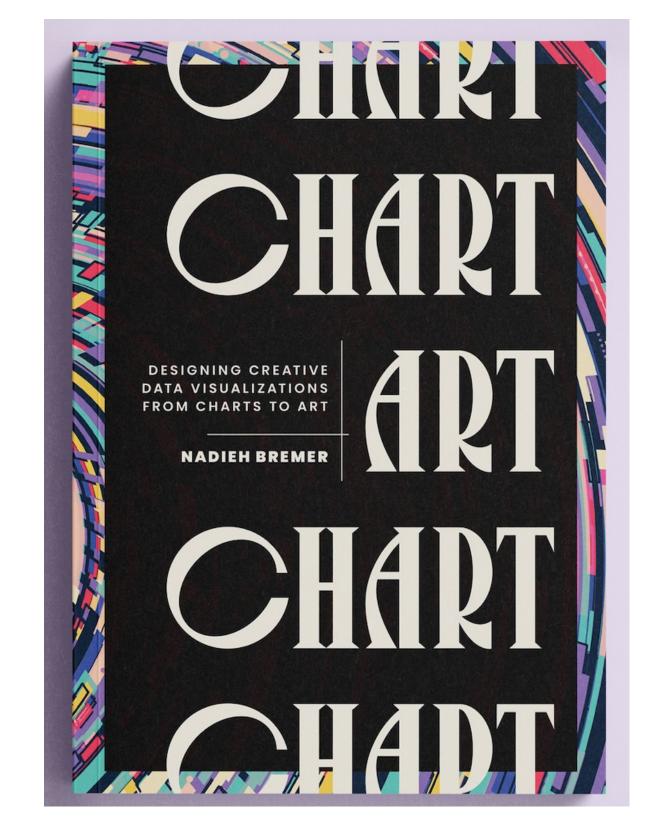
- Building Science Graphics:
 An illustrated guide to
 communicating science through diagrams and visualizations
 - -Jen Christiansen



https://www.buildingsciencegraphics.com/

CHART

- CHART: Designing Creative Data Visualizations from Charts to Art
 - -Nadieh Bremer



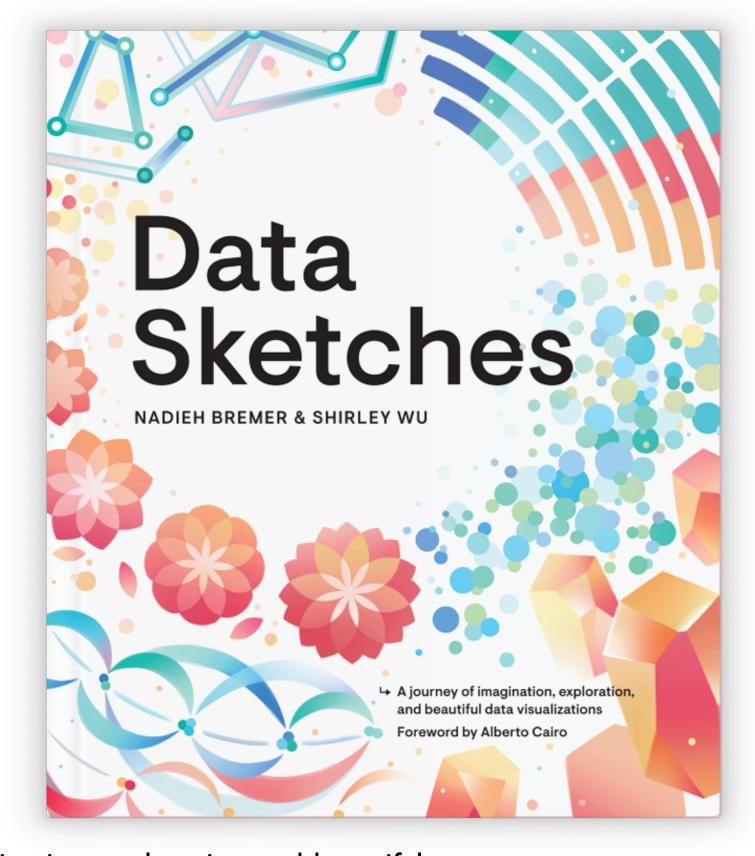
https://www.visualcinnamon.com/chart/

https://www.routledge.com/CHART-Designing-Creative-Data-Visualizations-from-Charts-to-Art/Bremer/p/book/9781032797755

Data Sketches

- Data Sketches:

 A journey of imagination,
 exploration, and beautiful data
 visualizations
 - Nadieh Bremer & Shirley Wu

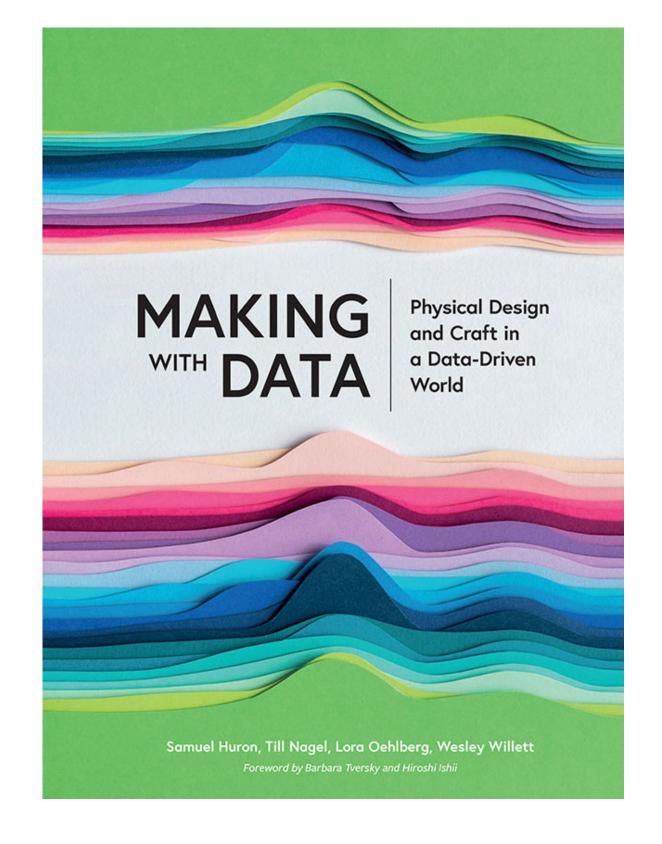


https://www.datasketch.es/book

https://www.routledge.com/Data-Sketches-A-journey-of-imagination-exploration-and-beautiful-data-visualizations/Bremer-Wu/p/book/9780367000080

Making with Data

- Making with Data: Physical Design and Craft in a Data-Driven World
 - editors:
 - Samuel Huron, Till Nagel, Lora Oehlberg, Wesley Willett



https://makingwithdata.org/

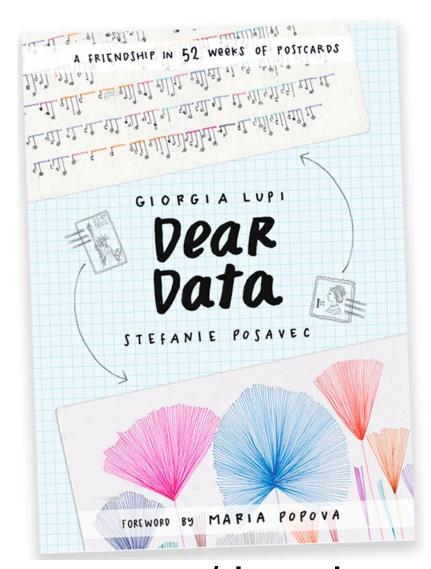
https://www.routledge.com/Making-with-Data-Physical-Design-and-Craft-in-a-Data-Driven-World/ Huron-Nagel-Oehlberg-Willett/p/book/9781032182223

Dear Data: A Friendship in 52 Weeks of Postcards

• inspiring celebration of data humanism



Giorgia Lupi and Stefanie Posavec



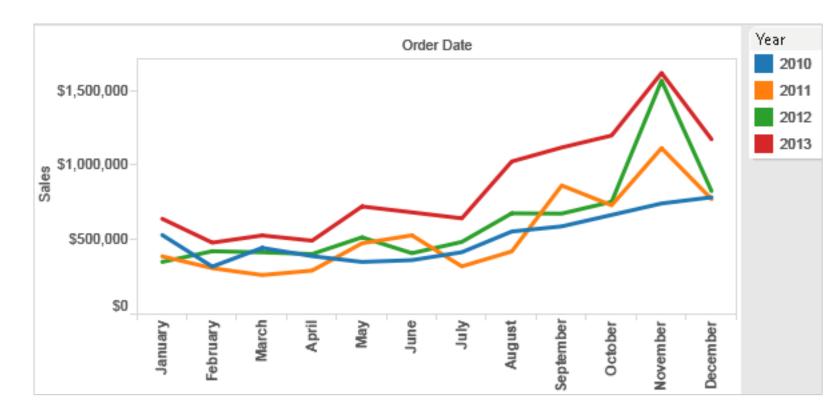
http://www.dear-data.com/by-week/

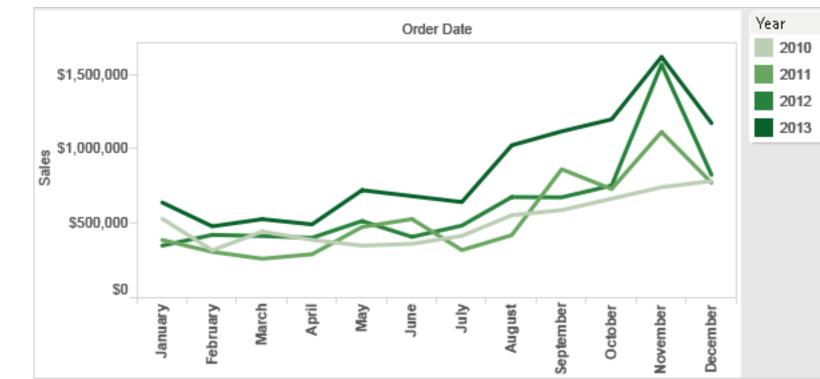
https://giorgialupi.com/dear-data

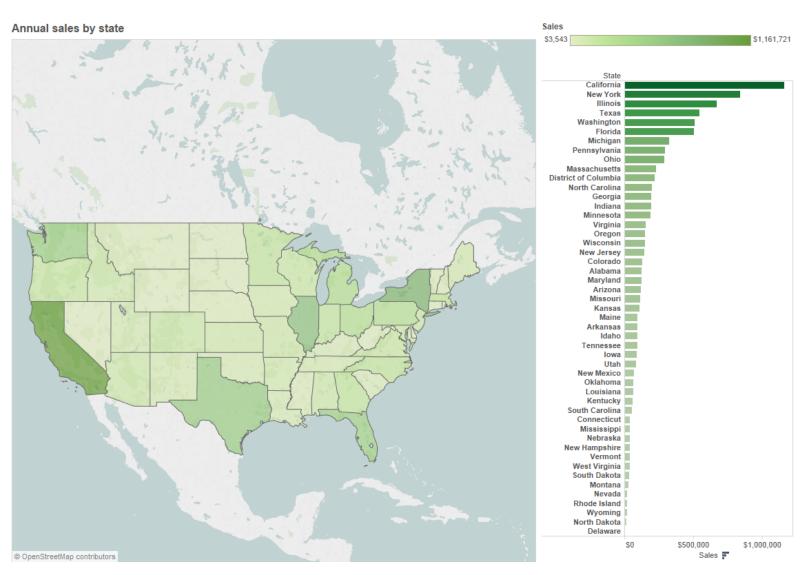
https://www.stefanieposavec.com/dear-data

Color

Categorical vs ordered color



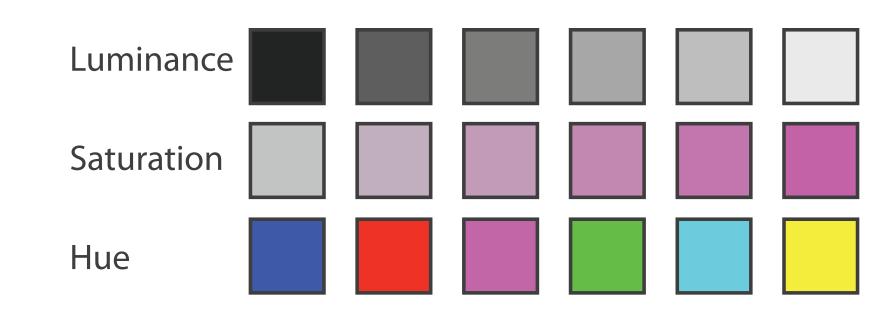




[Seriously Colorful: Advanced Color Principles & Practices. Stone.Tableau Customer Conference 2014.]

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



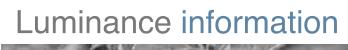
- channels have different properties
 - -what they convey directly to perceptual system
 - -how much they can convey: how many discriminable bins can we use?

Luminance

- need luminance for edge detection
 - fine-grained detail only visible through luminance contrast
 - -legible text requires luminance contrast!

• intrinsic perceptual ordering







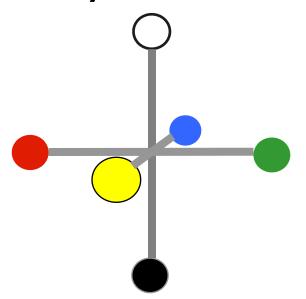
Chroma information



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

Opponent color and color deficiency

- perceptual processing before optic nerve
 - one achromatic luminance channel (L*)
 - -edge detection through luminance contrast
 - -2 chroma channels
 - -red-green (a*) & yellow-blue axis (b*)
- "color blind": one axis has degraded acuity
 - -8% of men are red/green color deficient
 - -blue/yellow is rare







Chroma information



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

Color spaces

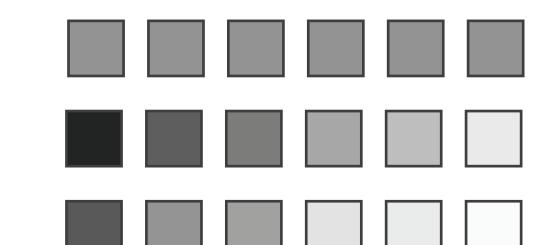
- CIE L*a*b*: good for computation
 - L* intuitive: perceptually linear luminance
 - a*b* axes: perceptually linear but nonintuitive
- RGB: good for display hardware
 - poor for encoding
- HSL/HSV: somewhat better for encoding
 - hue/saturation wheel intuitive
 - beware: only pseudo-perceptual!
 - lightness (L) or value (V) \neq luminance or L*
- Luminance, hue, saturation
 - good for encoding
 - but not standard graphics/tools colorspace

Corners of the RGB color cube

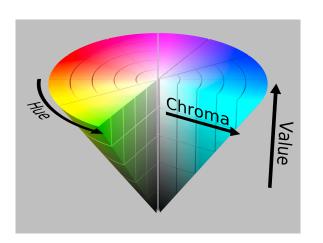
L from HLS All the same

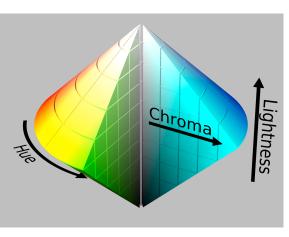
Luminance values

L* values









Designing for color deficiency: Check with simulator



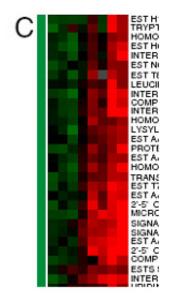
Normal vision

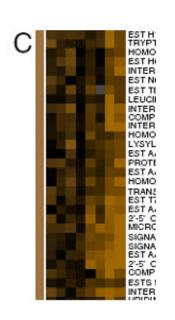


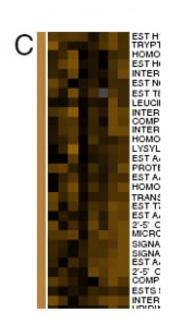
Deuteranope Protanope



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





\$120,000

\$160,000

COGS

Deuteranope simulation

\$240,000

\$280,000

\$200,000

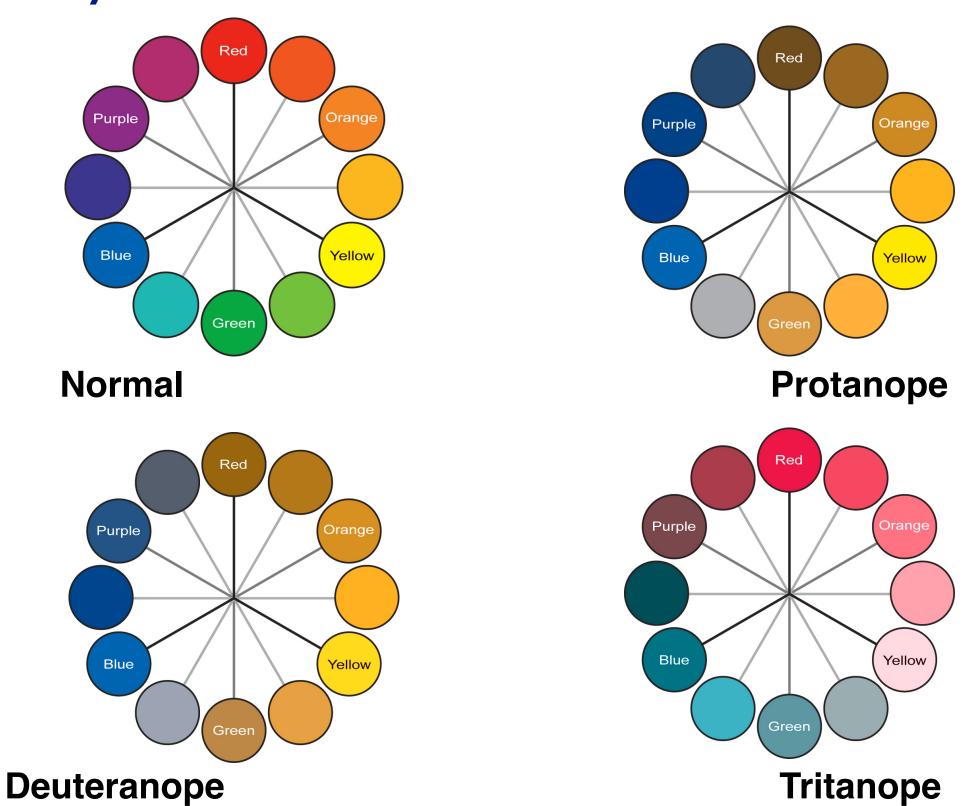
Change the shape

\$40,000

\$80,000

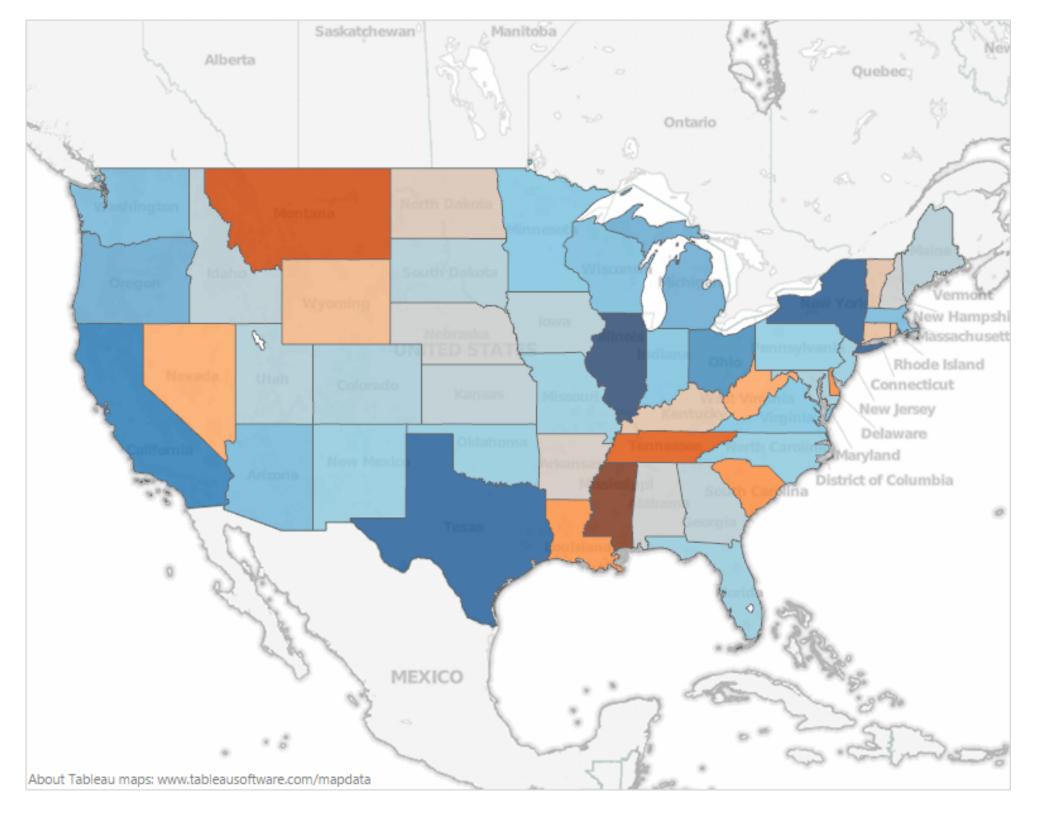
Vary luminance

Color deficiency: Reduces color to 2 dimensions



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

Designing for color deficiency: Blue-Orange is safe



Bezold Effect: Outlines matter

• color constancy: simultaneous contrast effect



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

Relative judgements: Color & illumination

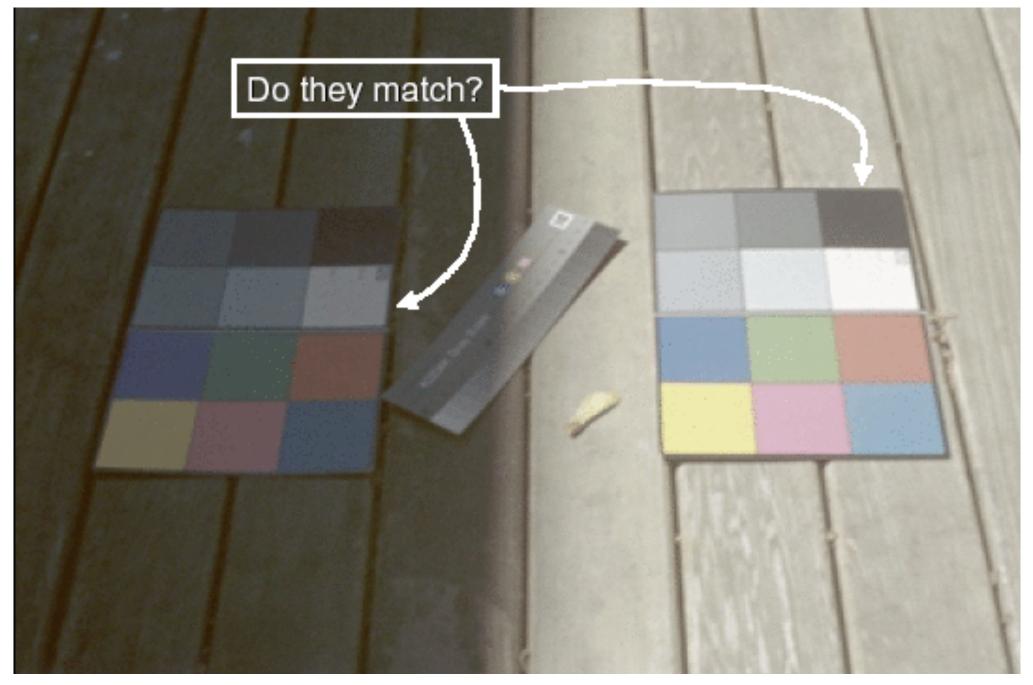


Image courtesy of John McCann

Relative judgements: Color & illumination

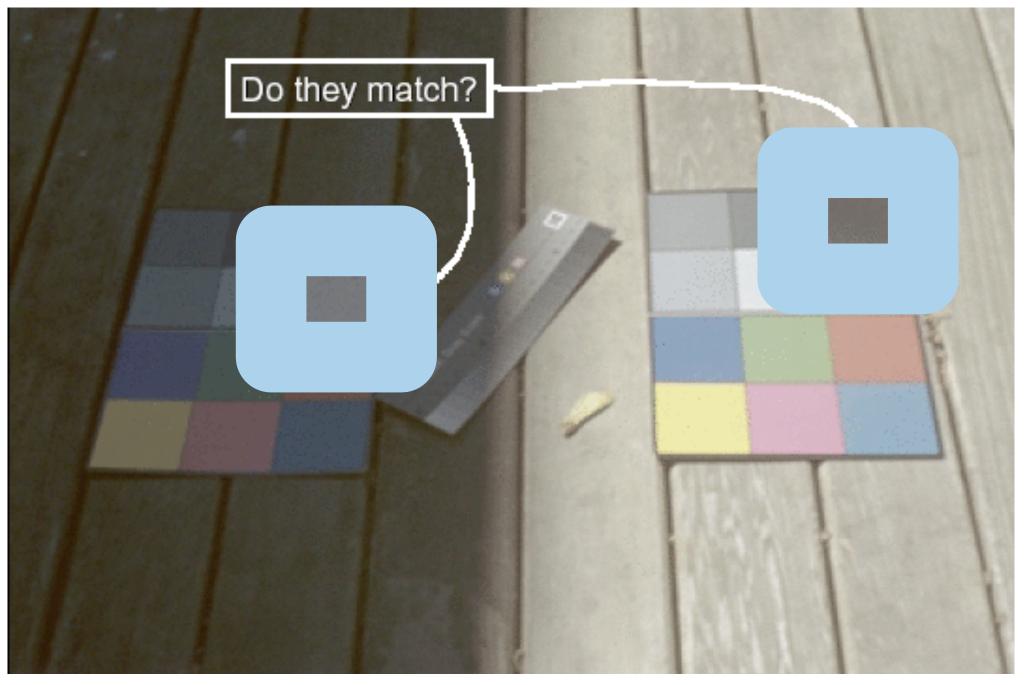
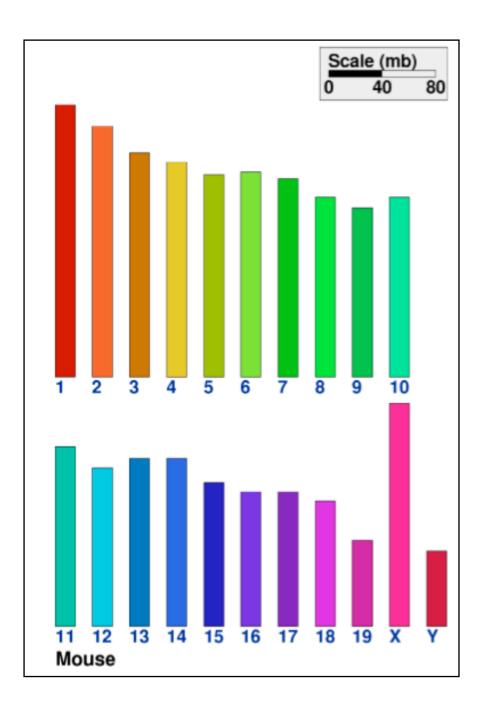
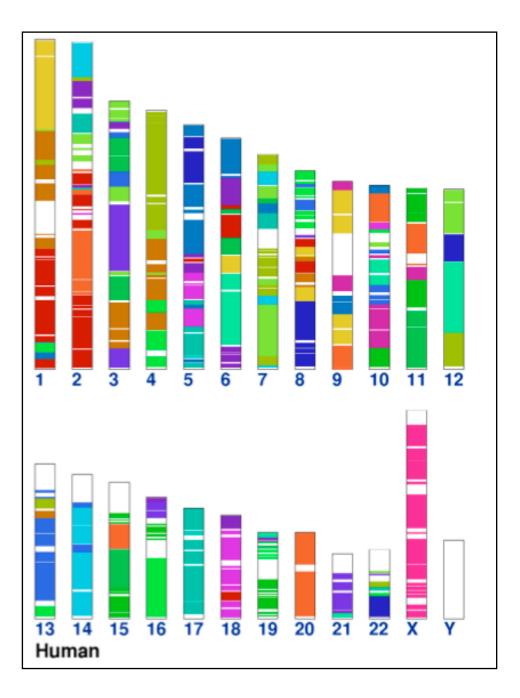


Image courtesy of John McCann via Maureen Stone

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

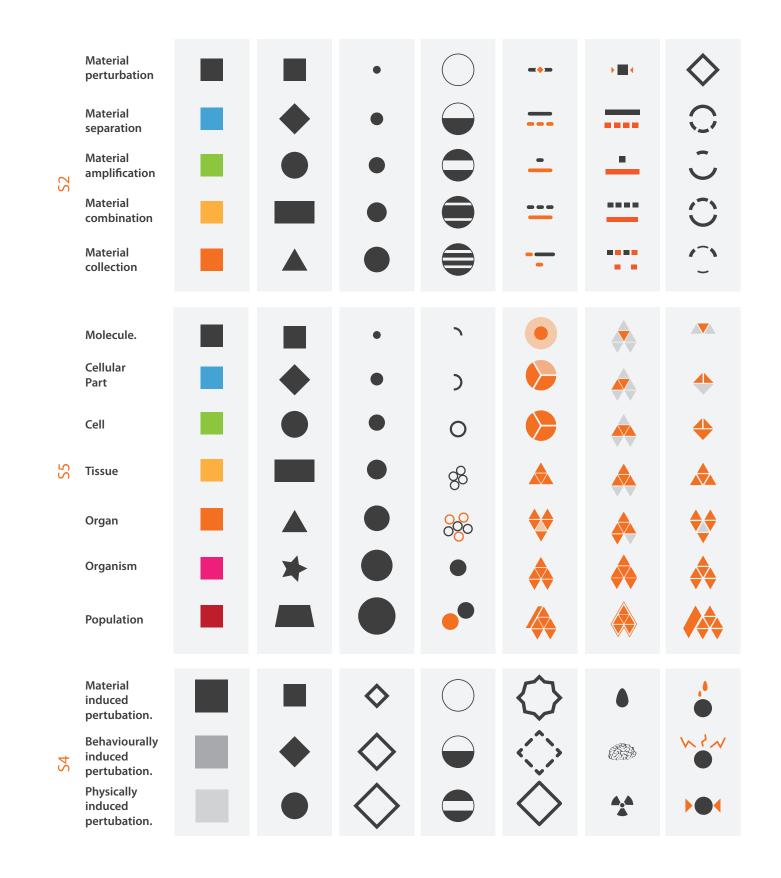




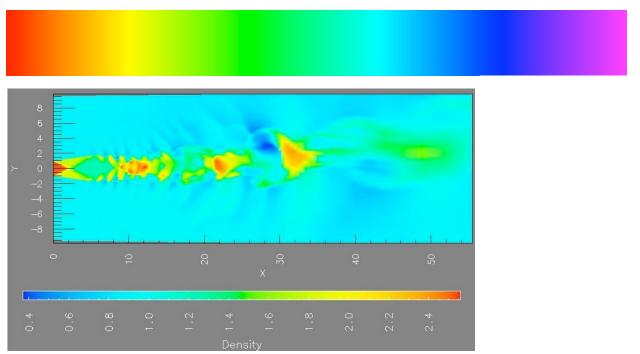
[Cinteny: flexible analysis and visualization of synteny and genome rearrangements in multiple organisms. Sinha and Meller. BMC Bioinformatics, 8:82, 2007.]

Glyphs

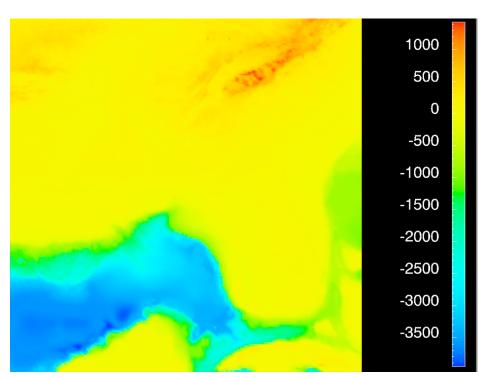
- glyphs: composite objects
 - -internal structure with multiple marks
- alternative to color coding
 - or coding with any single channel



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



[A Rule-based Tool for Assisting Colormap Selection. Bergman,. Rogowitz, and. Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. http://www.research.ibm.com/people/l/lloydt/color/color.HTM]

problems

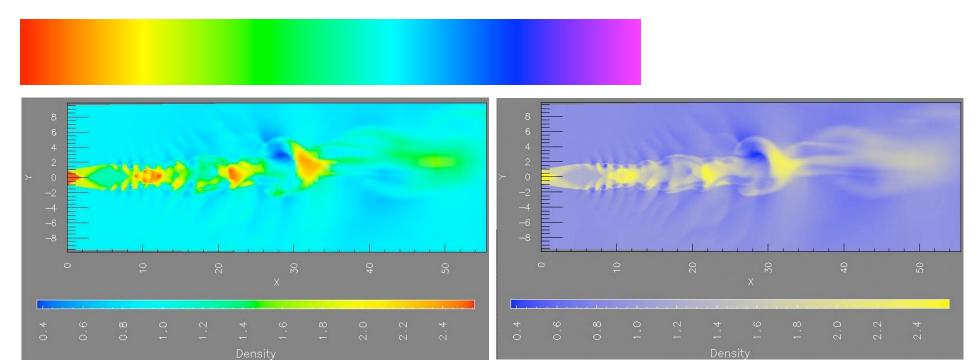
- -perceptually unordered
- -perceptually nonlinear

benefits

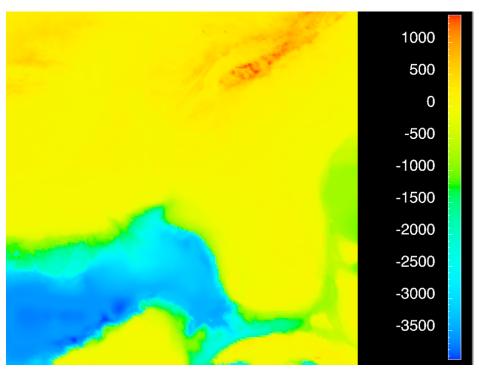
-fine-grained structure visible and nameable

alternatives

–large-scale structure: fewer hues



[A Rule-based Tool for Assisting Colormap Selection. Bergman,. Rogowitz, and. Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. http://www.research.ibm.com/people/l/lloydt/color/color.HTM]

problems

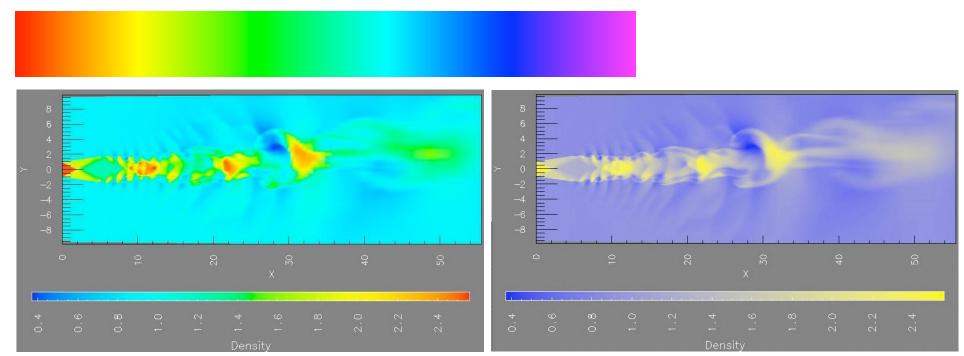
- -perceptually unordered
- -perceptually nonlinear

benefits

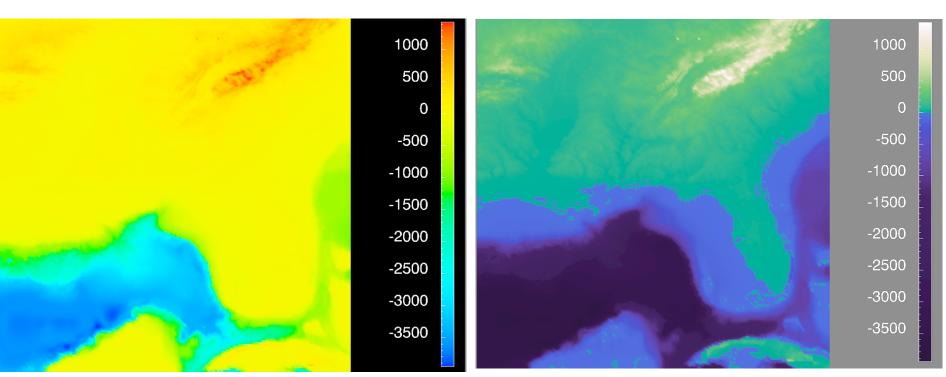
-fine-grained structure visible and nameable

alternatives

- –large-scale structure: fewer hues
- -fine structure: multiple hues with monotonically increasing luminance [eg viridis R/python]



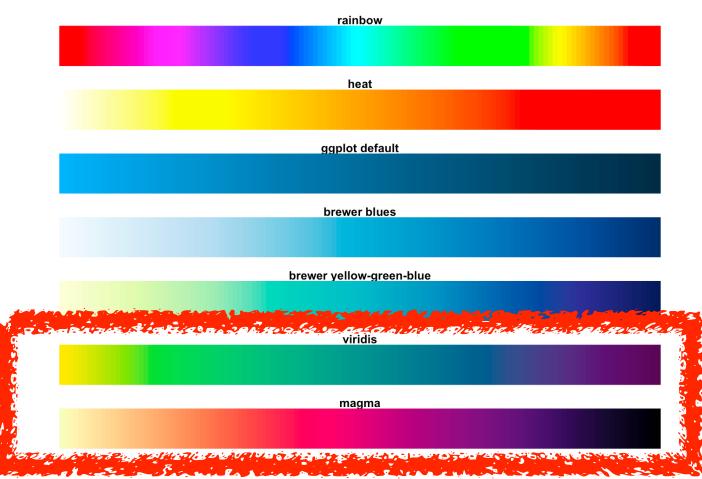
[A Rule-based Tool for Assisting Colormap Selection. Bergman,. Rogowitz, and Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



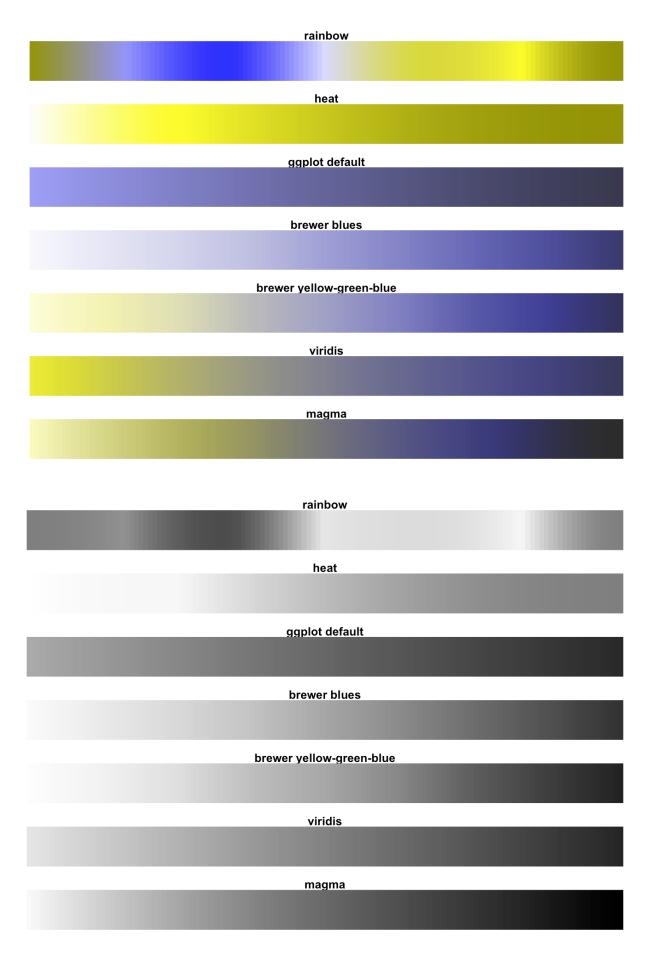
[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. http://www.research.ibm.com/people/I/lloydt/color/color.HTM]

Viridis

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



https://cran.r-project.org/web/packages/viridis/vignettes/intro-to-viridis.html



problems

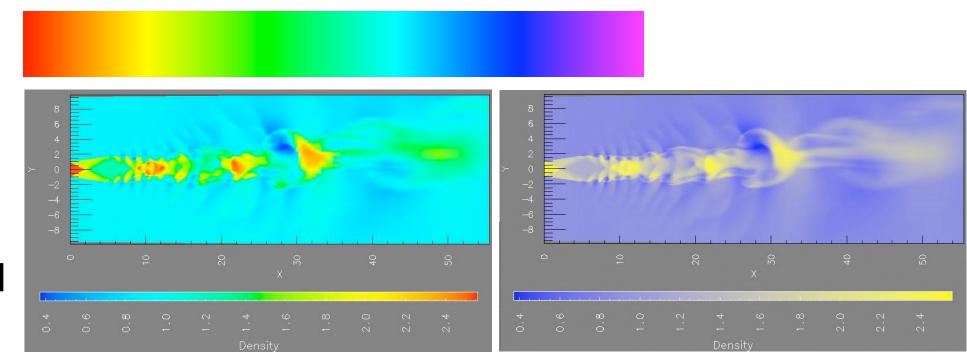
- -perceptually unordered
- -perceptually nonlinear

benefits

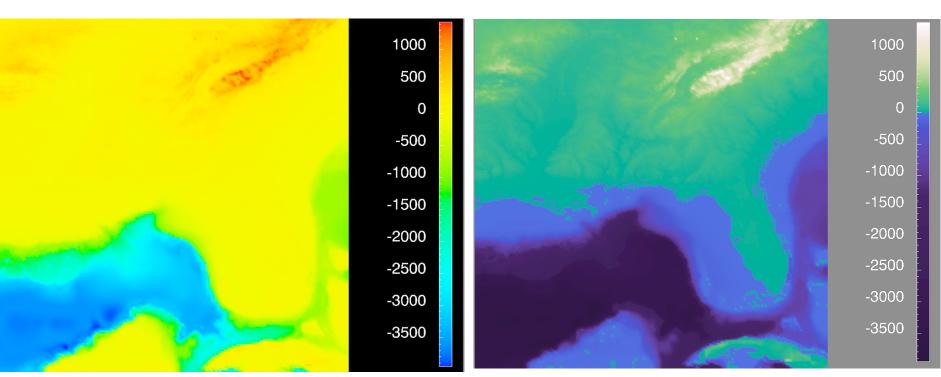
-fine-grained structure visible and nameable

alternatives

- -large-scale structure: fewer hues
- -fine structure: multiple hues with monotonically increasing luminance [eg viridis R/python]
- -segmented rainbows for binned or categorical



[A Rule-based Tool for Assisting Colormap Selection. Bergman,. Rogowitz, and. Treinish. Proc. IEEE Visualization (Vis), pp. 118–125, 1995.]



[Why Should Engineers Be Worried About Color? Treinish and Rogowitz 1998. http://www.research.ibm.com/people/I/lloydt/color/color.HTM]

More color resources: Muth

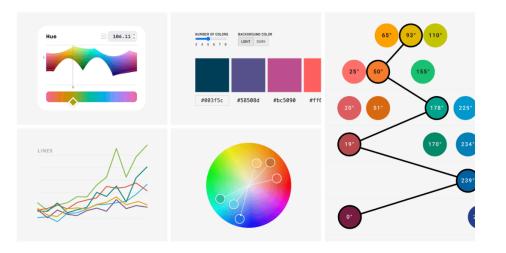
- DataWrapper guidance on color palette creation by Lisa Charlotte Muth https://blog.datawrapper.de/create-good-color-palettes/
 - -lots of practical advice, easy to understand

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Color in data vis
20 min
December 11th, 2024

How to find & create good color palettes





If you created a chart, chances are high you now want beautiful, accessible, appropriate colors for it. We already covered how to pick more beautiful colors for your data visualizations and what to consider when creating a color palette for your organizational style guide. The following article is more practical. It explains which tools and approaches you can use to find, come up with, or extend a color palette:

Where to find good color palettes:

Use the default colors in your tool | Use color palettes created by others for data visualization | Use color palettes created by others | Pick colors from other data visualizations | Pick colors from other works

How to create new color palettes:

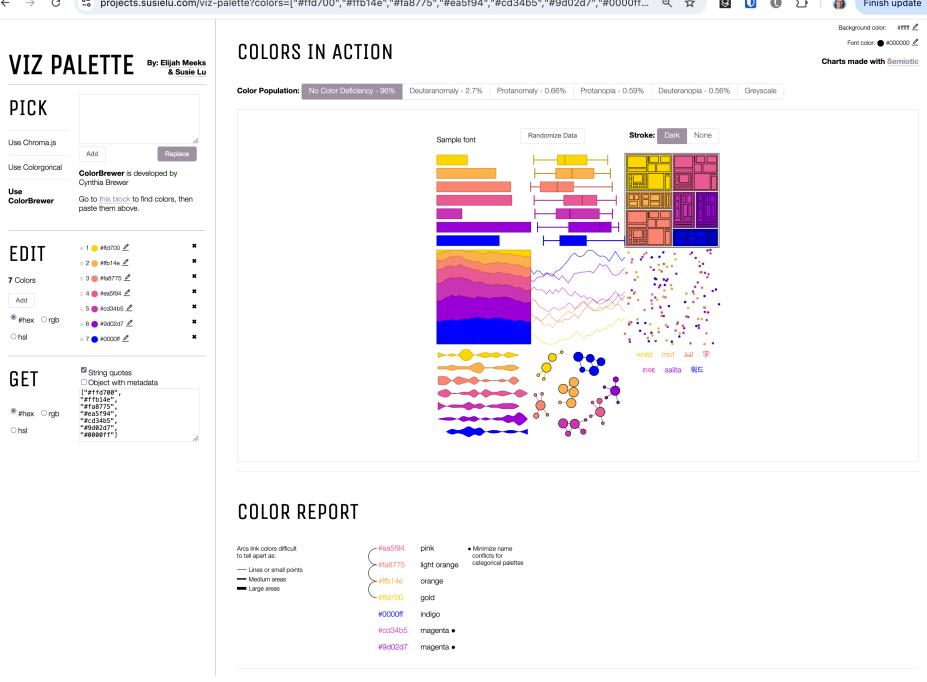
Use a tool that generates color palettes for data visualization | Use a palette generator based on the color wheel | Ask an LLM (like ChatGPT or Claude) | Cycle through color palette generators | Pick colors from a gradient | Define lightness and saturation first, then pick the hue | Don't use extra tools

How to extend an existing color palette:

More color resources: Viz Palette

• Viz Palette, by Elijah Meeks and Susie Lu https://projects.susielu.com/viz-palette

- colorblindness checks
- examples for different mark sizes/types
- report on distinguishability including names

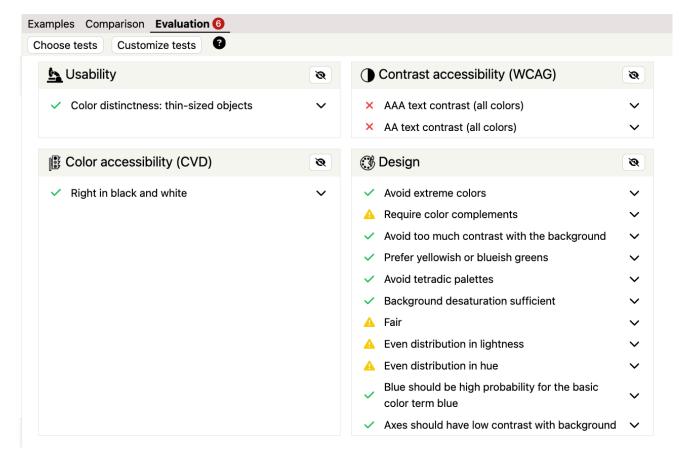


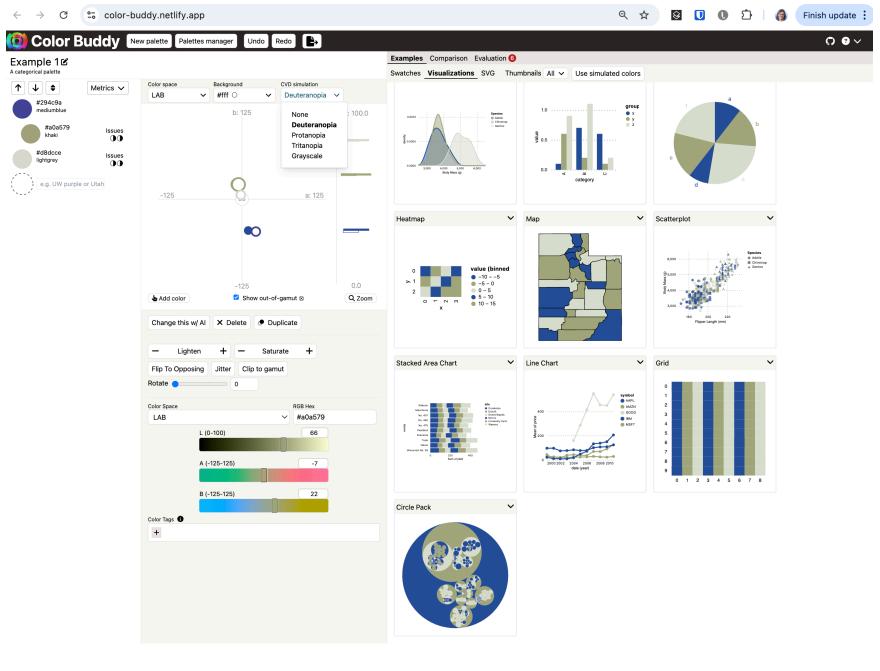
More color resources: Color Buddy

Color Buddy, by Andrew McNutt and Maureen Stone

https://color-buddy.netlify.app/

- check against colorblindness
- check different mark types/sizes
- evaluate against design guidelines

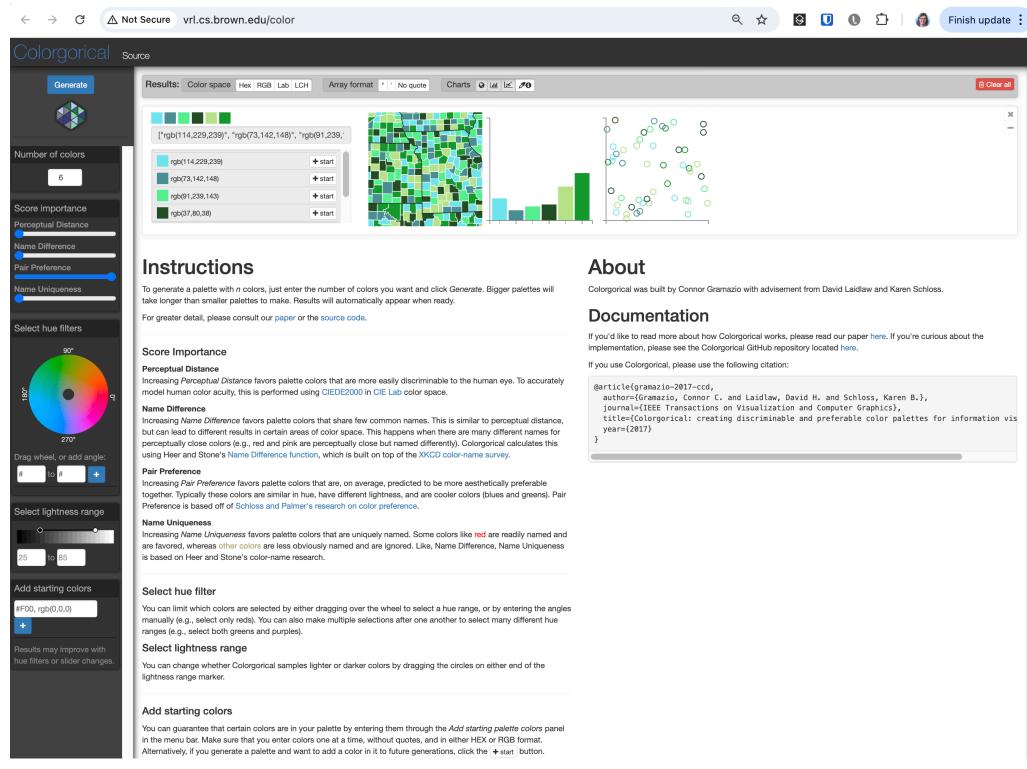




More color resources: Colorgorical

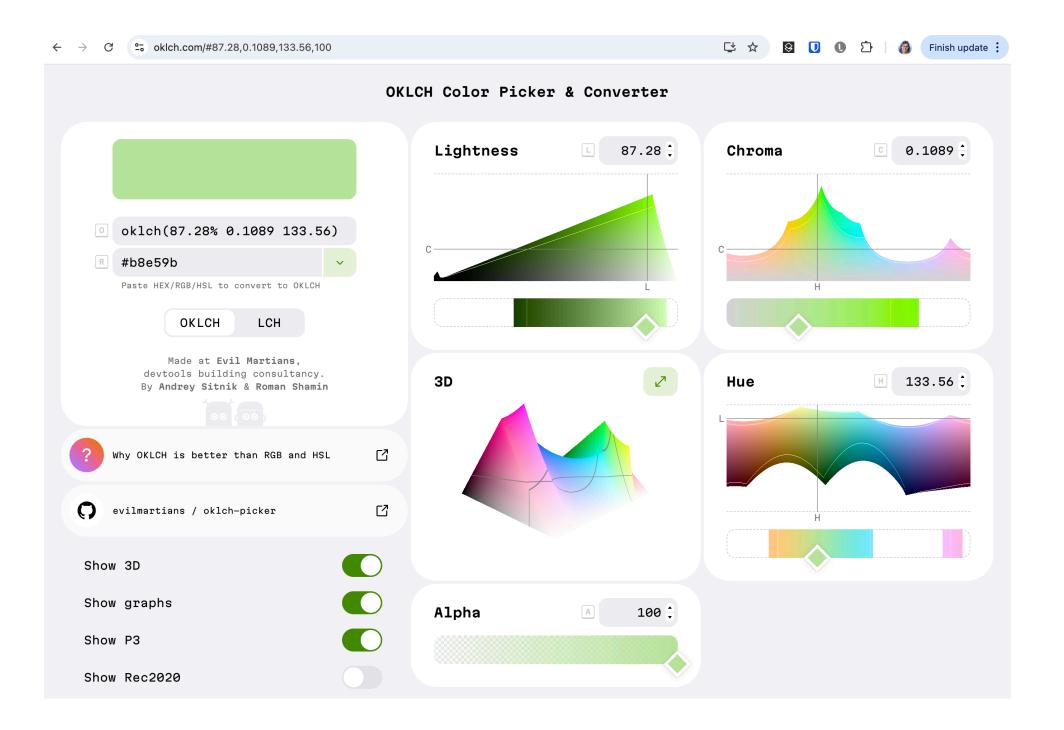
 Colorgorical, by Connor Gramazio http://vrl.cs.brown.edu/color

categorical color, including name



More color resources: oklch.com

- oklch perceptual space color picker/converter, by Andrey Sitnik & Roman Shamin (Evil Martians) https://oklch.com/
 - -inspect color space itself



More color resources: ColorBrewer

 ColorBrewer, by Cynthia Brewer colorbrewer2.com

