

Marks Revisited: Beyond Bertin

Tamara Munzner

Department of Computer Science
University of British Columbia

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<http://www.cs.ubc.ca/~tmm/talks.html#mit24>



 **[@tamara@vis.social](https://mstdn.social/@tamaramunzner)**

 **[@tamaramunzner](https://twitter.com/tamaramunzner)**

Co-conspirators

- Richard Brath, Uncharted Software
- UBC infovis course TAs
 - Mara Solen, Francis Nguyen
 - Steve Kasica, Ryan Smith

Marks and channels: Foundational model

- decompose visual encoding into marks & channels

- marks

- geometric primitives
 - represent items

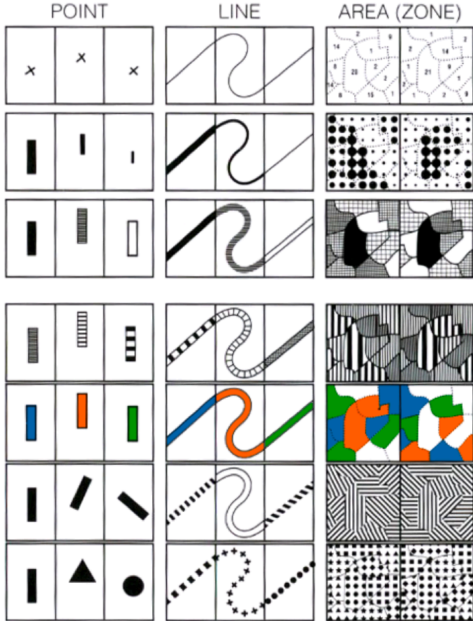
- channels

- control appearance of marks
 - representing attributes

- widely used

- Bertin 1967

- Semiology of Graphics



Marks

→ Points



→ Lines



→ Areas



Channels

→ Position

→ Horizontal



→ Vertical



→ Both



→ Color



→ Shape

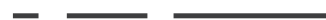


→ Tilt



→ Size

→ Length



→ Area



→ Volume

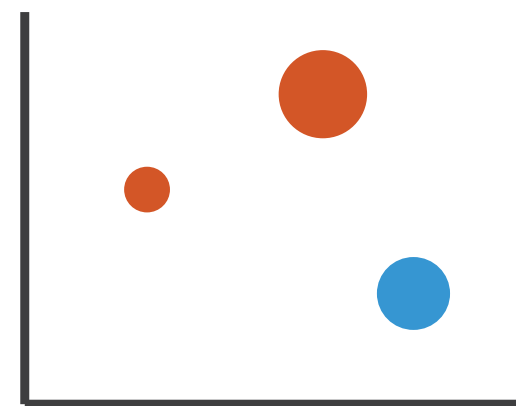
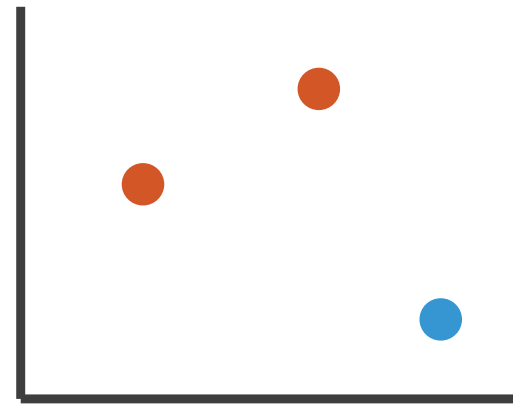
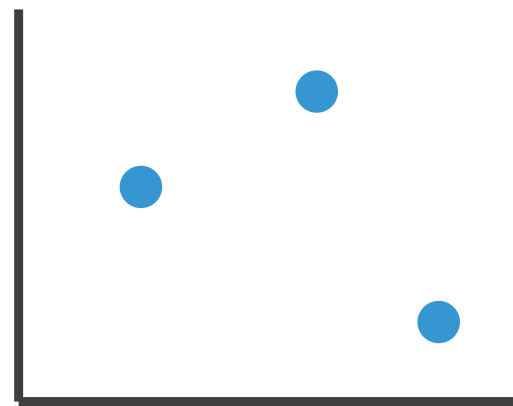
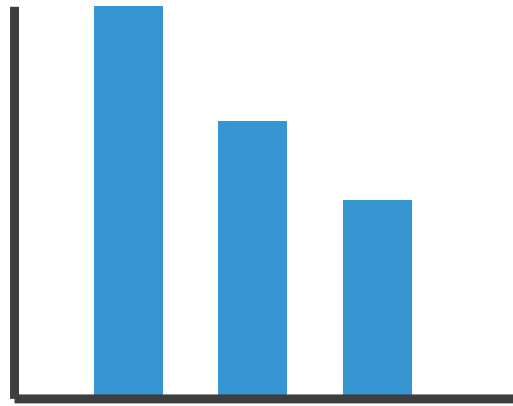


Talk outline

- explain current marks & channels model
- walk through many questions that arise when teaching it
- present preliminary ideas towards an alternative model

Visual encoding

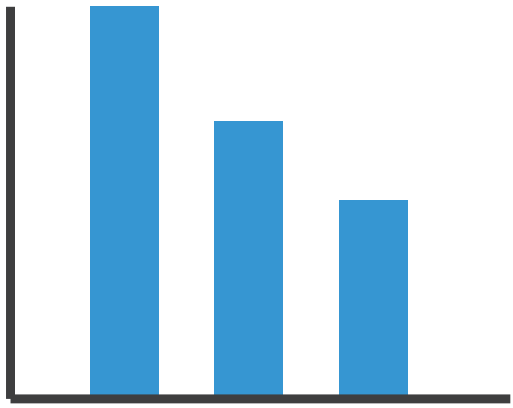
- analyze idiom structure as combination of marks and channels



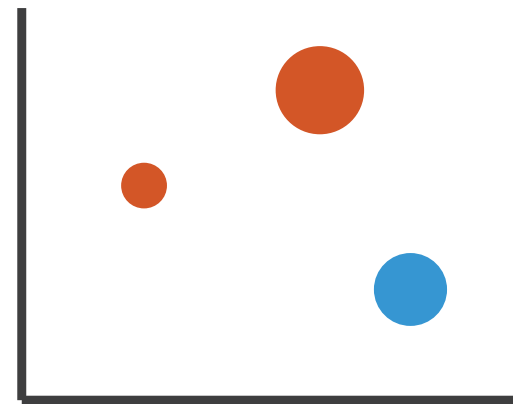
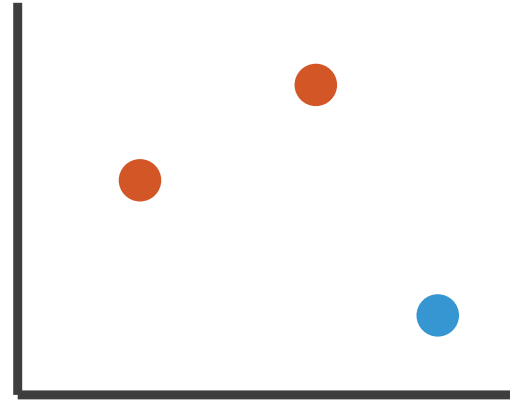
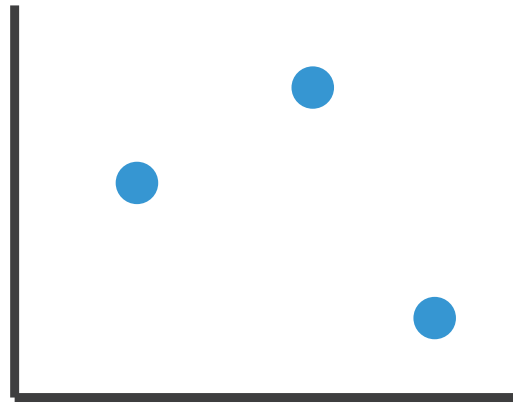
Visual encoding

- analyze idiom structure as combination of marks and channels

idiom: bar chart



1 channel:
vertical position

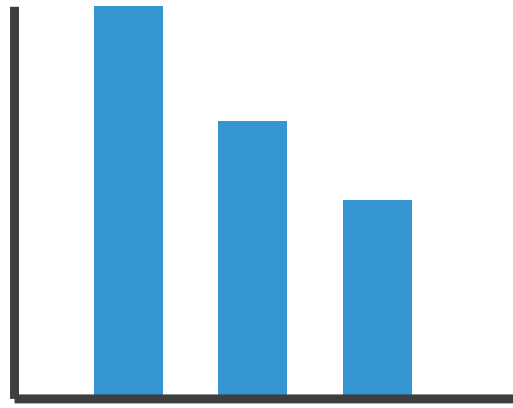


mark: line

Visual encoding

- analyze idiom structure as combination of marks and channels

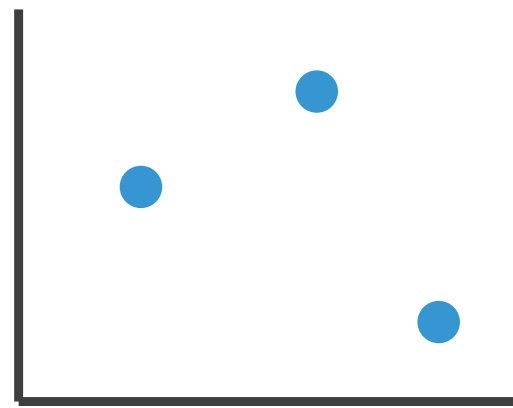
idiom: bar chart



1 channel:
vertical position

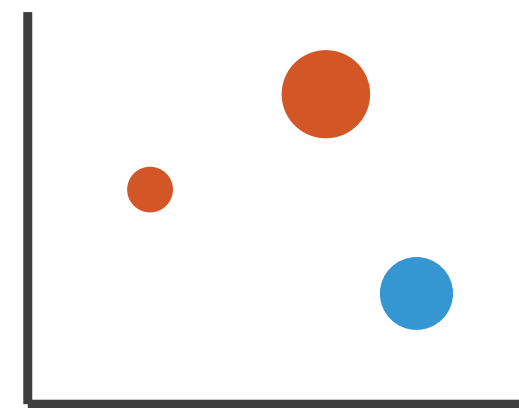
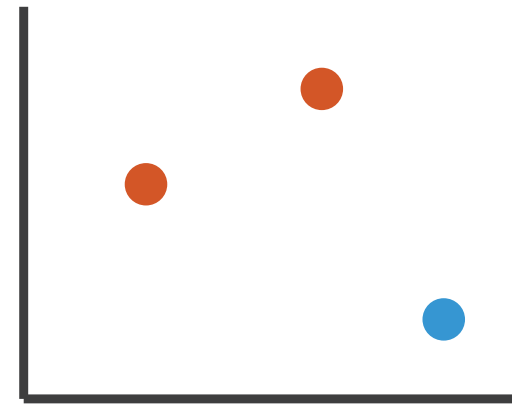
mark: line

idiom: scatterplot



2 channels:
vertical position
horizontal position

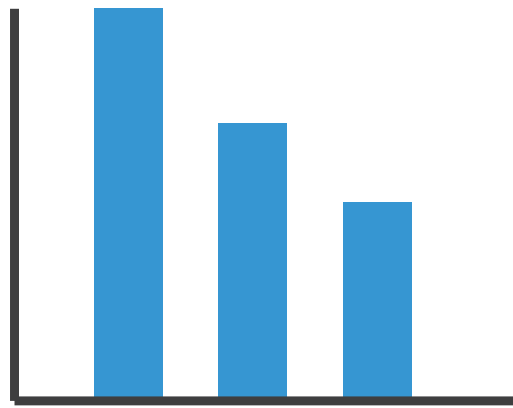
mark: point



Visual encoding

- analyze idiom structure as combination of marks and channels

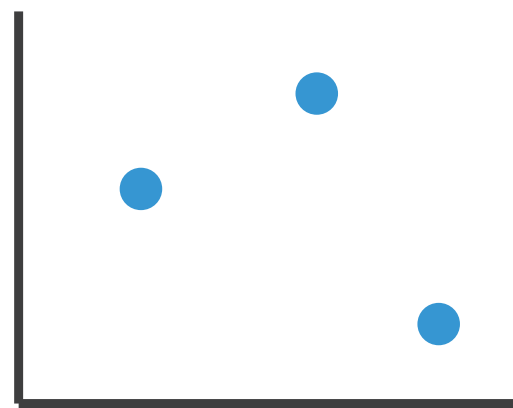
idiom: bar chart



1 channel:
vertical position

mark: line

idiom: scatterplot



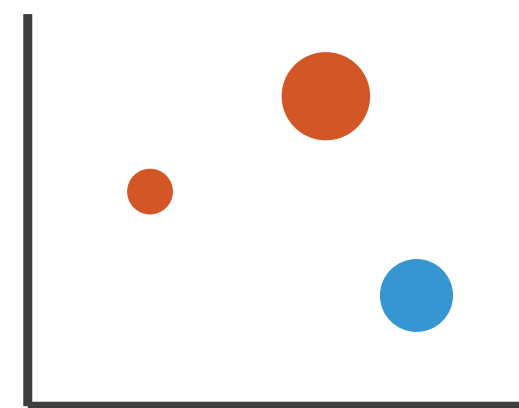
2 channels:
vertical position
horizontal position

mark: point



3 channels:
vertical position
horizontal position
color hue

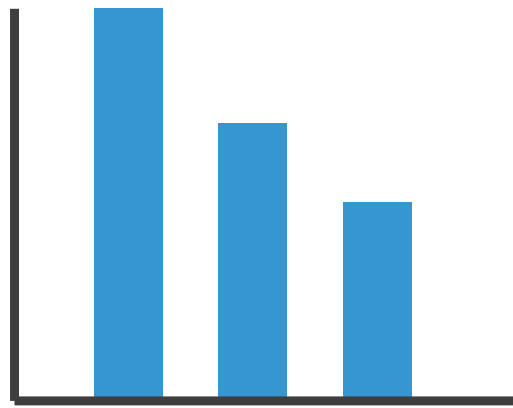
mark: point



Visual encoding

- analyze idiom structure as combination of marks and channels

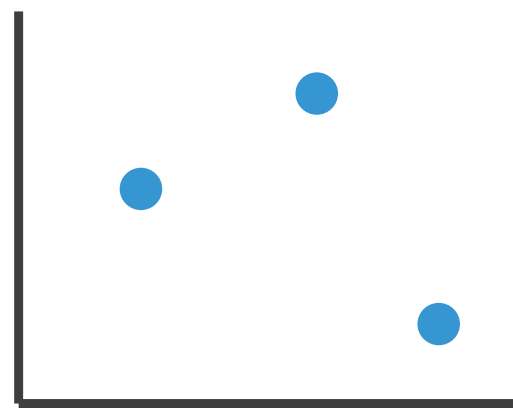
idiom: bar chart



1 channel:
vertical position

mark: line

idiom: scatterplot



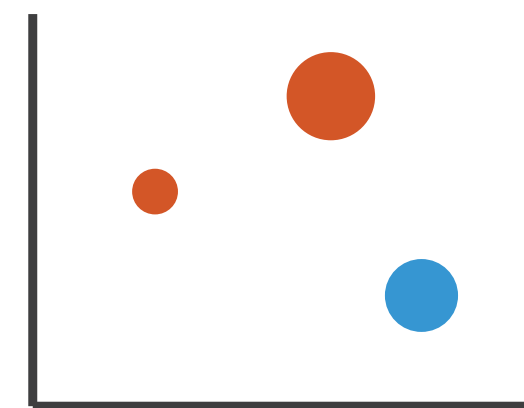
2 channels:
vertical position
horizontal position

mark: point



3 channels:
vertical position
horizontal position
color hue

mark: point



4 channels:
vertical position
horizontal position
color hue
size (area)

mark: point

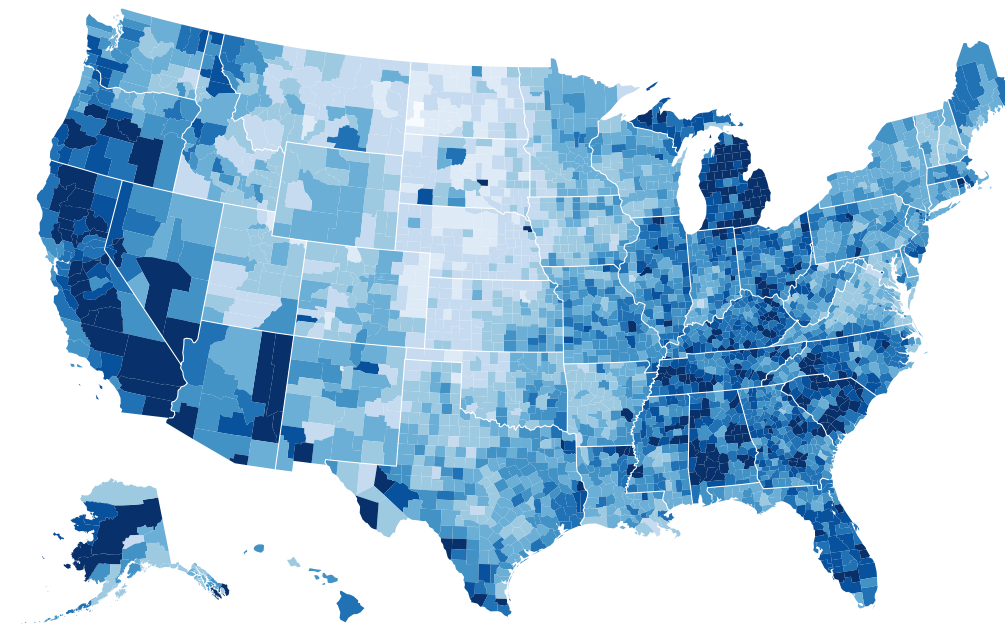
Visual encoding: Spatial data

- marks for given spatial data (boundaries)

idiom: choropleth map

channels:
position
color (saturation)

mark: area



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

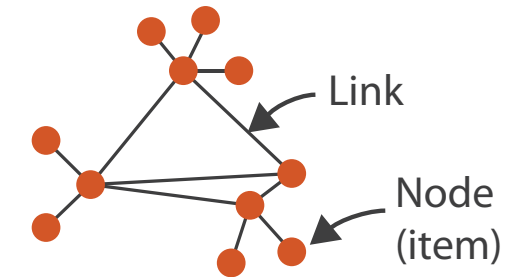
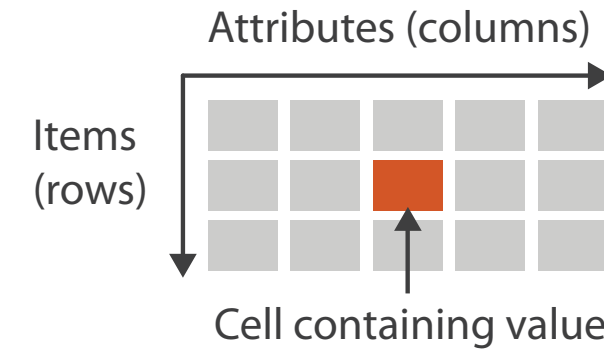
Visual encoding: Network data

- marks for items vs marks for links

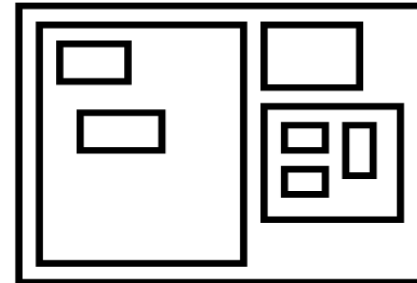
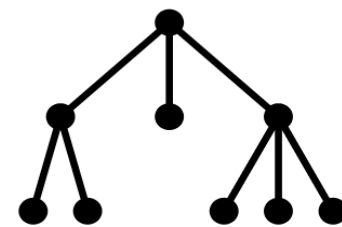
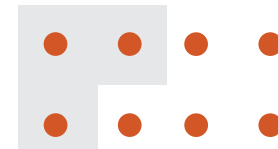
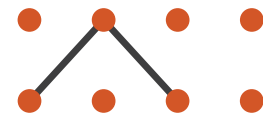
➔ Dataset Types

➔ Tables

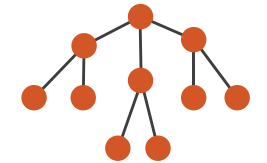
➔ Networks



➔ Connection ➔ Containment



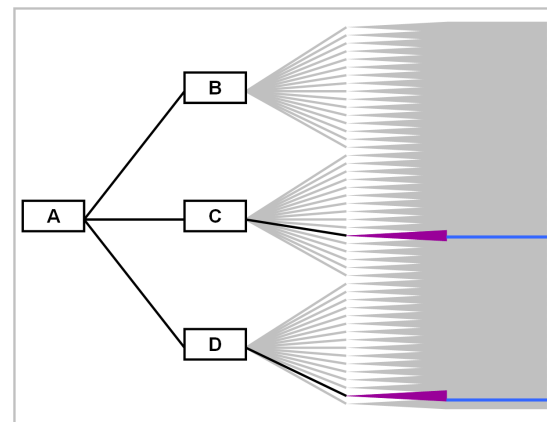
➔ Trees



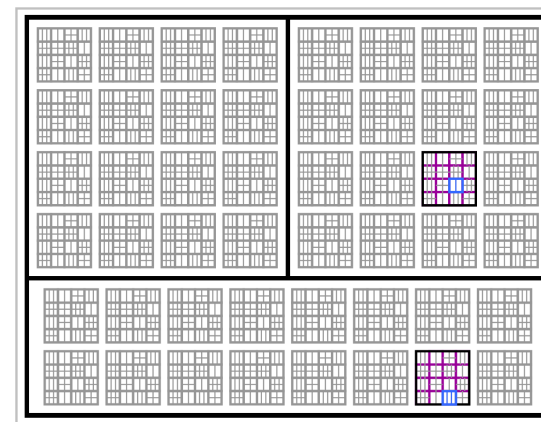
idiom: node-link diagram

channel:
position

marks:
point for items/nodes,
connection line for links



Node-Link Diagram



Treemap

position

marks:
area for items/nodes,
containment area for parent-child links

Why analyze visual encodings?

- marks & channels model is a **design space**
 - **descriptive** power: ability to describe significant range of existing examples
 - **evaluative** power: ability to help assess multiple design alternatives
 - **generative** power: ability to help designers create new designs
 - **criteria:** *Michel Beaudoin-Lafon, Designing Interaction, not Interfaces. AVI 2004.*
- many names: taxonomies, typologies, classifications, frameworks, models...
 - delineate: axes / dimensions / categories
 - that are cross-cutting / independent / orthogonal
- design spaces help us reason
 - impose systematic & actionable structure on set of possibilities for specific problem
 - to support reasoning about design choices
 - capture the key variables at play
 - increase cognitive efficiency & support inferences by grouping similar instances together to facilitate reasoning about classes

Design spaces in visualization: continuing theme

The Structure of the Information Visualization Design Space

Stuart K. Card and Jock Mackinlay
Xerox PARC

Exploring the Design Space of Composite Visualization

Waqas Javed* Niklas Elmqvist†

2366

IEEE TRANSACTIONS ON VISUALIZATION AND COMPUTER GRAPHICS, VOL. 19, NO. 12, DECEMBER 2013

A Design Space of Visualization Tasks

Hans-Jörg Schulz, Thomas Nocke, Magnus Heitzler, and Heidrun Schumann

A Design Space of Vision Science Methods for Visualization Research

Madison A. Elliott, Christine Nothelfer, Cindy Xiong, and Danielle Albers Szafir



Fig. 1. Overview of design space of experimental methods. We present a four component design space to guide researchers in creating visualization studies grounded in vision science research methods.

1 INTRODUCTION

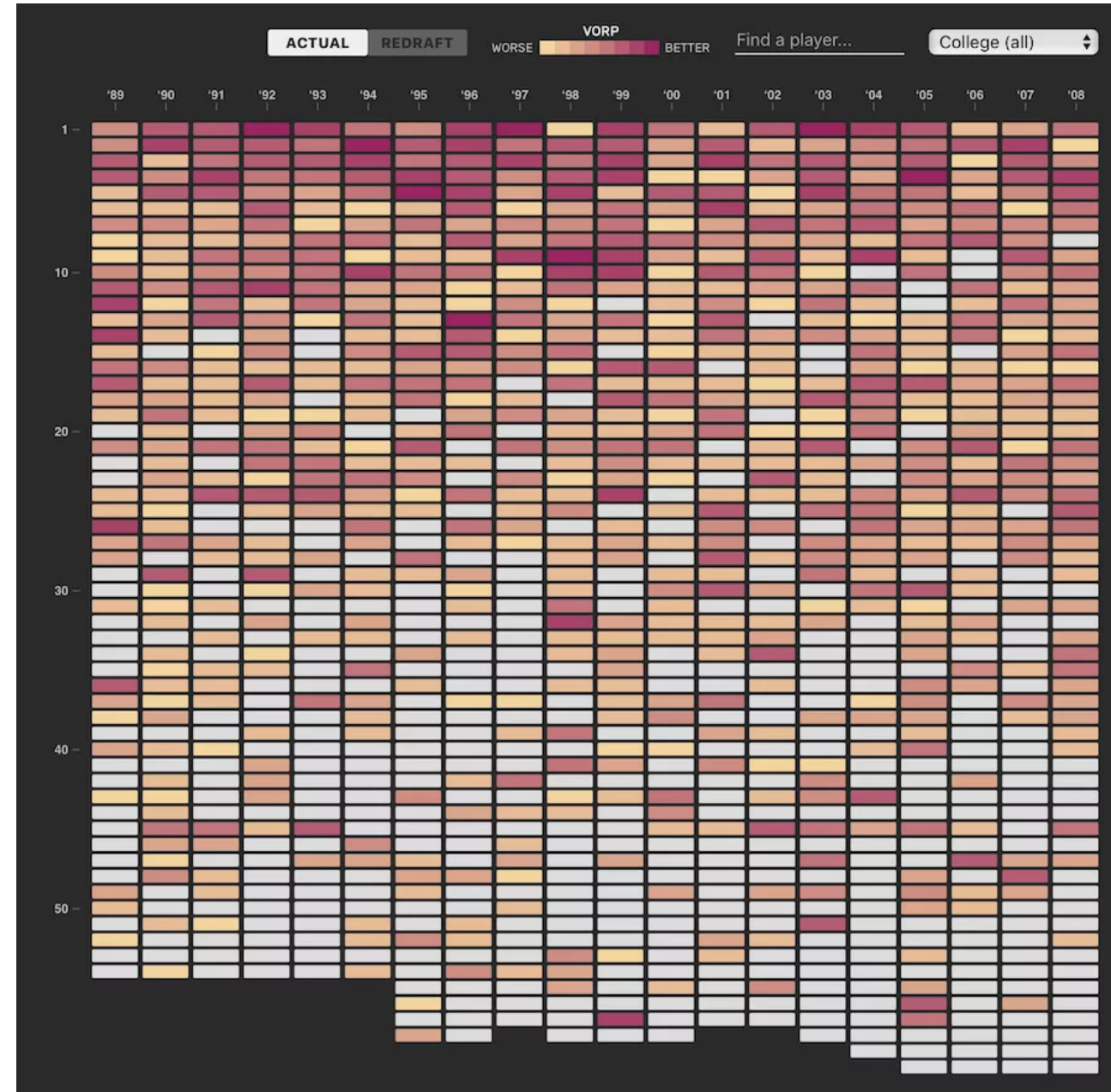
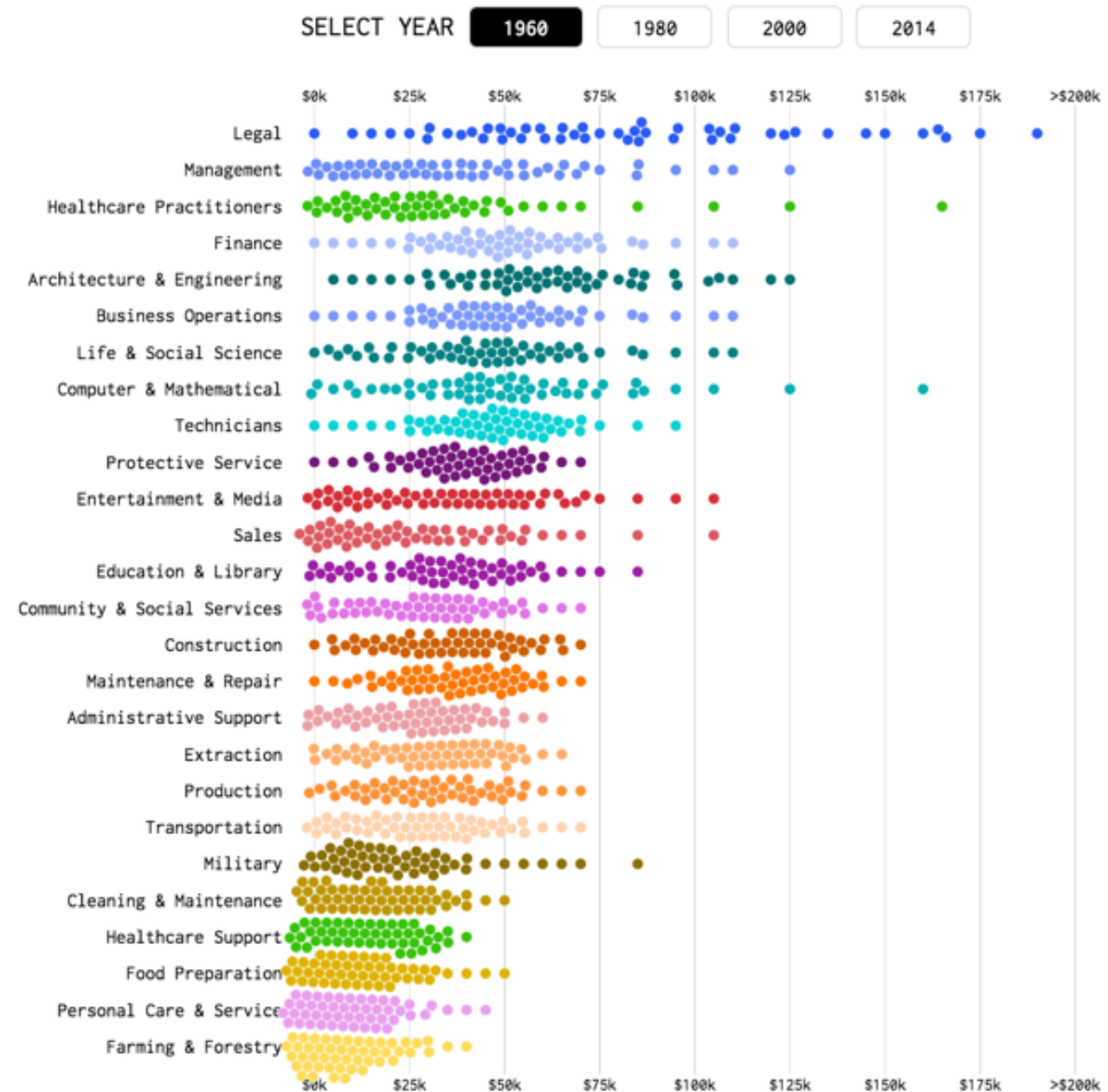
As the field of visualization sets its

Teaching design space: analyze visual encoding & map to data

- analyze existing encoding with marks & channels
 - Visual channels used?
 - Channel X encodes attribute Y
 - Channel X encodes attribute Y
 - Marks used?
 - Mark of type X encodes item Y
 - Mark of type X encodes item Y

Teaching: Bertini in-class exercises, catalyst for questions

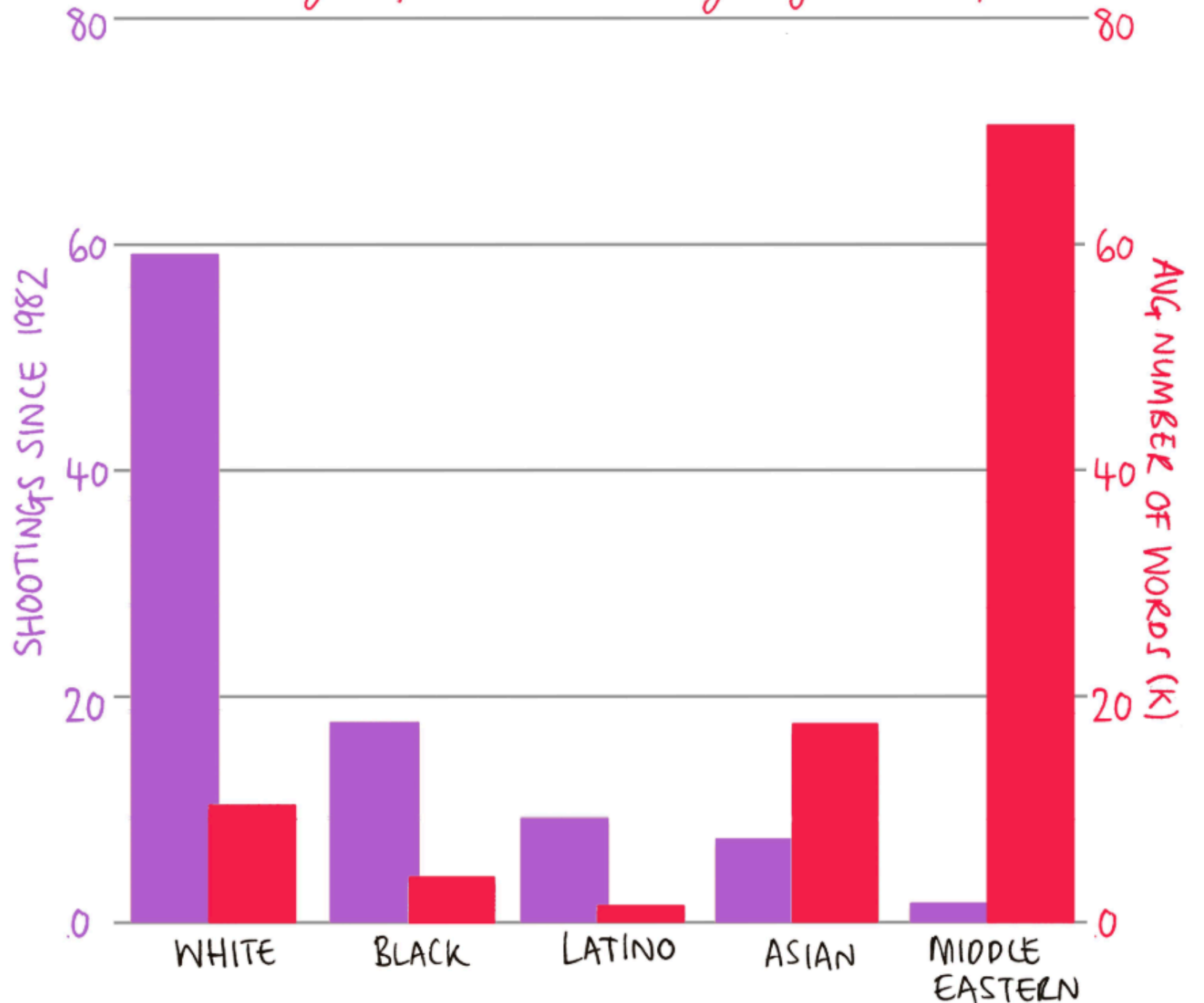
- decoding marks & channels
 - <https://enrico.bertini.io/teaching>



Quiz: Name marks/channels

- Shooting Media Coverage
- marks
 - A: points
 - B: lines
 - C: areas
- channels
 - A: position
 - B: color
 - C: motion
 - D: area
 - E: angle

Mass Shootings By Race Of Shooter
NYTimes Coverage Of Mass Shootings By Race Of Shooter



<https://twitter.com/MonaChalabi/status/1158779046693679106?s=20>

Quiz: Name marks/channels

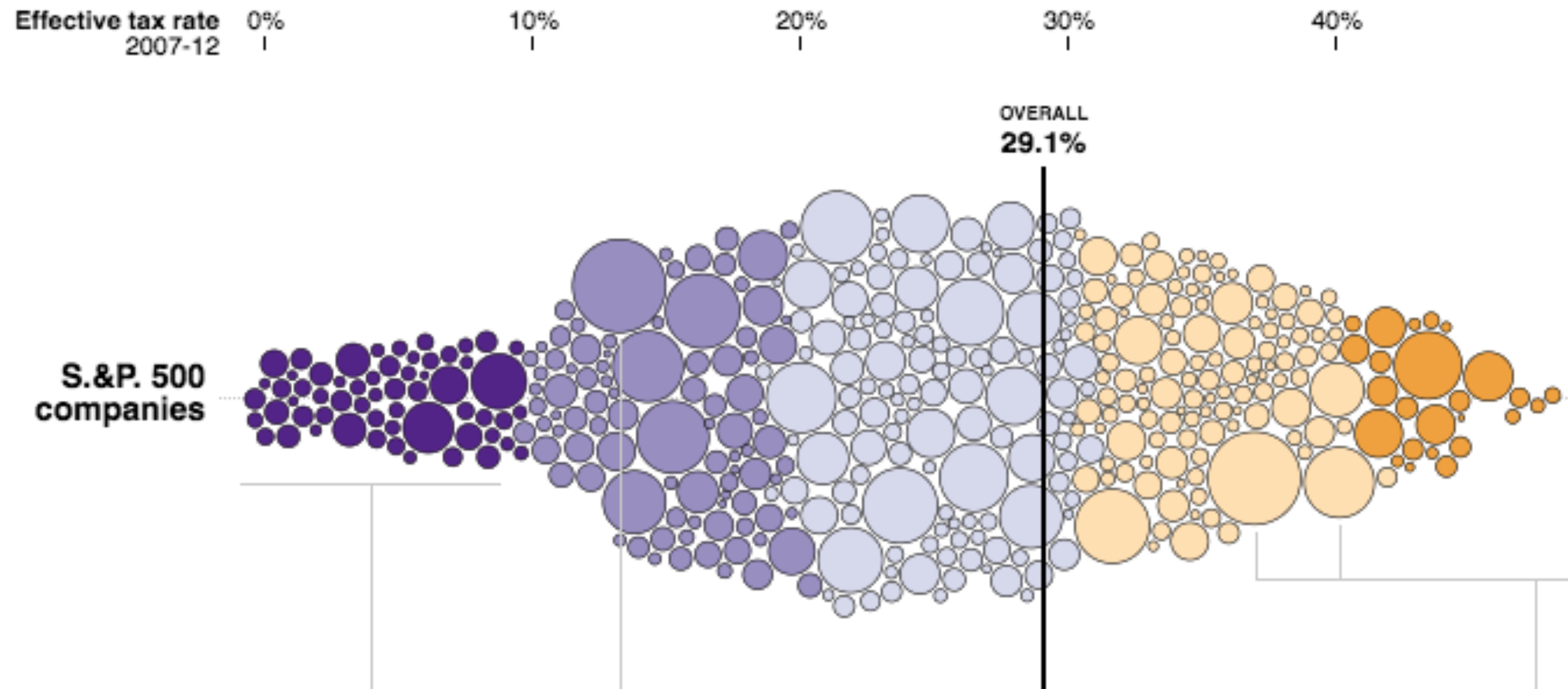
- Tax Rates

- marks

- A: points
- B: lines
- C: areas

- channels

- A: position
- B: color
- C: motion
- D: area
- E: angle



Many, many questions

- so what?
 - evidence that this design space could be improved!

Channels: Model evolves, heavily studied

- effectiveness rankings
- expressiveness matches, data & task

Crowdsourcing Graphical Perception: Using Mechanical Turk to Assess Visualization Design

Jeffrey Heer and Michael Bostock

ABSTRACT

Understanding perception is critical to visualization design. With its low cost and scalability, crowdsourcing presents an attractive option for evaluating the design space of visualizations; however, it first requires a model of perception. In this paper, we assess the viability of Amazon Mechanical Turk as a platform for graphical perception experiments. We replicate previous studies of spatial encoding (e.g., dot plots, cartograms) and on chart size and gridline visibility. Our results demonstrate that crowdsourced perception experiments are viable and contribute new insights for visualization design. Lastly, we report cost and performance metrics from our experiments and distill recommendations from crowdsourced studies.

ACM Classification: H5.2 [Information Systems]: User Interfaces—Evaluation/Negotiation Support

General Terms: Experimentation, Human Factors

Keywords: Information visualization, human-computer interaction, user study, evaluation, Mechanical Turk, crowdsourcing

INTRODUCTION

“Crowdsourcing” is a relatively new phenomenon where web workers complete one or more small tasks for micro-payments on the order of \$0.01. Such services are increasingly attractive as a low-cost means of conducting user studies. They lower the cost of recruiting participants, provide almost immediate access to hundreds of users. Similarly, by reducing the burden on the subject pool is greatly increased and diversified.

The reduced cost structure of crowdsourcing is particularly attractive in visualization, where the space of possible visual encodings is large and complex [2, 7, 10, 19, 27, 34]. Crowdsourcing experiments to canvas a wide range of visual encodings, effectively swapping experimental

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IEEE TRANSACTIONS ON VISUALIZATION AND COMPUTER GRAPHICS, VOL. 20, NO. 12, DECEMBER 2014

1943

Ranking Visualizations of Correlation Using Weber's Law

Lane Harrison, Fumeng Yang, Steven Franconeri, Remco Chang

Abstract—Despite years of research on identifying the best visualization for a given task, we conduct a large scale (n=1687) experiment to assess how well used visualizations can be modeled using Weber's Law. We find that visualization by establishing that: 1) correlation judgment precision follows Weber's Law, 2) correlation judgment precision models provide a concise means to compare visualizations.

Index Terms—Perception, Visualization

1 INTRODUCTION

The theory and design of information visualization have advanced in part because of the way since Bertin's seminal work on the design of visualization research has led to the development of design principles [5, 24] that aid the designer in choosing a visualization based on general data characteristics such as type. Unfortunately, many aspects of visualization are more art than science. For example, given a set of data, there are almost always multiple theoretically valid and therefore difficult to choose between. Beyond selecting a visualization, one must also account for many other aspects of the design such as color, shape, and size. These considerations such as context, and user expectations, make it tremendously difficult for even experts to choose the most accurate and appropriate visualization.

One method for objectively identifying the best visualization is to conduct multi-factor human-subject experiments, each design or usage consideration being a factor, often resulting in a large number of possible visualizations. These experiments produce actionable results that generalize beyond the scope of the experiment. A key question in the explanation as to why one visualization is better than another is that exhaustive comparative experiments are growing needs of the infovis community.

What is needed then are quantitative models of visualization that are generalizable beyond the specific design elements while still providing designers with tradeoffs between “valid” visualization elements challenge conventional wisdom in infovis [13], recent research has suggested that the psychology and cognitive science [12, 3] humans perceive certain data properties

CHI 2019 Paper

CHI 2019, May 4–9, 2019, Glasgow, Scotland, UK

Measuring the Separability of Shape, Size, and Color in Scatterplots

Stephen Smart

University of Colorado Boulder
Boulder, Colorado
stephen.smart@colorado.edu

Color Difference Comparisons

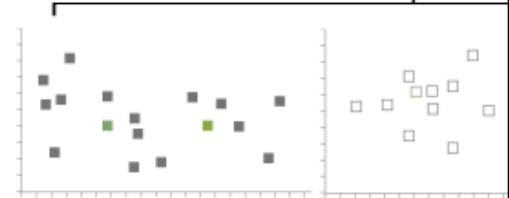


Figure 1: We examine how mark shape, size and color affect the separability of different shape categories.

ABSTRACT

Scatterplots commonly use multiple visual channels to code multivariate datasets. Such visualizations often use shape, size, and color as these dimensions are considered “separable”—dimensions represented by one channel do not significantly interfere with viewers’ abilities to perceive another. However, recent work shows that size significantly impacts color difference perceptions, raising broader questions about the separability of these

23 Jul 2021



Danielle Albers Szafir

Rethinking the Ranks of Visual Channels

Caitlyn M. McColeman*, Fumeng Yang*, Timothy F. Brady, and Steven Franconeri

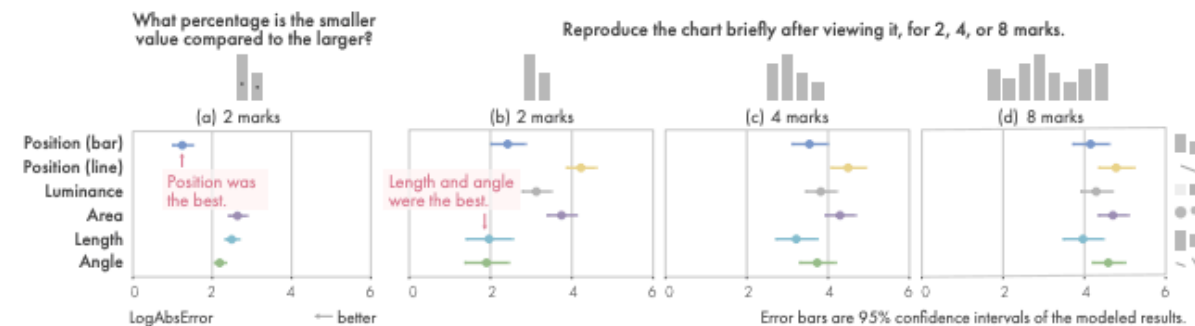
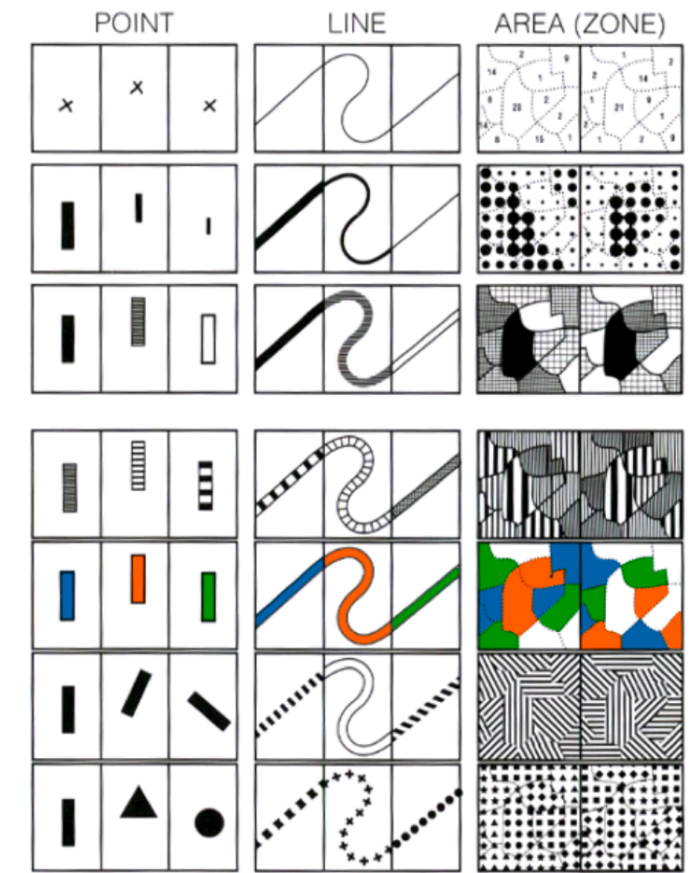


Fig. 1. One core guideline for data visualization design is that some visual channels offer better perceptual precision than others, drawing those precision estimates from two-value ratio judgment tasks [17]. (a) This figure depicts typical data (from [33], 50 participants) showing these judgments are more precise for position (e.g., bar graphs) than for area (e.g., bubble charts). We tested whether that ranking generalizes to the new task of reproducing 2 to 8 previously seen values, and analyzed reproduction bias, precision, and error using a Bayesian modeling approach. (b) This figure shows our modeled results (49 participants). The ranking did not hold, and other factors besides channel choice—like the number of values in the series—had an order of magnitude more influence on performance.

Marks: Model stays static

- model inherited from Bertin (Semiology of Graphics, 1967)
 - never questioned
- geometric motivation
 - geometric primitives have dimensions
 - how could we argue with math?!



→ Points



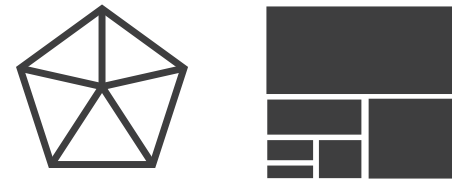
0D

→ Lines



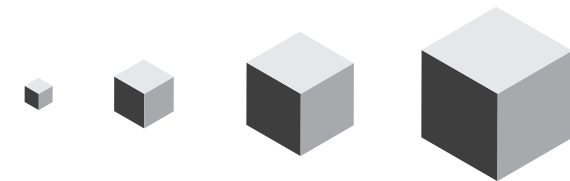
1D

→ Areas



2D

→ Volume



3D

What do marks do?

- idiom: pie chart

- **area** marks with angle channel: **2D area varies**

- separated & ordered radially, uniform length

- accuracy: area less accurate than rectilinear aligned line length

- task: part-to-whole judgements

- idiom: coxcomb chart

- marks with length channel: **ID length varies**

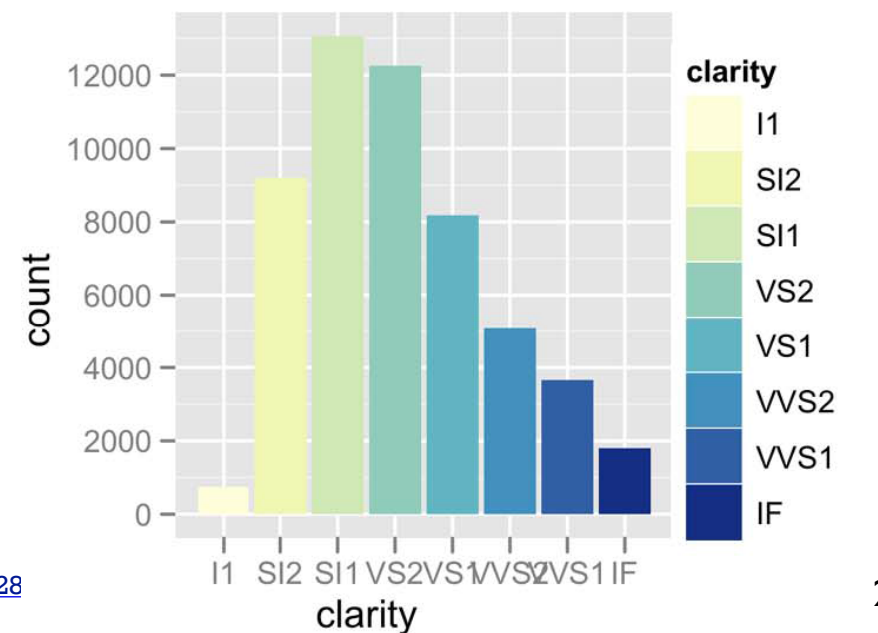
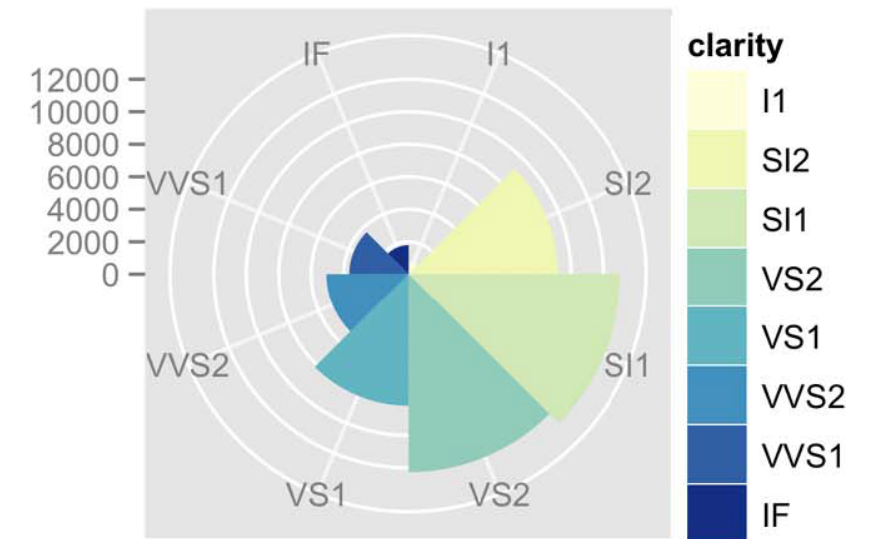
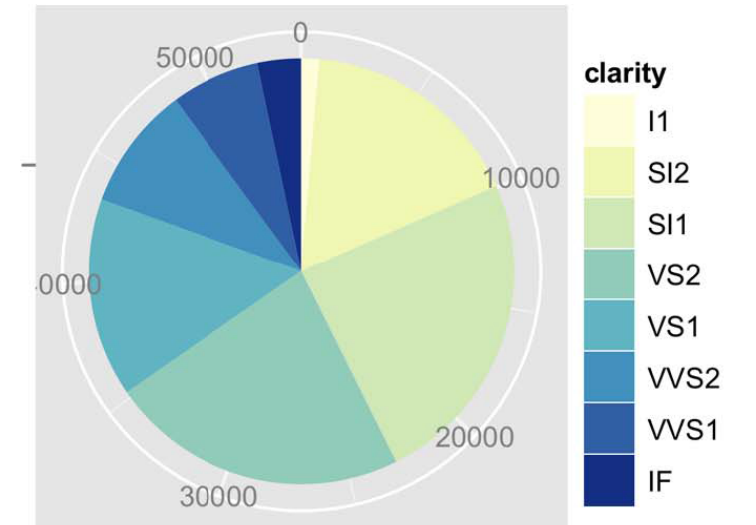
- separated & ordered radially, uniform width

- direct analog to radial bar charts

- what's the mark type?

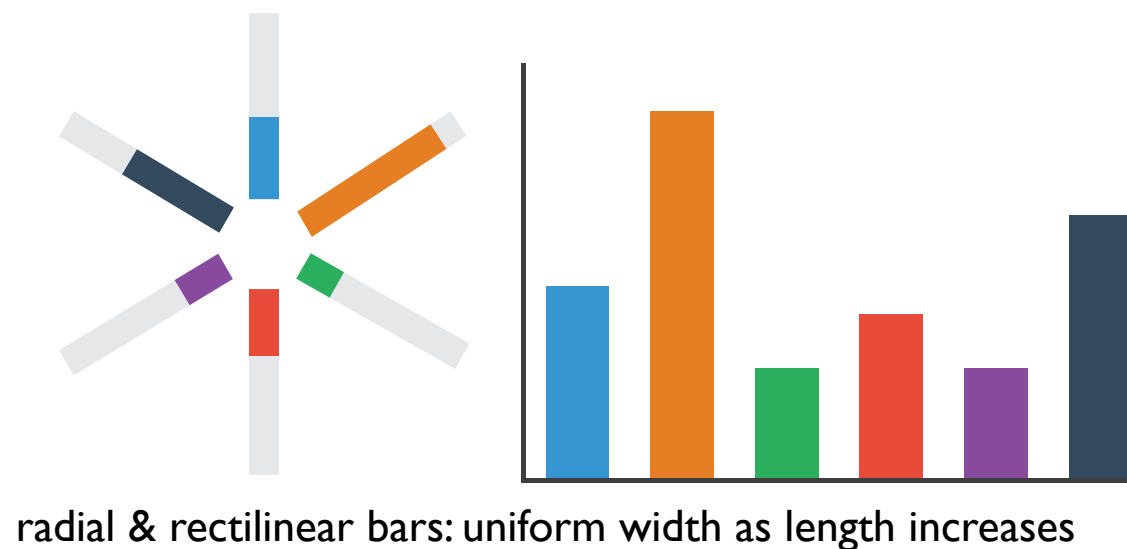
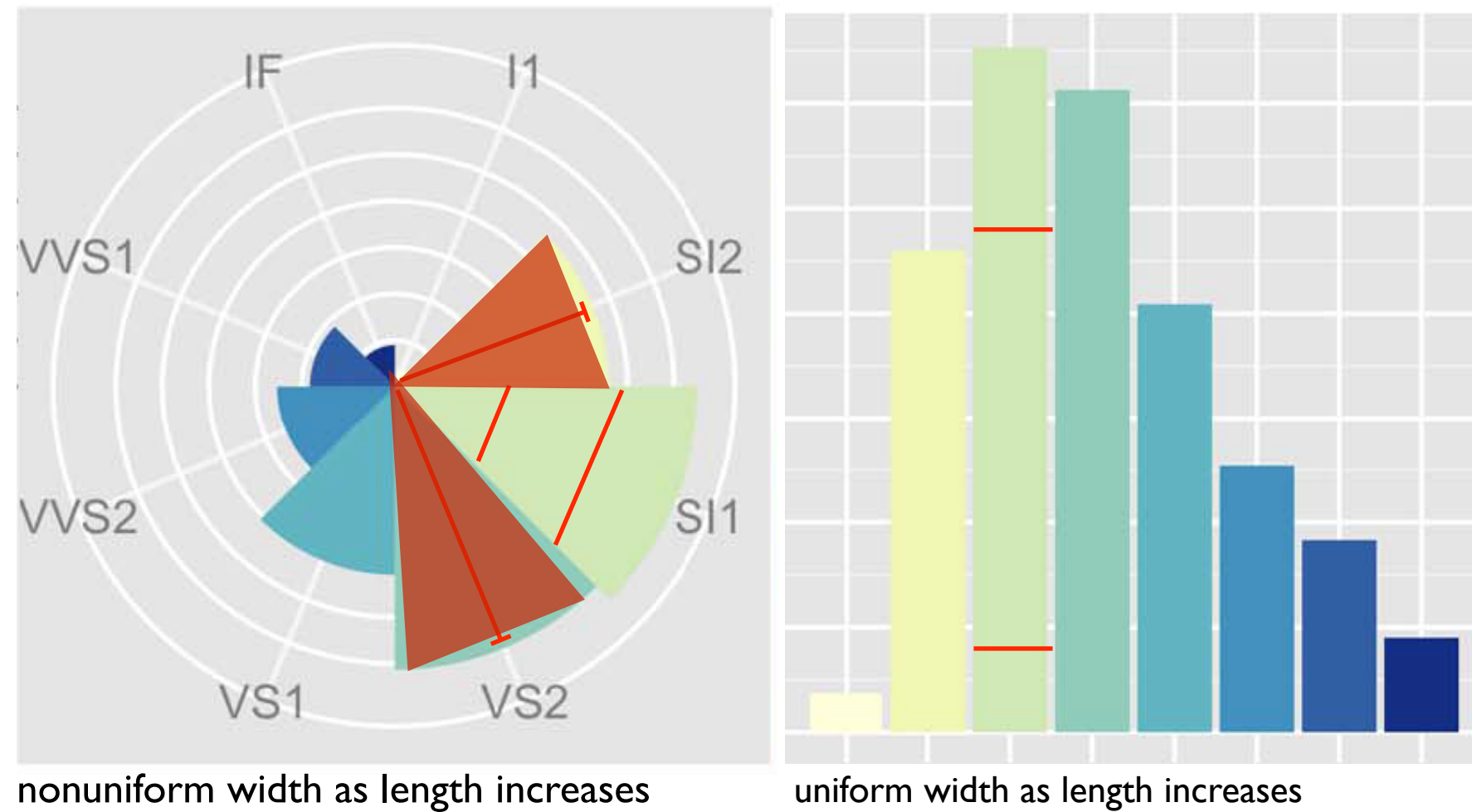
- line, because it's length coded?

- area, because area varies too?



Coxcomb channel analysis

- encode: **ID size (length)**
- decode/perceive: **2D area**
- nonuniform line/sector width as length increases
 - so area variation is nonlinear wrt mark length!
- bar chart safer
 - uniform width, so area is linear with mark length
 - both radial & rectilinear cases
- **mark type: encode or decode?**
 - infer designer intent
 - predict viewer response: if channels differ, which "wins"?

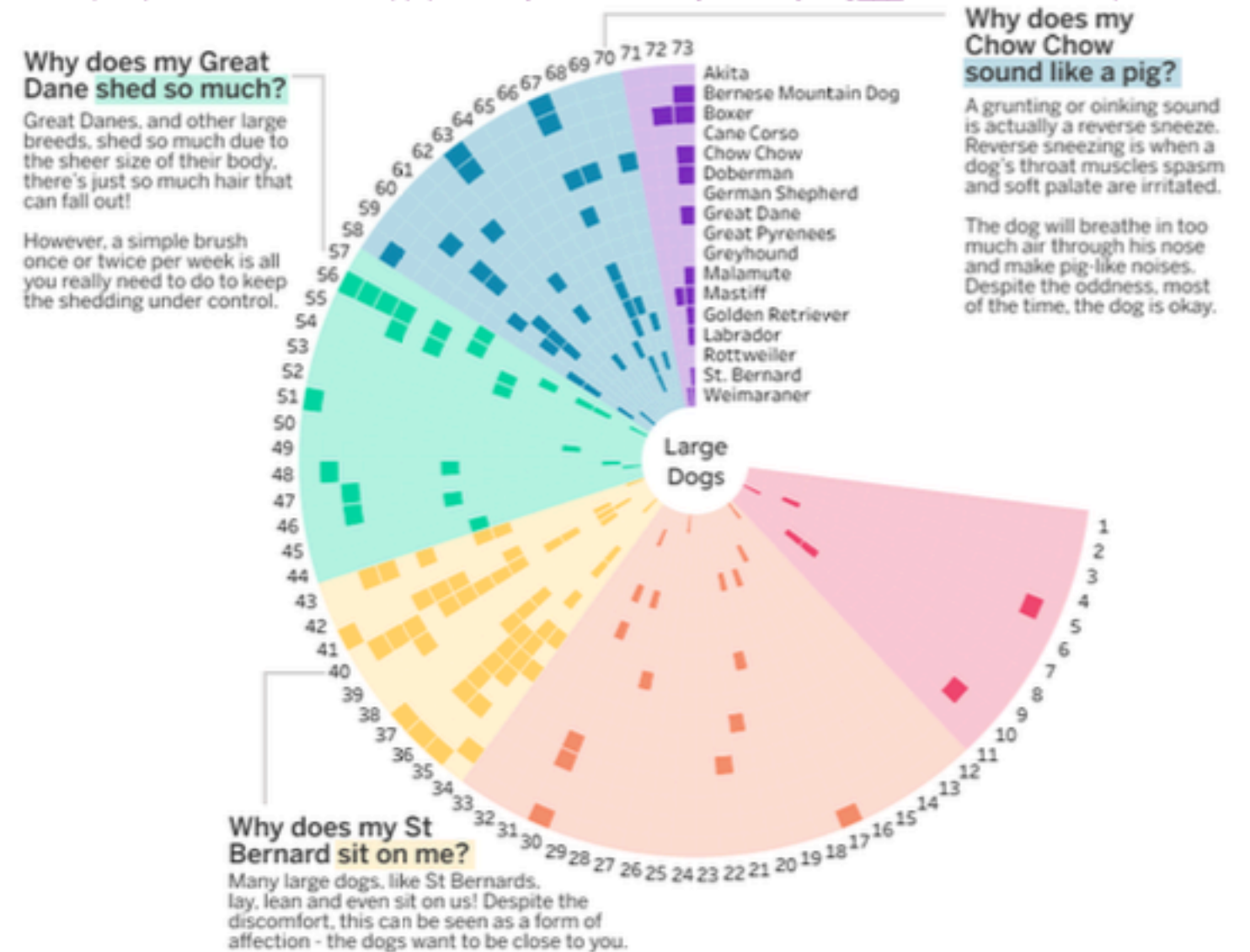


Channels used: what does it mean?

- Does channel size encode attribute?
 - yes: sizes differ
 - according to dog name (alphabetical order)
 - no: size differences not meaningful
 - just emerges from choice of layout, radial vs rectilinear
 - not a "real" attribute
- Can we use size channel to encode another attribute?
 - no!
 - it's "taken" already, would change meaning

Q5 Marks & Channels: Why Does My Dog [4 pts]

Credit: https://public.tableau.com/app/profile/wjsutton/viz/WhyDoesMyDog_IronQuest/DesktopVersion



https://public.tableau.com/app/profile/wjsutton/viz/WhyDoesMyDog_IronQuest/DesktopVersion

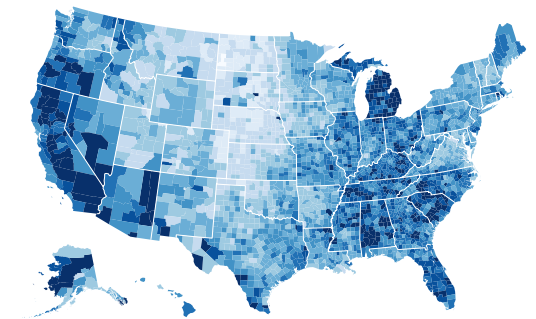
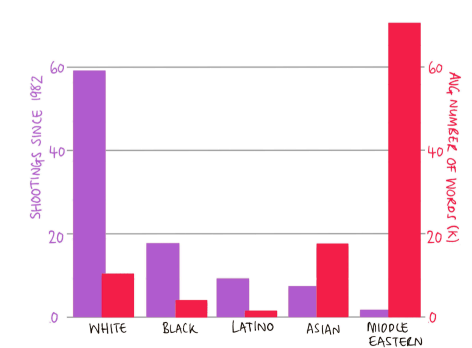
Channel availability

- channels as constraints
 - when does using one channel constrain another channel?
- Channel Availability Model
 - Encoded: which channels directly used to encode attributes?
 - clear meaning
 - multiple channels can be directly used for redundant encoding
 - Unavailable: which channels unavailable / precluded / taken?
 - general dependencies between channels
 - specifics of idiom/algorithm design
 - Free: which channels free to encode another attribute?
 - without changing usability of existing encoding

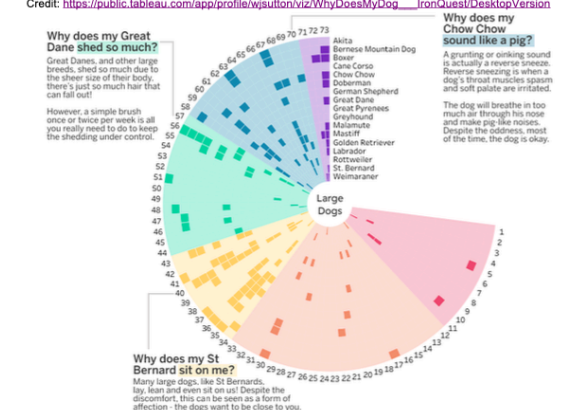
General dependencies: Position

- need fine-grained ability to specify for adequate descriptive power
 - rectilinear (horizontal and/or vertical)
 - high precision
 - depth (3D position): very low precision
 - radial (angular position and/or radial distance)
 - lower precision
 - general dependencies for unavailability?
 - cannot use both rectilinear and radial simultaneously
 - in same layer, using one type precludes other
 - but horizontal doesn't preclude vertical & vice versa

Mass Shootings By Race Of Shooter
NYTimes Coverage Of Mass Shootings By Race Of Shooter

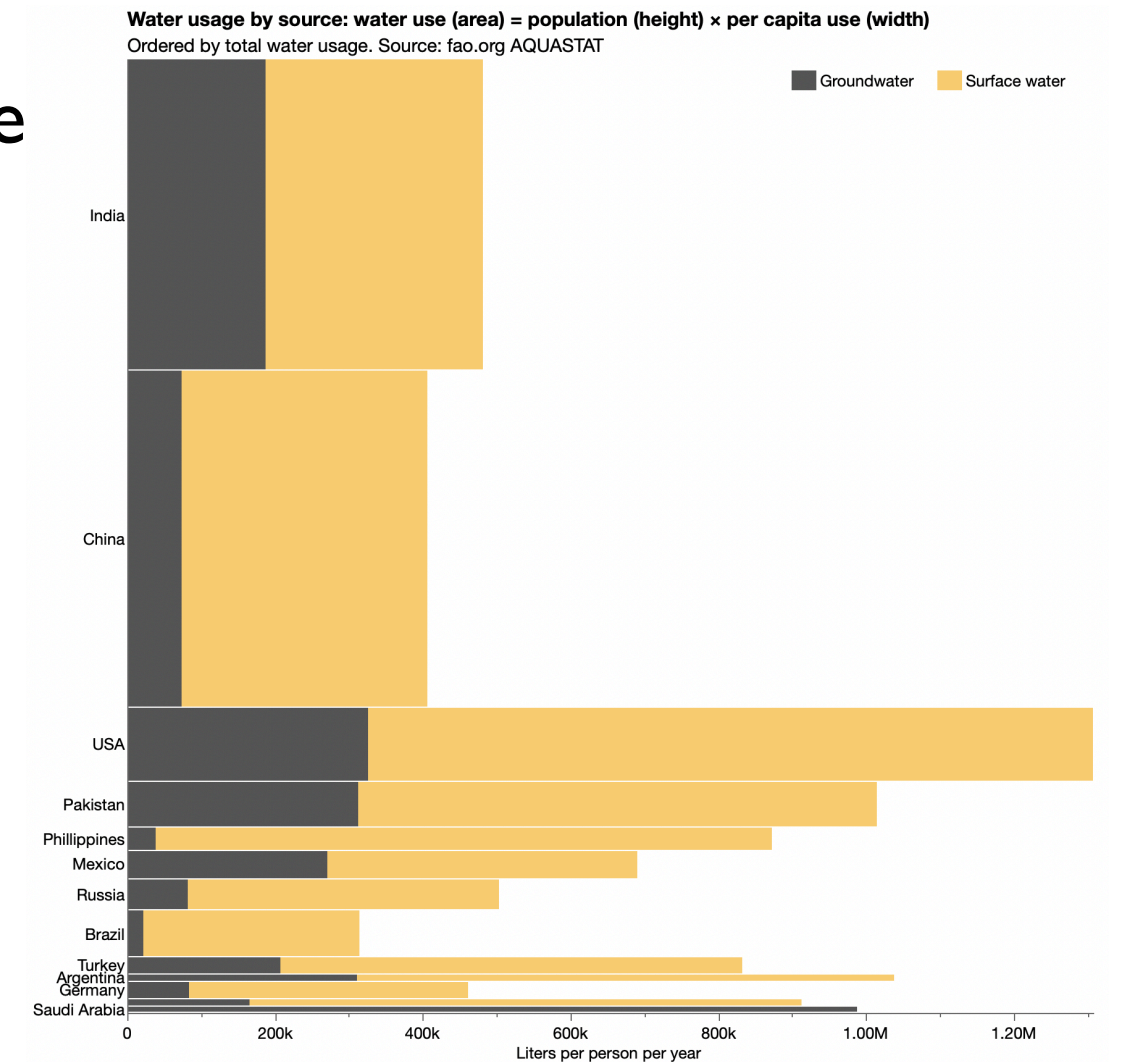


Q5 Marks & Channels: Why Does My Dog [4 pts]
Credit: https://public.tableau.com/app/profile/wjauton/viz/WhyDoesMyDog_IronQuest/DesktopVersion



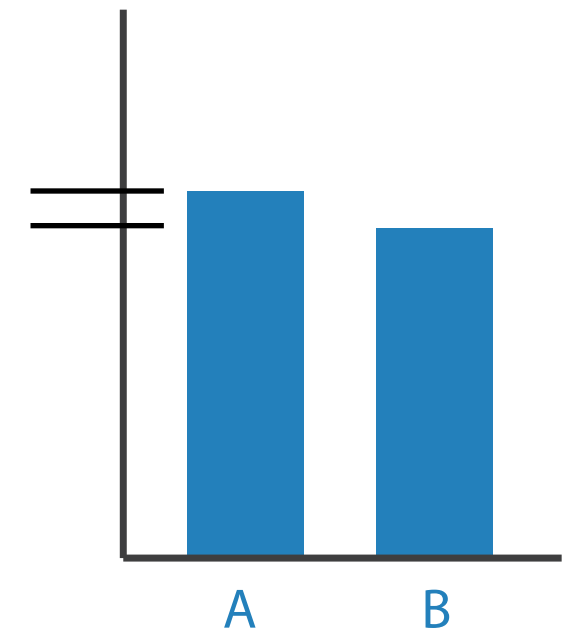
General dependencies: Size

- 1D (length) \ll 2D (area) \ll 3D (volume)
- dependencies for unavailability?
 - larger dimension subsumes smaller ones
 - encode with area channel means length channel unavailable
 - volume means area & length unavailable
 - but not vice versa: can augment from length to area
 - add second attribute for 1D size coding in other direction

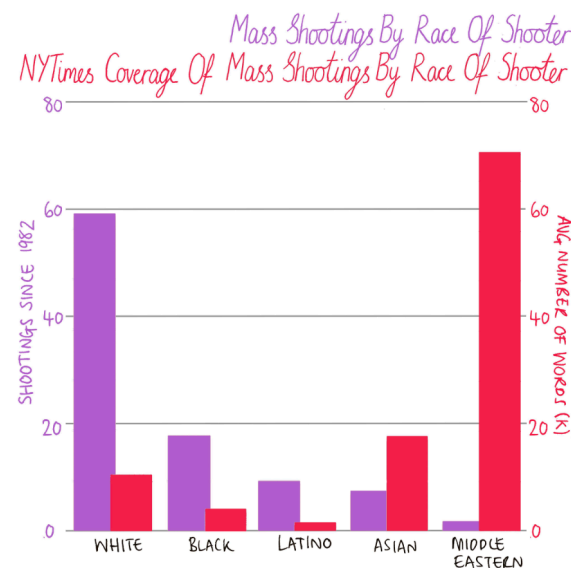


General dependencies: Position vs length

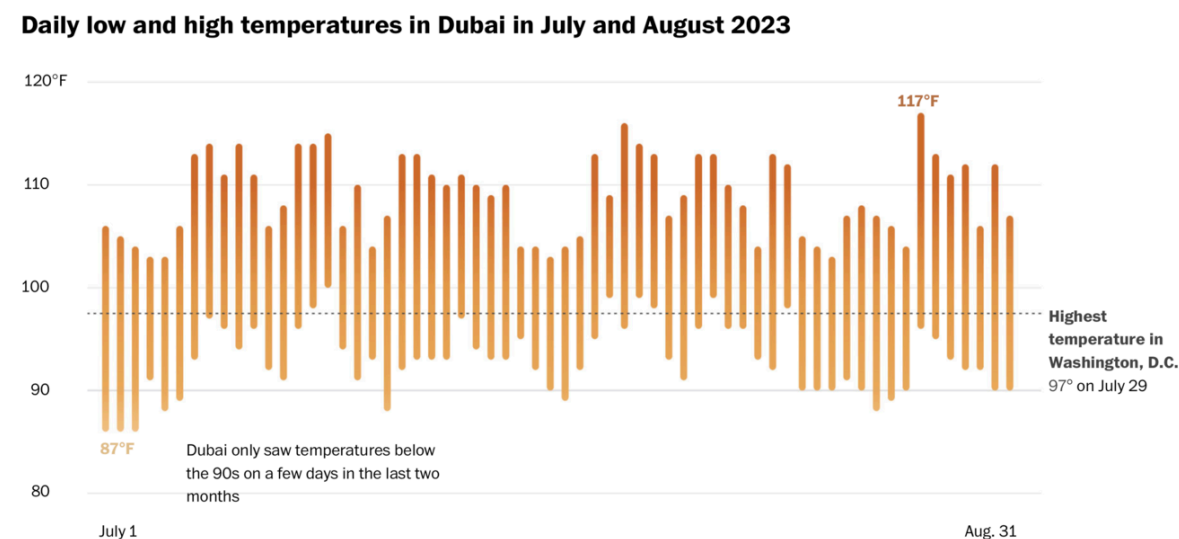
- alignment
 - position (horizontal and/or vertical) is usually shortcut for "aligned position", highest precision channel of all
 - reference frame of explicit axis
 - implicit boundaries of view / window / region
- general dependencies: position (ID) vs length (ID size)?
 - for line marks, position encoded implies length encoded
 - but not vice versa: can have length without position



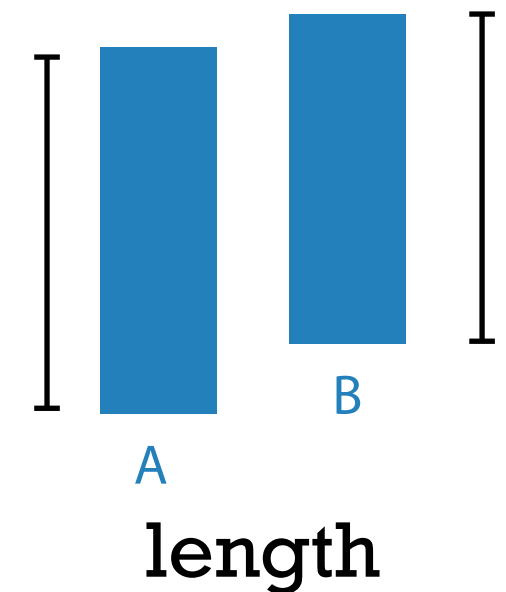
position along aligned scale



<https://twitter.com/MonaChalabi/status/1158779046693679106>



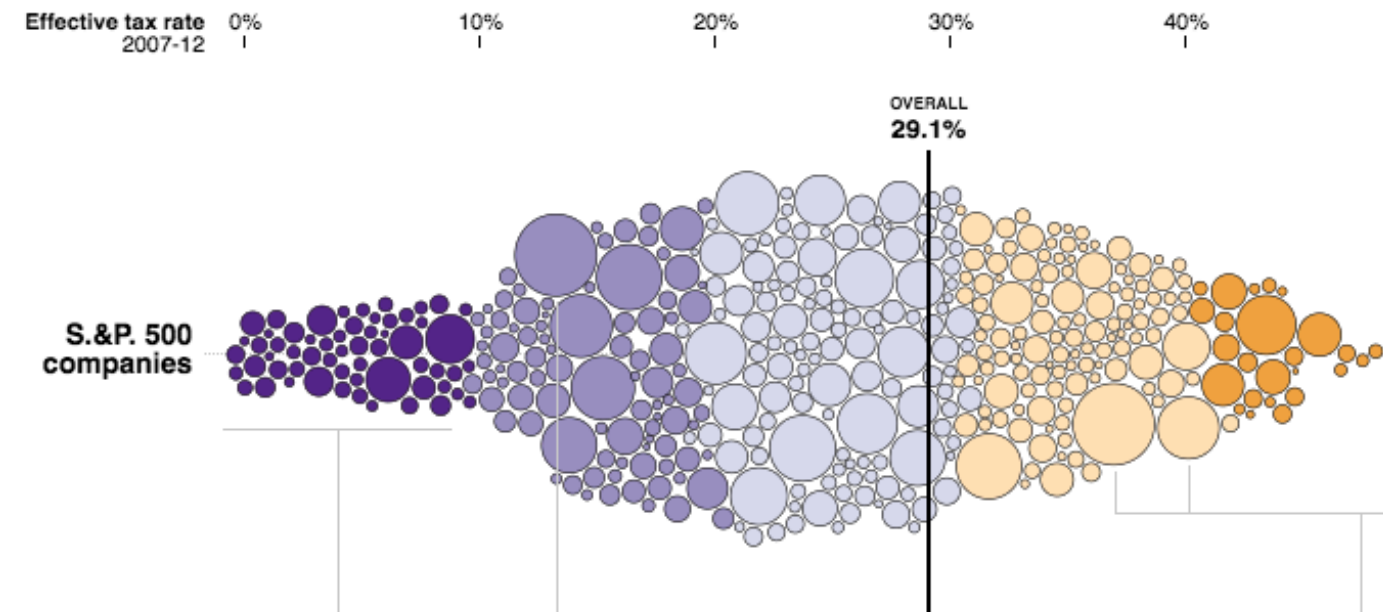
<https://www.washingtonpost.com/world/2023/09/10/dubai-heat-staying-cool>



length

Channel availability analysis: Circle packings

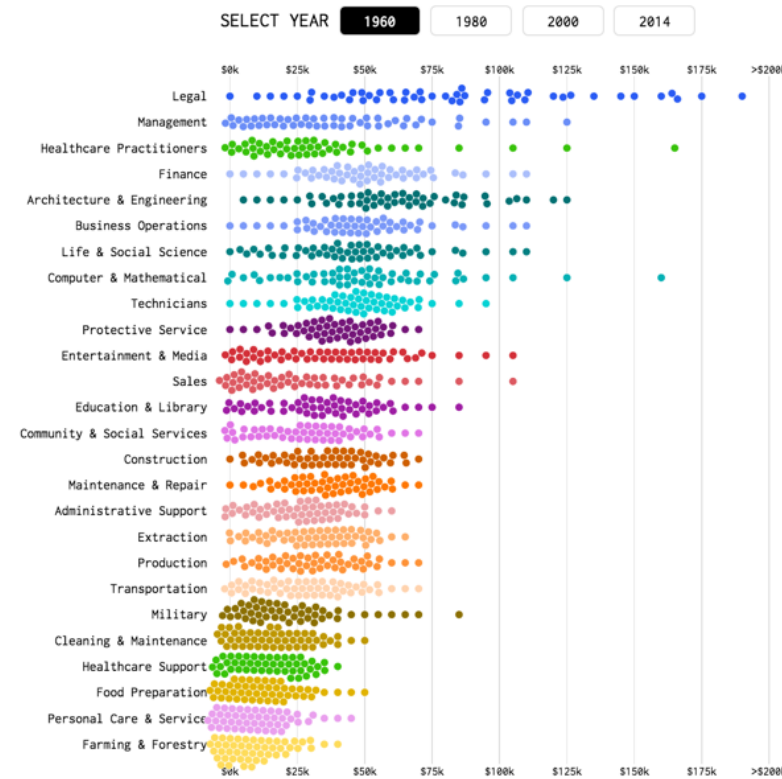
- customized circle packing
 - Encoded channels
 - horizontal position: encodes tax rate
 - color: rate, redundant with horizontal position
 - size: market cap
 - Unavailable channels
 - vertical position: used by algorithm to avoid overlap
 - radial positions (angular, distance): precluded by horizontal position channel use
 - Free channels
 - motion
 - unclear
 - shape? orientation?



<https://archive.nytimes.com/www.nytimes.com/interactive/2013/05/25/sunday-review/corporate-taxes.html>

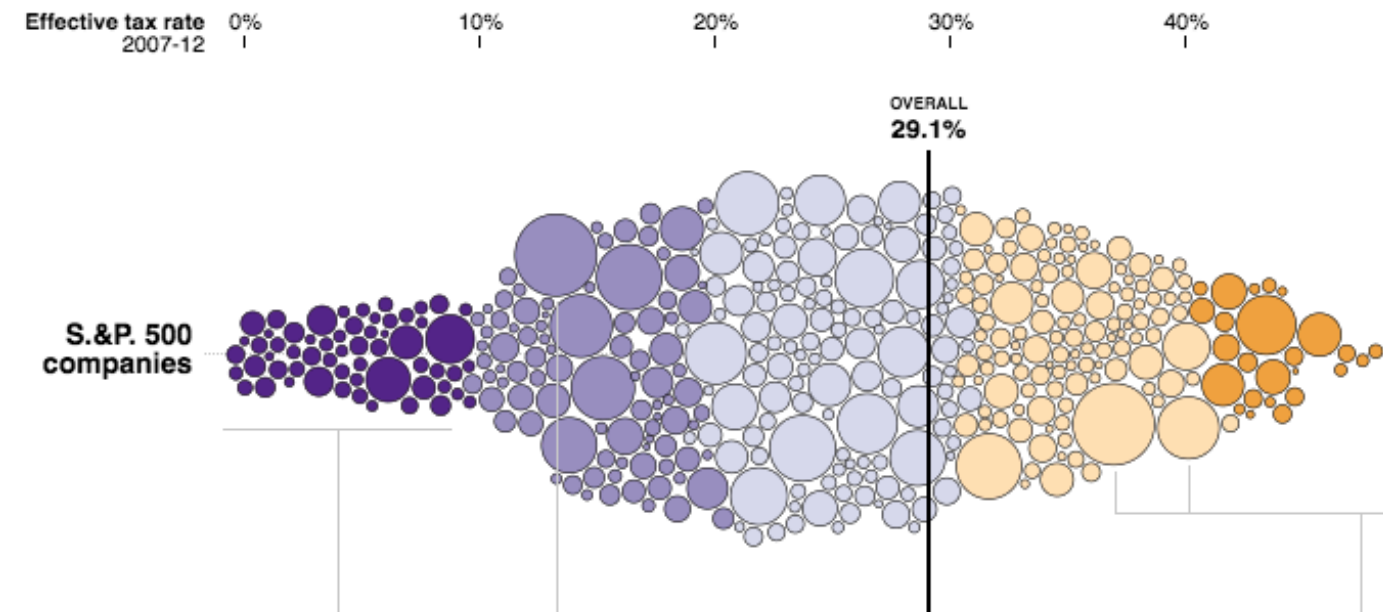
Channel availability analysis: Circle packings

- customized circle packings are special case
 - beeswarm plots



<https://flowingdata.com/2016/06/28/distributions-of-annual-income/>

- general circle packing
 - algorithmic constraint: no overlaps, minimal gaps
 - Unavailable: position used by algorithm

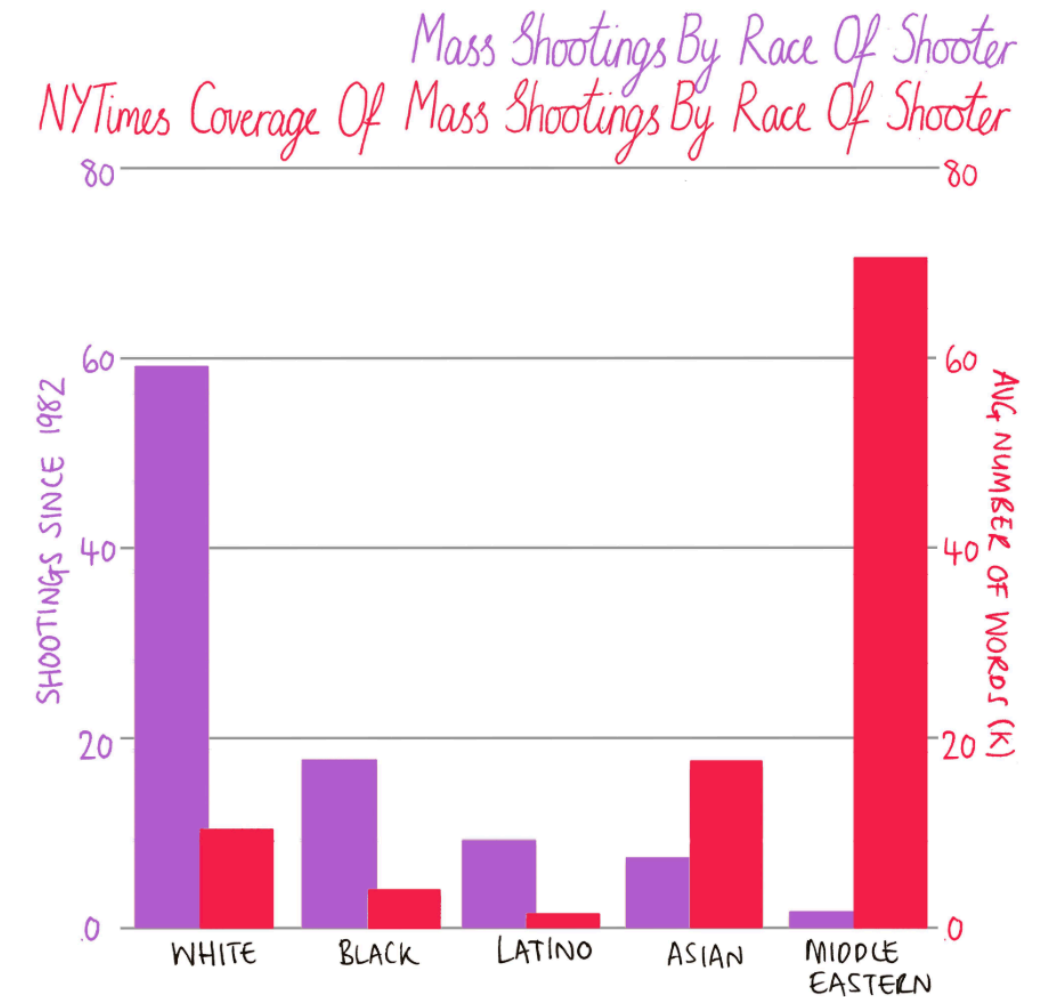


<https://archive.nytimes.com/www.nytimes.com/interactive/2013/05/25/sunday-review/corporate-taxes.html>



Channel availability analysis: Grouped bar charts

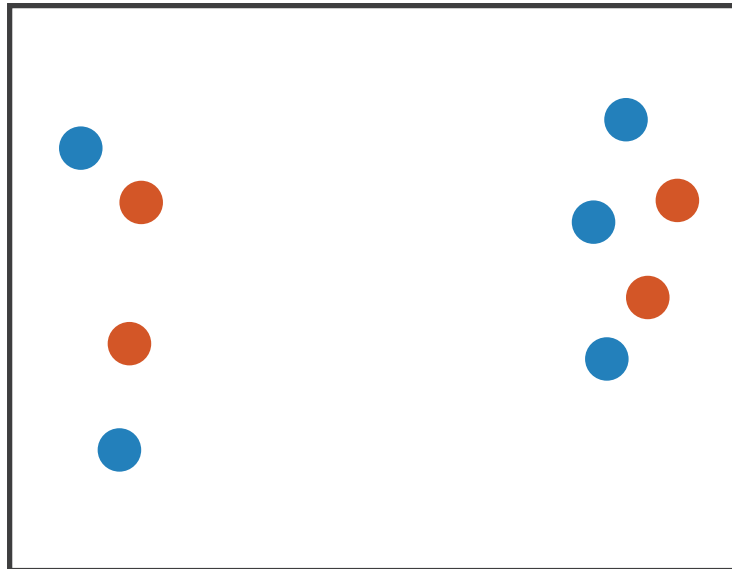
- Encoded
 - vertical position encodes quantitative attributes
 - shootings & coverage counts
 - length (ID size) redundantly encodes same thing
 - color encodes categorical attrib (shooting vs coverage)
 - horizontal position
 - low-level (within group) encodes same as color
 - high-level (across groups) encodes race (shooting & coverage)
- Unavailable
 - any other position channel (radial) precluded
- Free
 - motion, shape, ...



<https://twitter.com/MonaChalabi/status/1158779046693679106?s=20>

Separability vs integrality: Existing framework

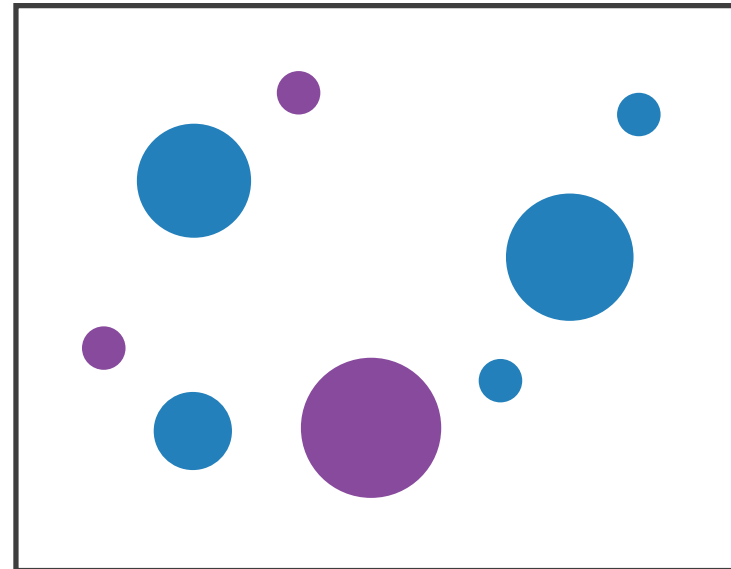
Position
+ Hue (Color)



Fully separable

2 groups each

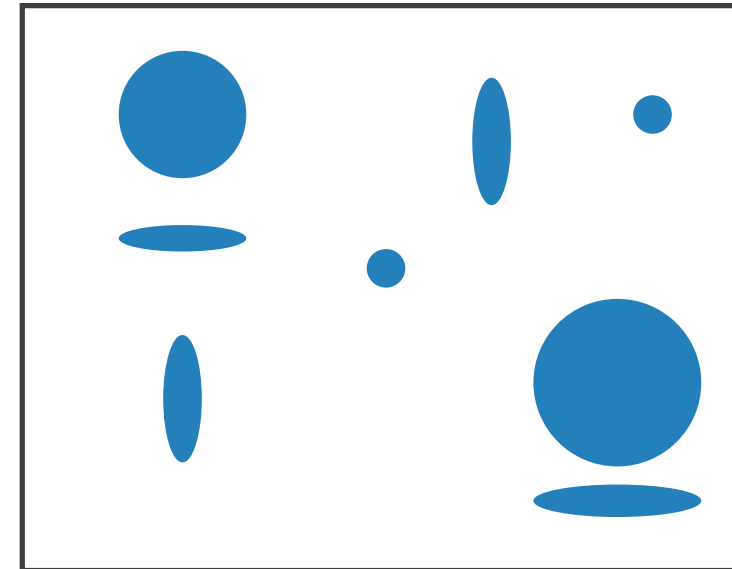
Size
+ Hue (Color)



Some interference

2 groups each

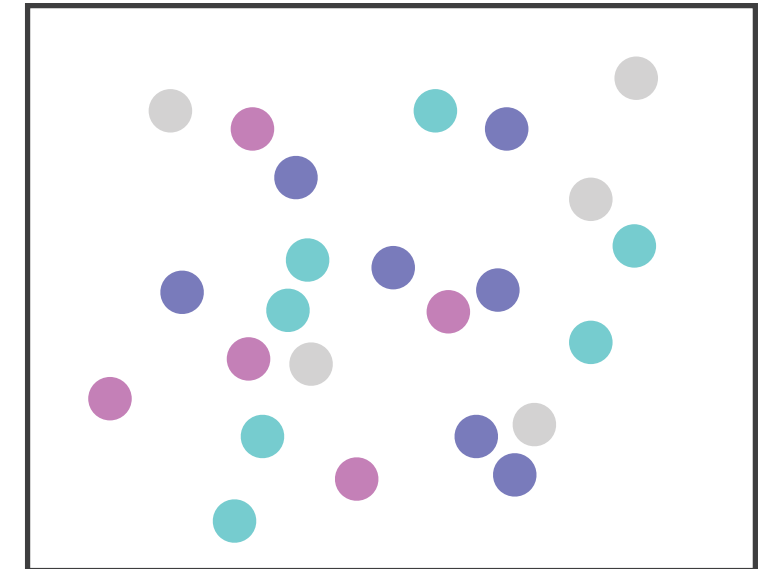
Width
+ Height



Some/significant
interference

3 groups total:
integral area

Red
+ Green



Major interference

4 groups total:
integral hue

Channel availability model augments, not replaces

Channel availability model

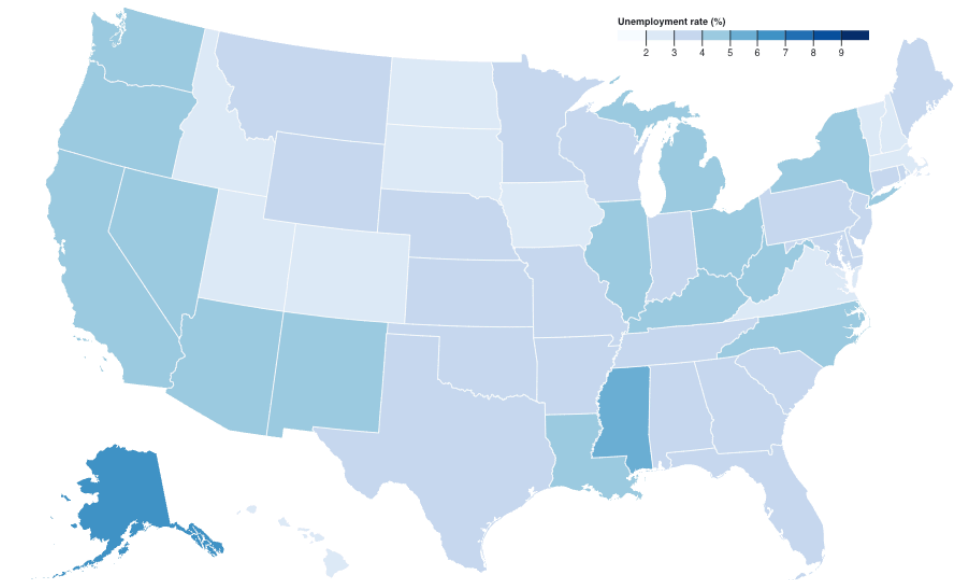
- Channel availability
 - Encoded
 - clear meaning
 - Unavailable: which channels unavailable / taken?
 - general dependencies
 - idiom/algorithm design
 - Free: which channels free to encode another attribute?
 - without changing usability of existing encoding
- how does this idea relate to mark types?

Area marks: Rethinking

- area marks is a terrible name
 - other marks all have graphical area too
 - allowing us to encode with color
 - there's also an "area" channel, which is confusingly different
 - area is not the only channel in play with these marks!

Area marks

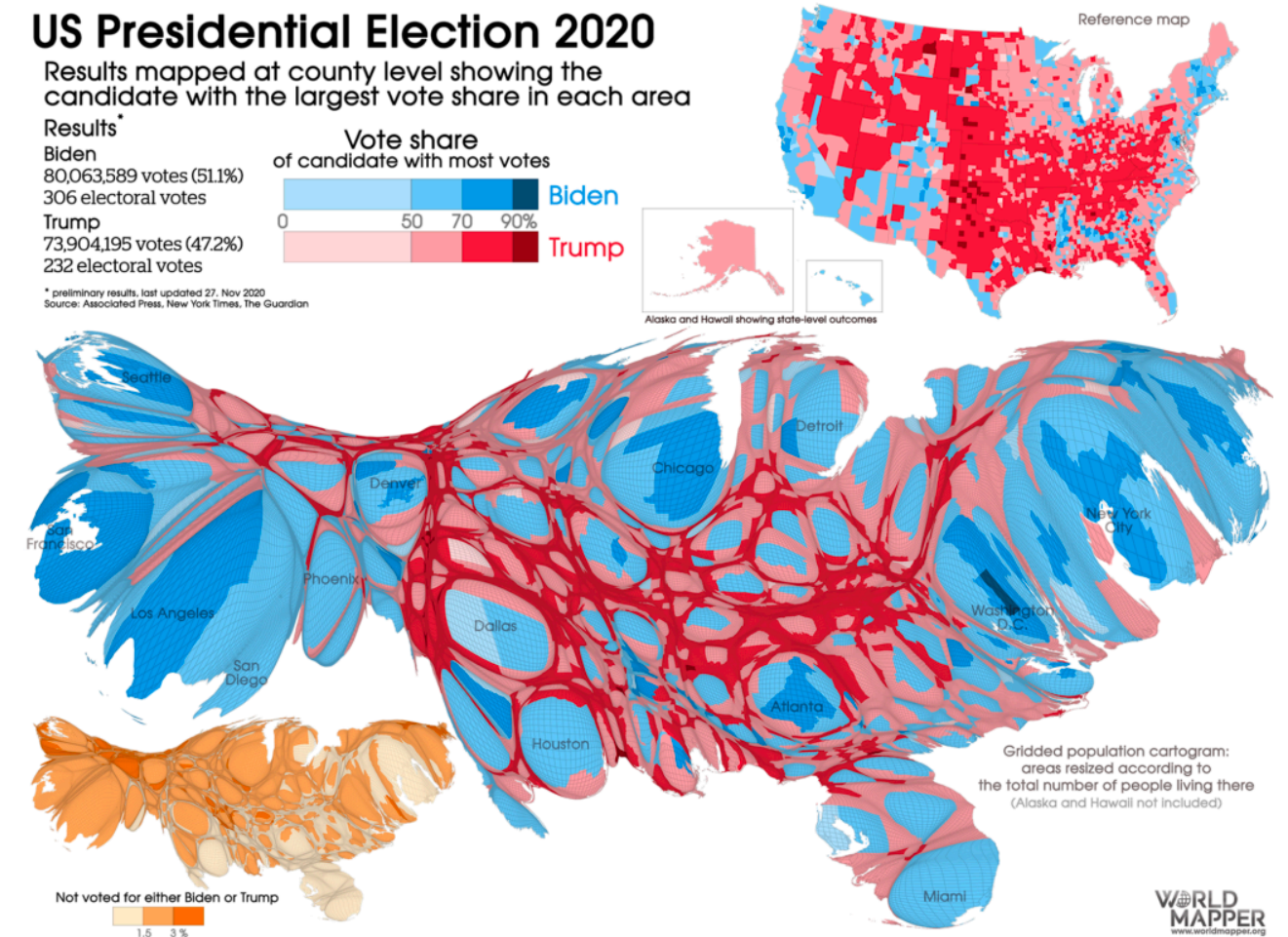
- obvious example: choropleth maps
- what can we do to California? could we encode additional data?
 - cannot shrink/grow (size channel)
 - cannot translate (position channel)
 - cannot rotate (orientation channel)
 - cannot reshape (shape channel)
 - why not?
 - would lose meaning of that mark: boundary is the data
 - also lose meaning for other occluded marks
- "area" mark is not specific enough
 - AreaPositionOrientationShape mark??? nah...
 - idea: **interlocking**



<https://observablehq.com/@d3/us-state-choropleth>

Interlocking (area) marks

- many channels locked down with interlocking marks
 - boundary encodes meaning
 - cannot change size, shape, position, orientation
 - mark type as a constraint
- but... what about cartograms?
 - cannot change just one mark (California)
 - but could change them all!
- interlocking marks as global constraint:
 - cannot change just one independently
 - but can change all simultaneously!
 - typically with algorithm

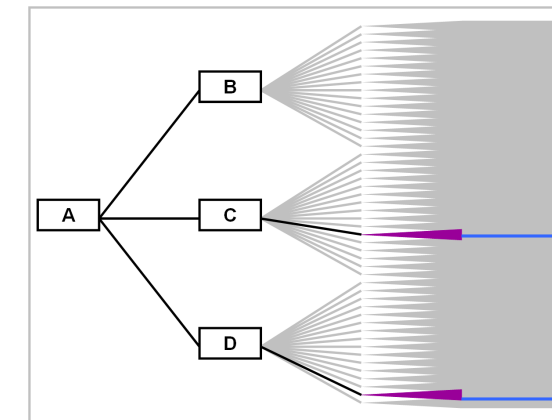
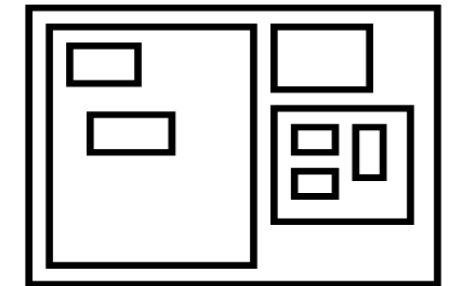
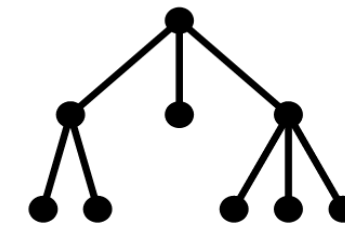
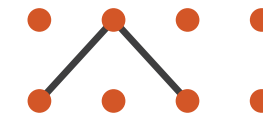


<https://worldmapper.org/us-presidential-election-2020/>

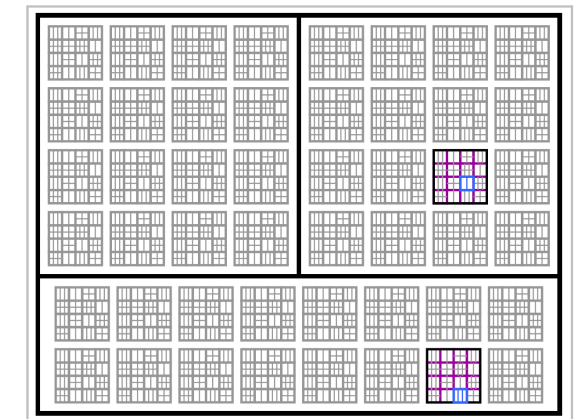
Interlocking marks: Non-spatial

- example with non-spatial data?
- treemaps
 - show hierarchy with containment, not connection
 - encode additional attribute with area/size
- again, cannot change just one mark alone
 - but could recompute layout to change all at once
- combined layout of all marks together carries meaning
 - unlike spatial data mark boundaries
 - individual mark boundaries have no intrinsic meaning

→ Connection → Containment



Node-Link Diagram



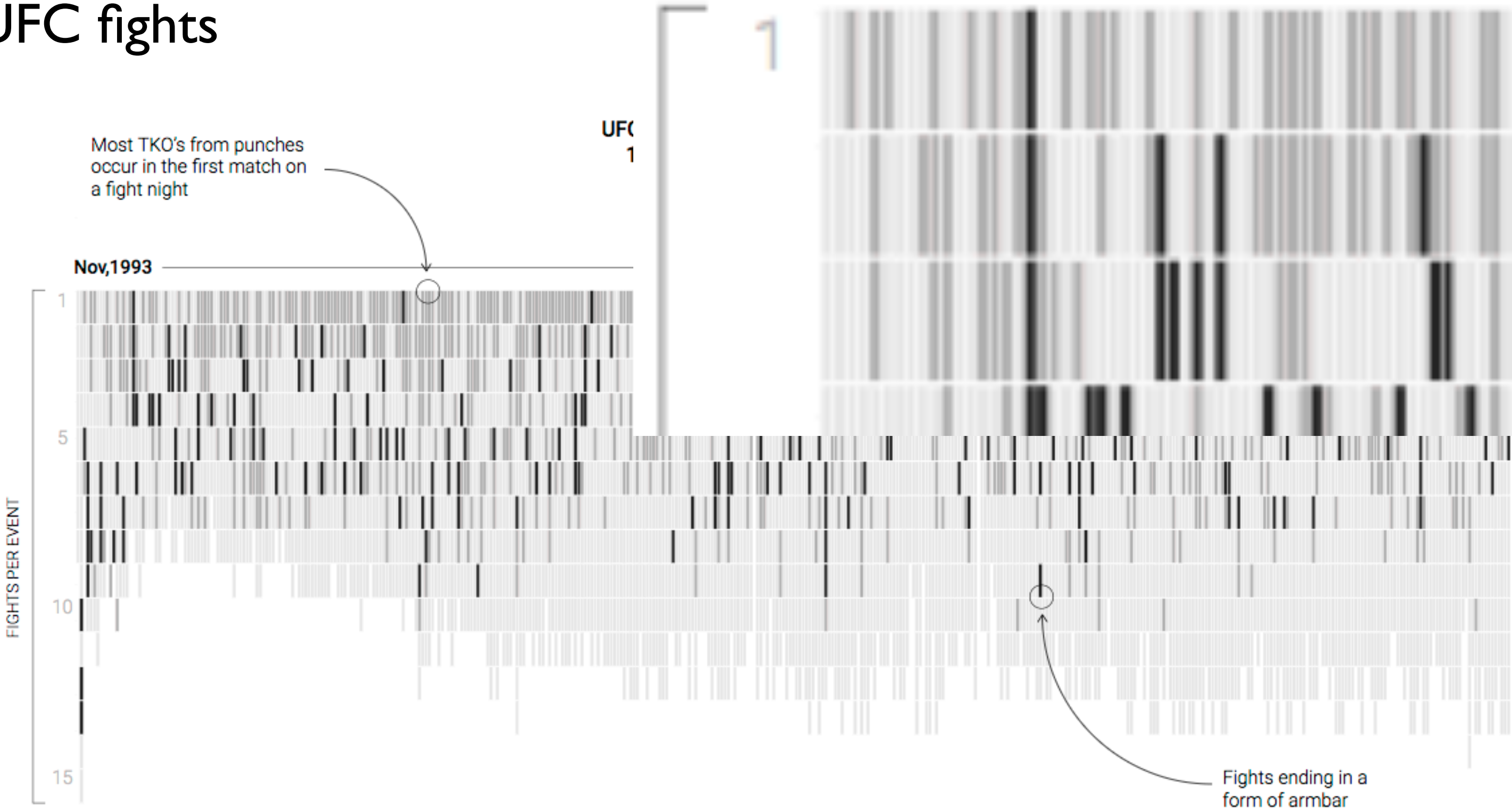
Treemap

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

Quiz: Name that mark

- C: UFC fights

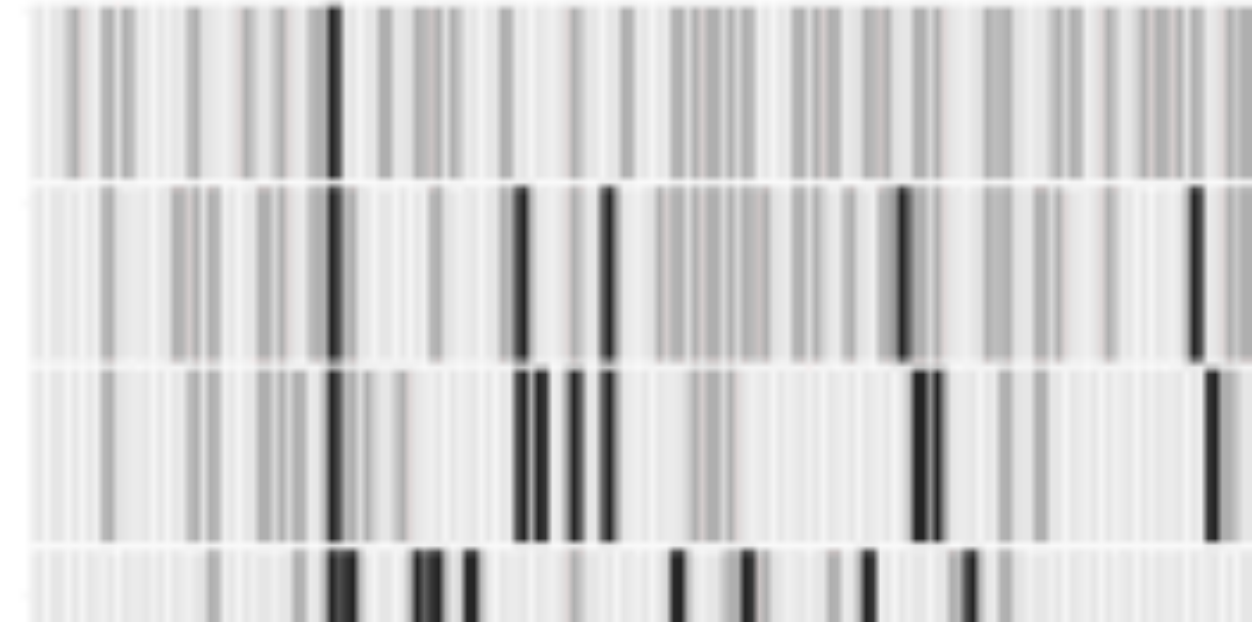
Nov, 1993



Analyzing marks

- what type of mark?
 - line?
 - no, not length coded
 - point mark with rectangular shape?
 - 2020: yes!
 - 2023: no!
 - cannot change position / size / orientation
 - area?
 - 2020: no, area/shape does not convey meaning
 - 2023: yes!
 - fully interlocking
 - position, size, shape, orientation all locked

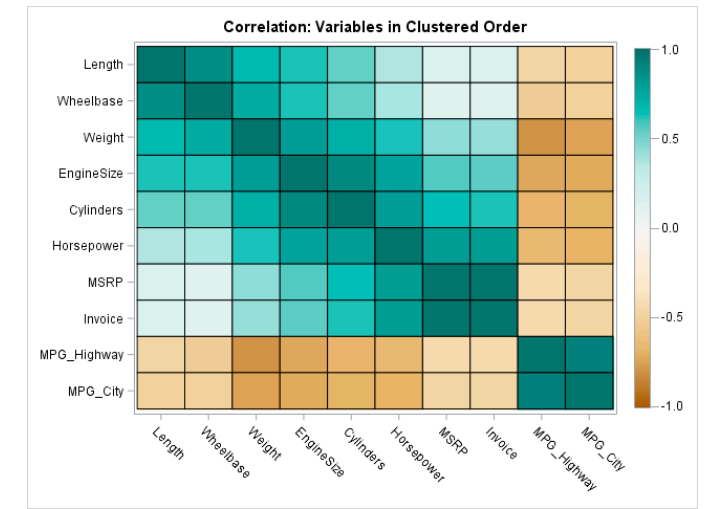
Nov, 1993



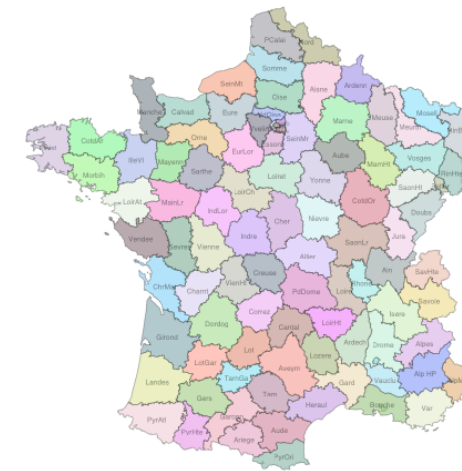
<https://multimedia.scmp.com/infographics/sport/article/3010883/bruce-lee-and-mixed-martial-arts/>

Interlocking marks: Tile heatmaps

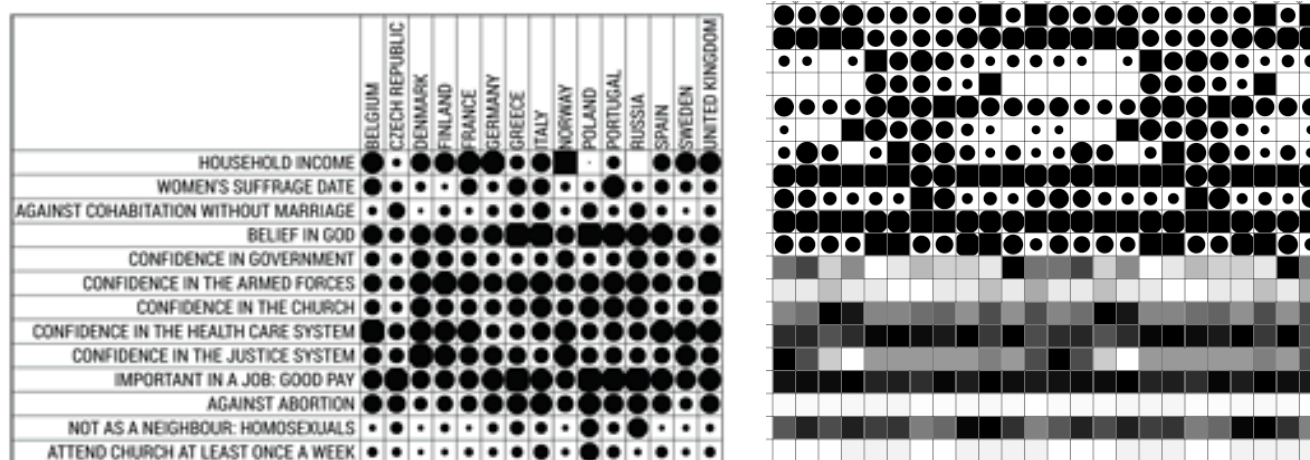
- 2D matrix/grid as index
 - position in use as index
 - size/area & shape & orientation all equal (& locked down)
- simplest possible case of interlocking marks?
 - more regular than choropleths or treemaps
 - but underlying similarities
- full extent of cell used for color coding
 - different from using a point mark within the cell



<https://blogs.sas.com/content/iml/2018/05/02/reorder-variables-correlation-heat-map.html>



Spatially ordered treemaps.
Wood and Dykes.
IEEE TVCG (Proc. InfoVis) 14(6):1348-1355, 2008.



Bertier, Charles Perin, Pierre Dragicevic, Jean-Daniel Fekete (2014).
[Revisiting Bertin's Matrices: New Interactions for Crafting Tabular Visualizations](#). TVCG, VIS' 2014.

Interlocking marks: Circle packings

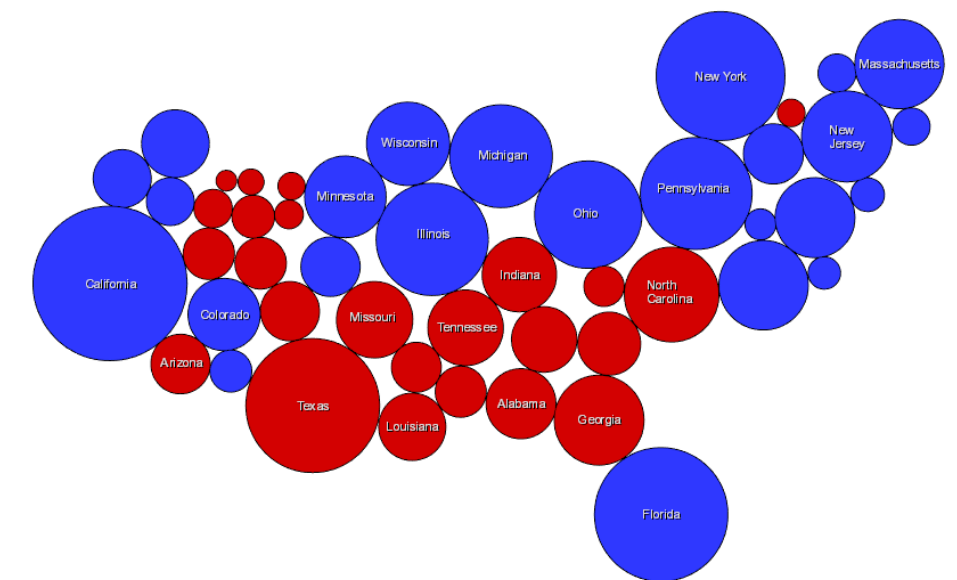
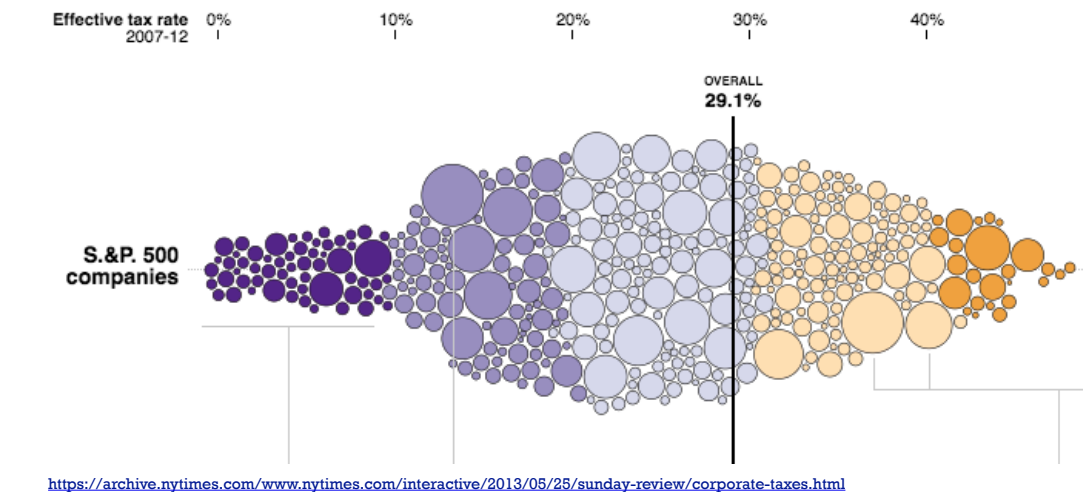
- interlocking marks

- not size-coded point marks (with circle shape)
- shape / position / orientation: equal & unavailable
- more like treemap than scatterplot



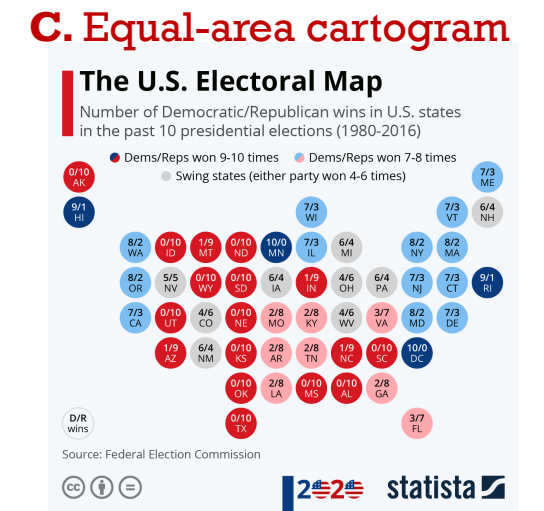
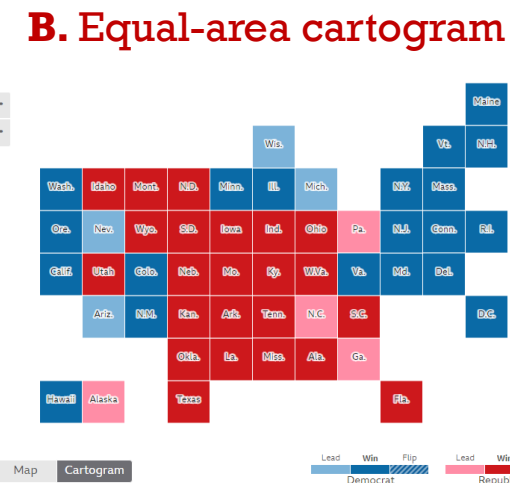
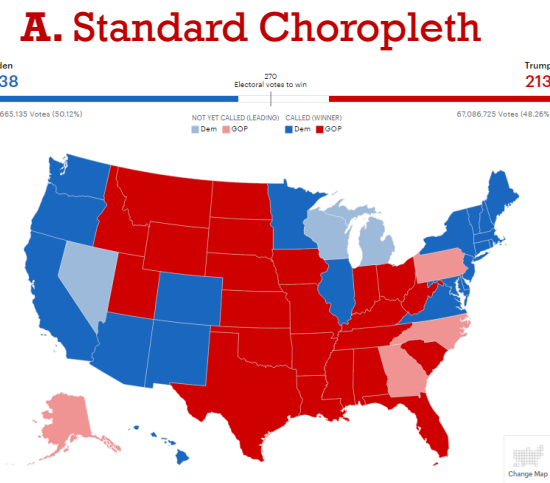
- Dorling cartogram

- can treat as special case of circle packing, with additional constraints based
- maintain relative positioning from geographic location
- throw away shape by regularizing to circles
- add size coding



Interlocking? Election maps roundup

- yes interlocking
 - A: already covered
 - B/C: equal-area alg algorithm simplifies shape



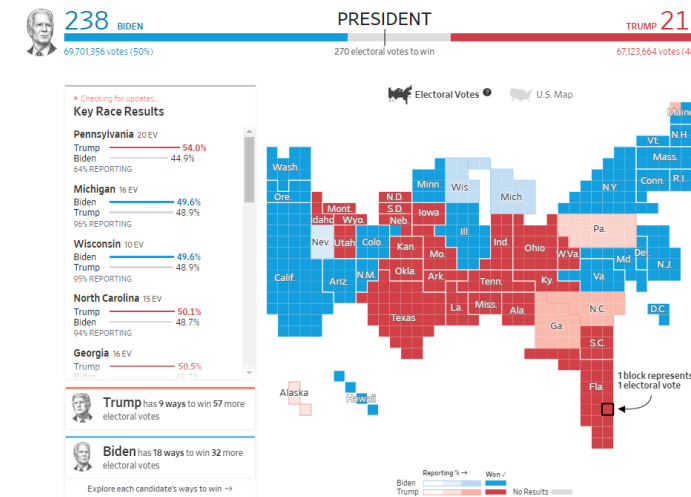
Area not encoding election data

- yes interlocking
 - E/F multi-level
 - top level: interlocking marks
 - bottom level: square units
 - E/F: countability for votes
 - F whitespace: population density

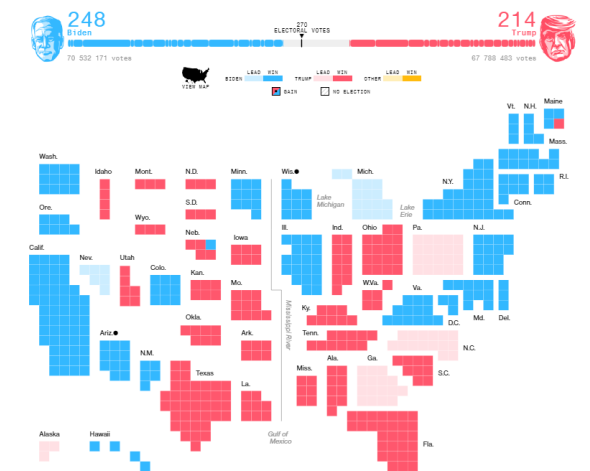
D. Area-proportional continuous cartogram



E. Units assembled into state-like shapes



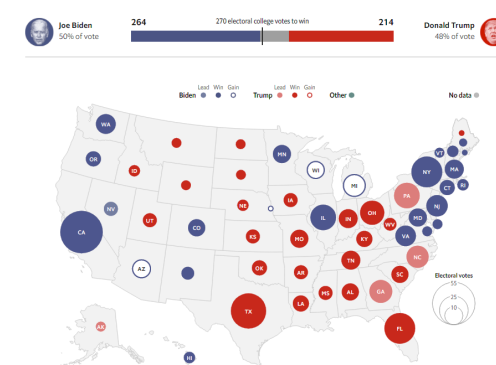
F. Units in state-like shapes spaced apart



Area encoding election data

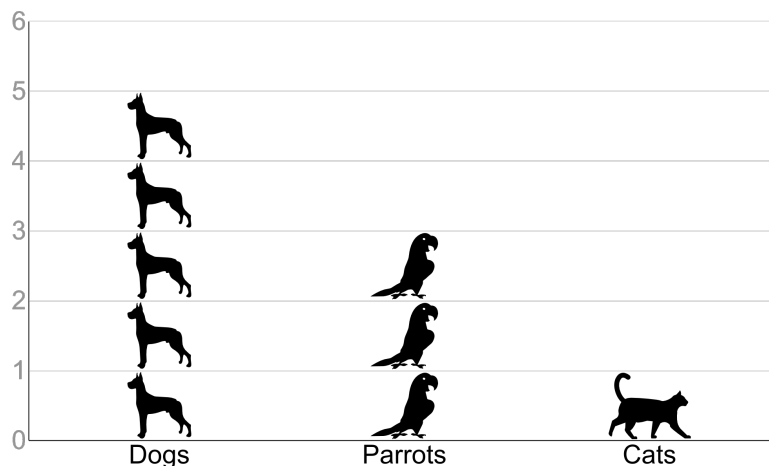
- no, point marks
 - size coded by area

E. Map with bubbles

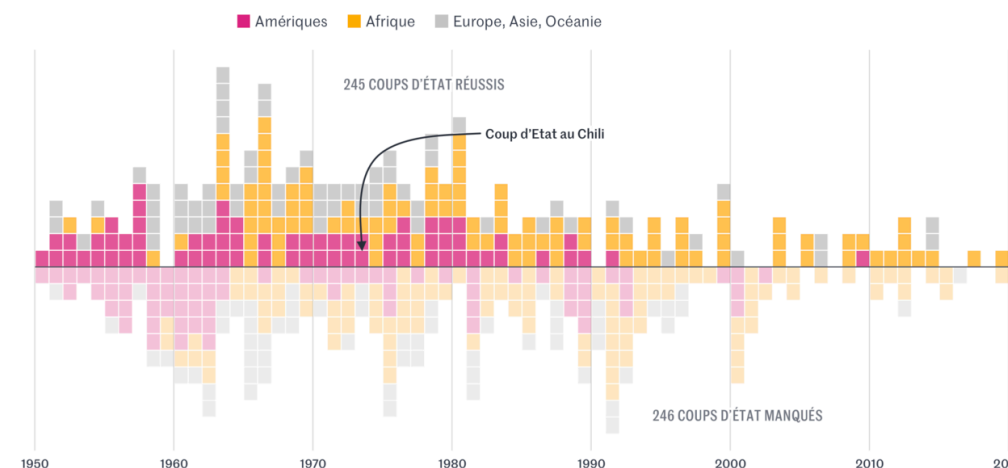


Unit encodings

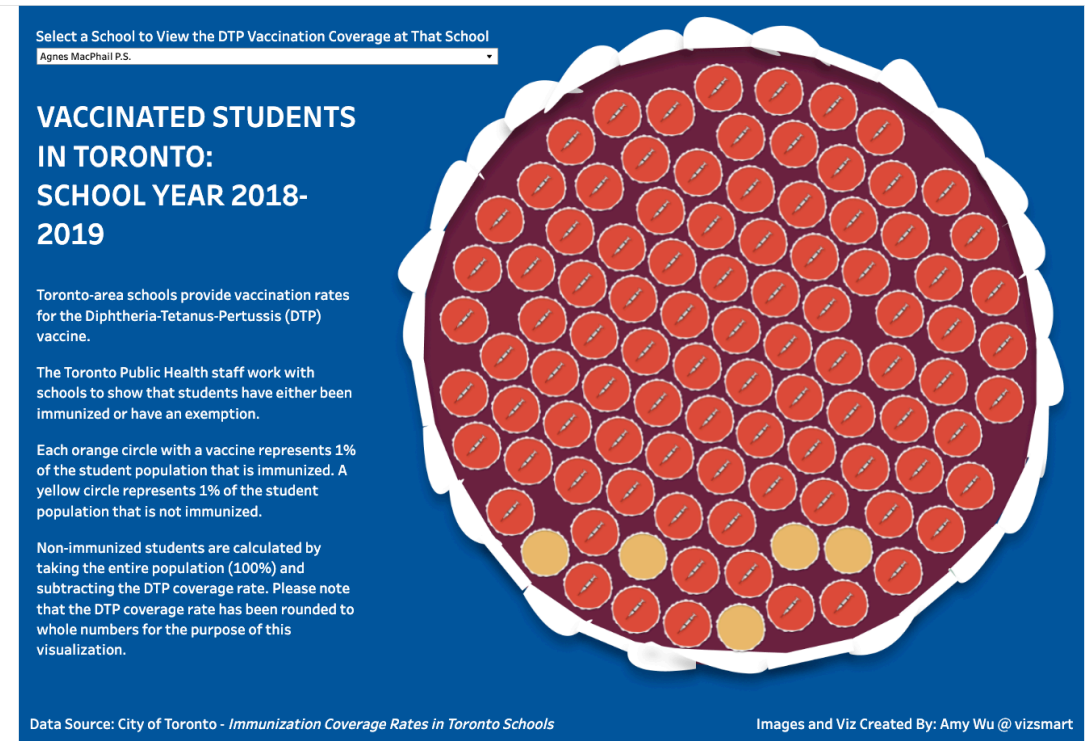
- point marks
 - general case: quantity only!
 - position channel not necessarily in use
- often constrained by idiom
 - multi-level interpretation
 - top level: interlocking mark
 - rectilinear: support counting width & height separately
 - bottom level: unit point marks
 - can be independently color-coded (or interactively highlighted)



<http://steveharoz.com/research/isotype/>



https://www.lemonde.fr/les-decodeurs/article/2023/09/11/depuis-1950-pres-de-cinq-cents-coups-d-etat-tentes-ou-reussis-surtout-en-amerique-latine-et-en-afrique_6188906_4355770.html



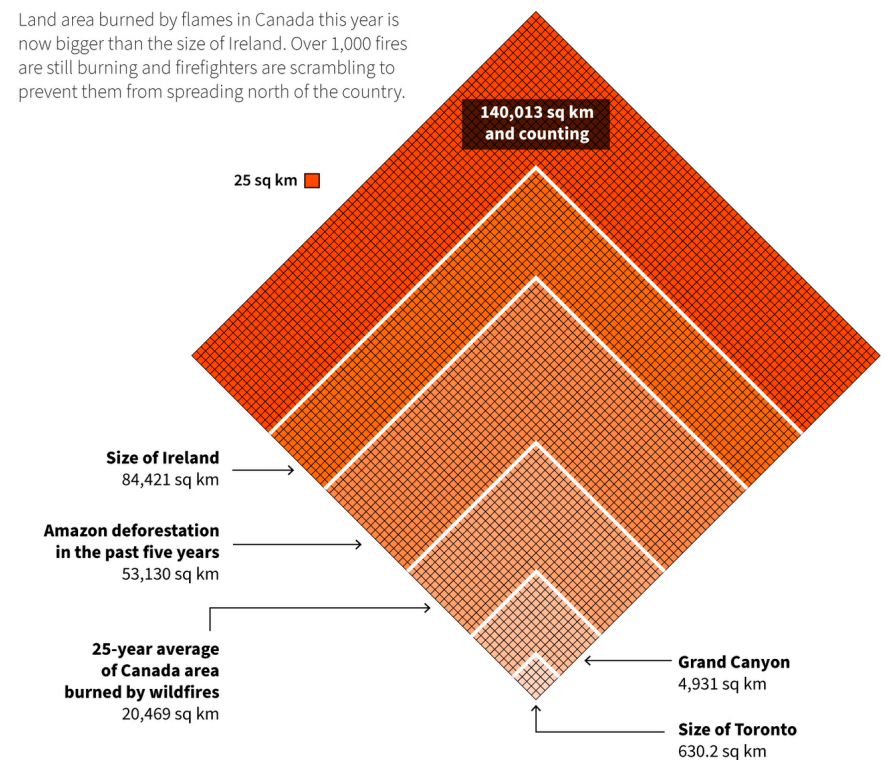
Data Source: City of Toronto - Immunization Coverage Rates in Toronto Schools Images and Viz Created By: Amy Wu @vizsmart

<https://www.informationisbeautifulawards.com/showcase/6163-vaccinated-students-in-toronto-school-year-2018-2019>

https://public.tableau.com/app/profile/amy.r.wu/viz/DTP_16715915193810/DTPDashboard

Wildfires in Canada burned this much land so far this year

Land area burned by flames in Canada this year is now bigger than the size of Ireland. Over 1,000 fires are still burning and firefighters are scrambling to prevent them from spreading north of the country.

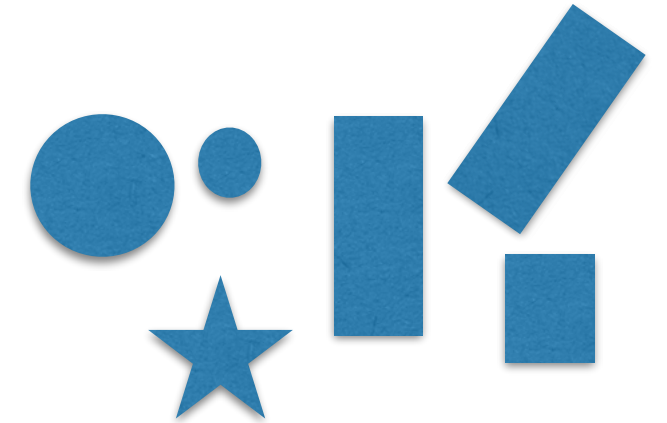


Note: As of Aug. 20, 4 p.m.
Source: Canadian Interagency Forest Fire Centre
Prinz Magtulis | Reuters, Aug. 21, 2023

<https://twitter.com/prinzmagtulis/status/1693684342818574675>

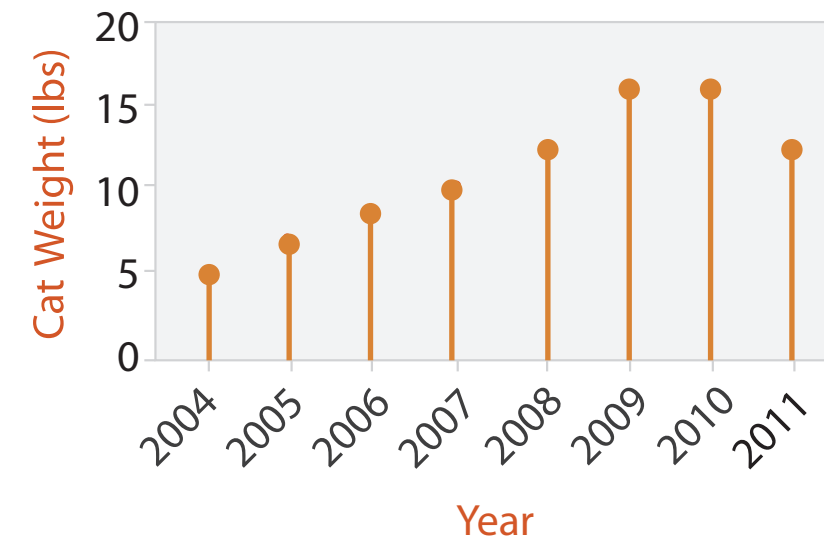
Distinguishing marks through constraints

- highly constrained: interlocking marks
 - many channels unavailable: size, position, shape, orientation
 - proposal: rename from "area" to "interlocking"
- unconstrained: point marks
 - can encode more info with any channel at all!
 - size, position, shape, orientation
 - color, motion, ...
 - does "point" imply circular shape?
 - proposal: is "unconstrained" a better / more evocative name?
- so... what about line marks?



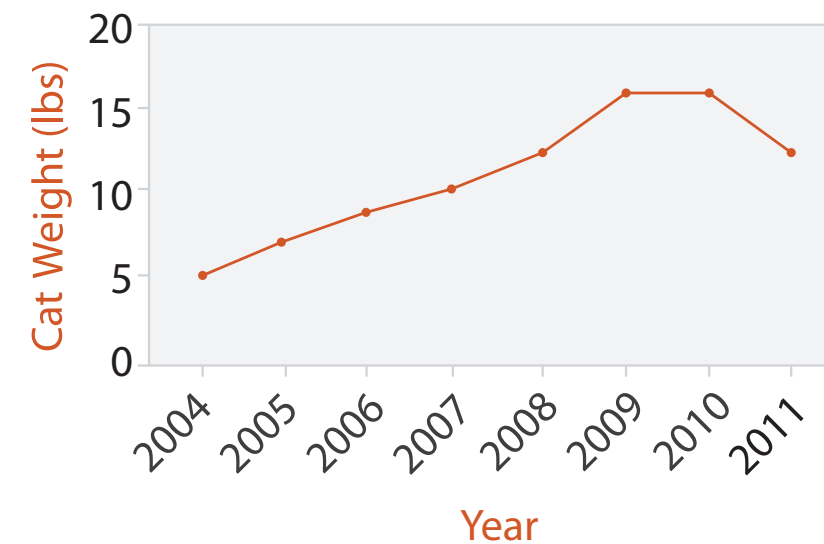
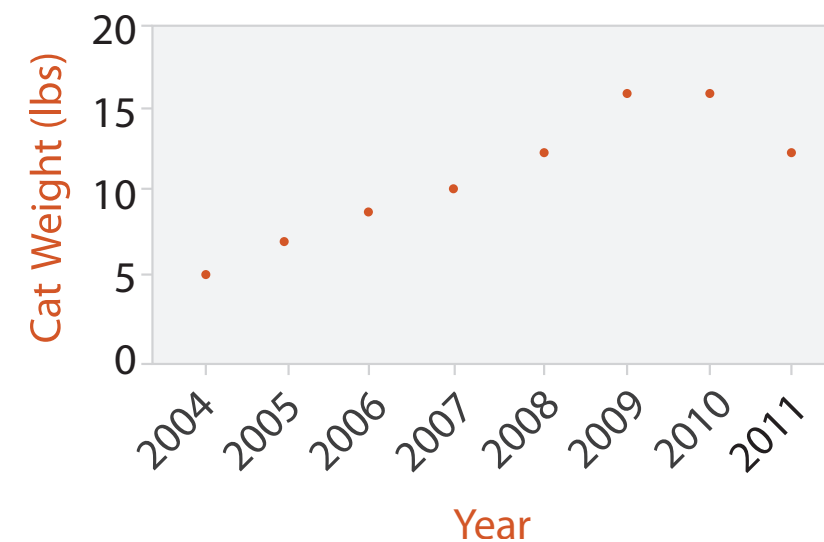
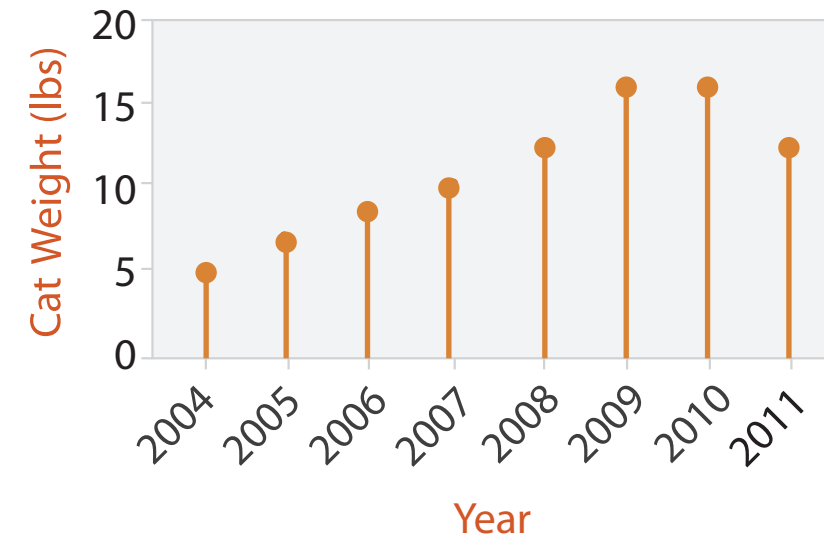
Line marks: Also confusing

- curved lines are not line marks!
 - model: one mark represents one item of data
 - line mark: express single quantitative attribute for one item with length
- also misnamed?
 - proposal: call them "segment" marks instead!



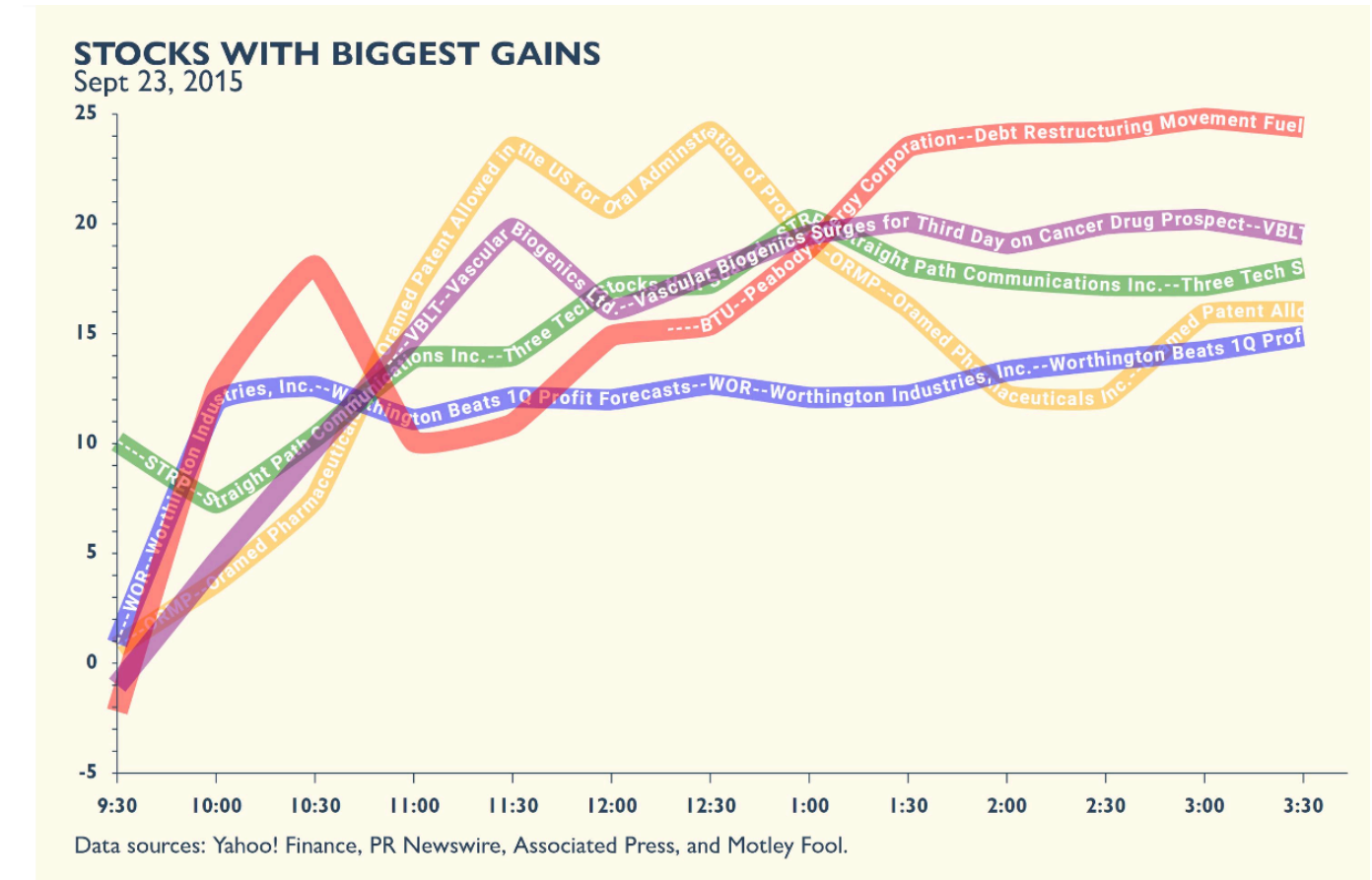
Line marks: Also confusing

- do line charts use line marks?
 - no
 - at least, not exactly like bar charts do
 - connection line segments between points
 - trend task: emphasize relationships between items
 - line chart encodes many items, not just one
 - with many piecewise-linear segments
 - or smoothed into curve



Line marks: Also confusing

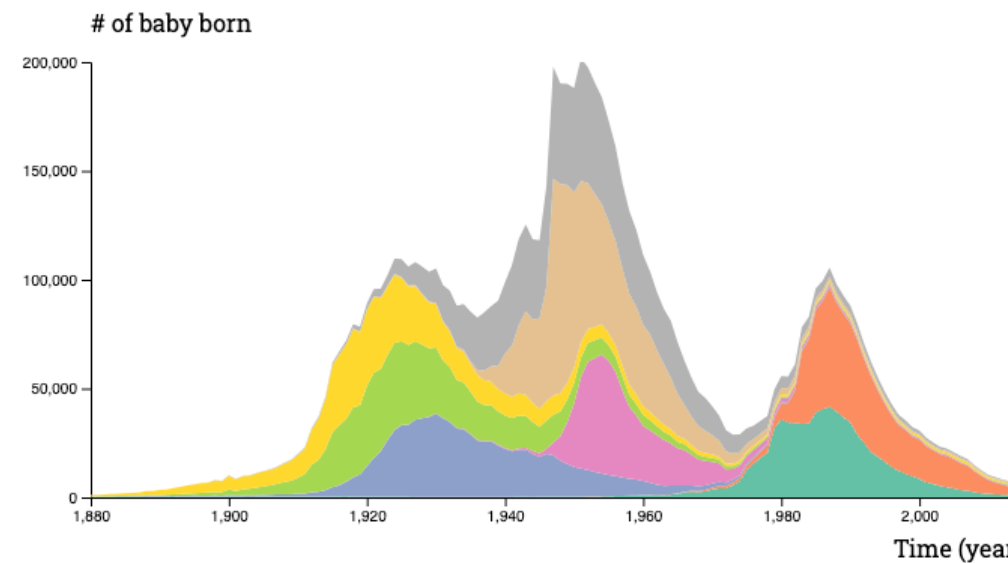
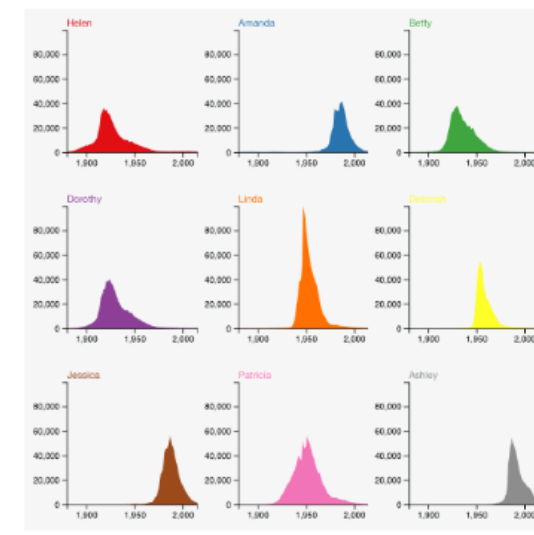
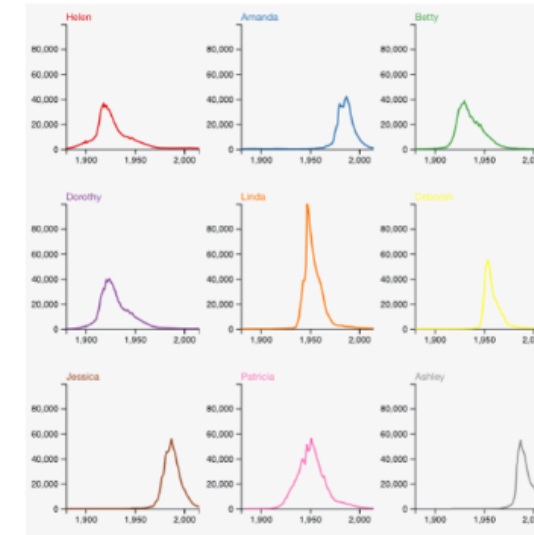
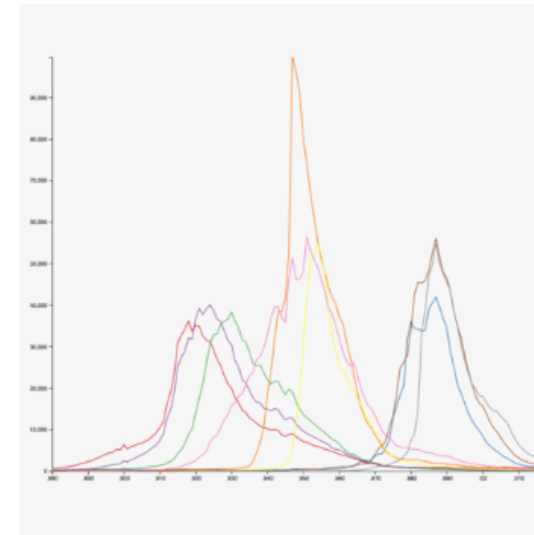
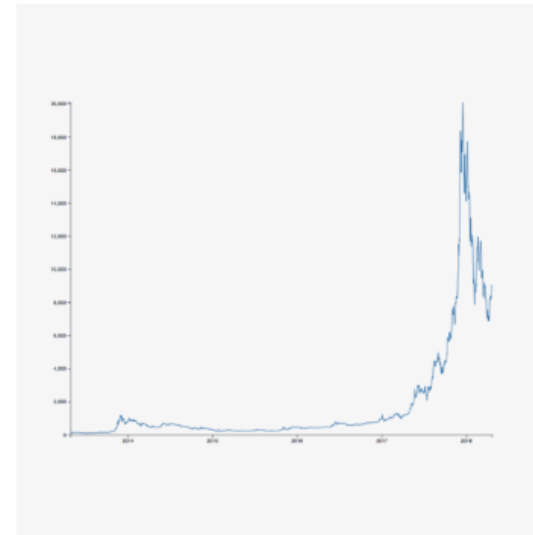
- when does region inside curved "line" boundary act like an area mark?
 - can encode information within its boundary
 - color, even text
 - only if it's wide enough?
 - or always!?
- what about the region below it?...



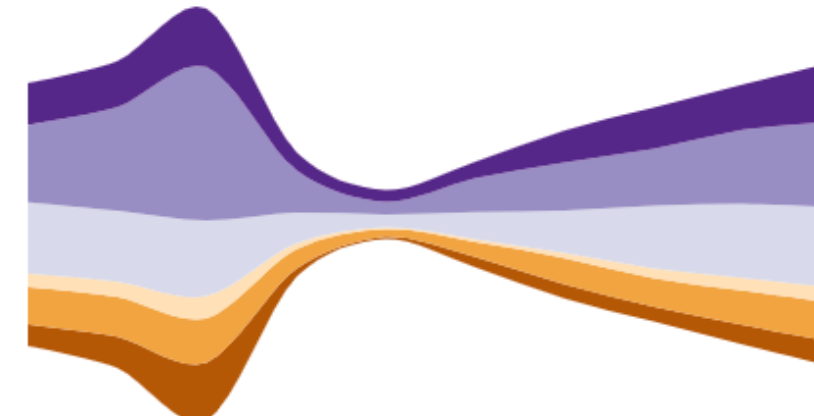
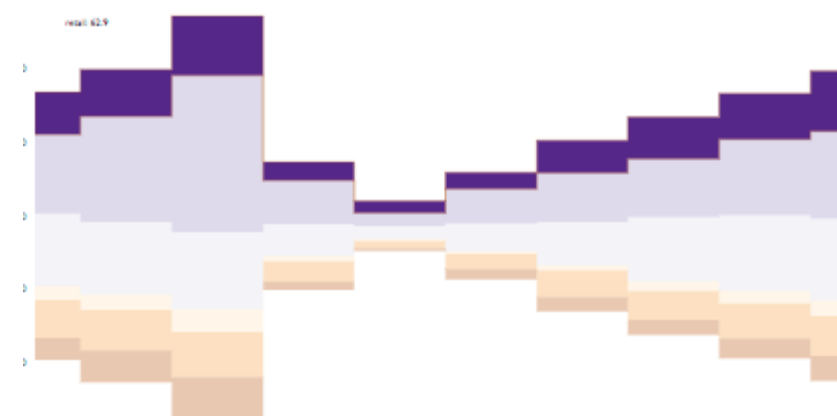
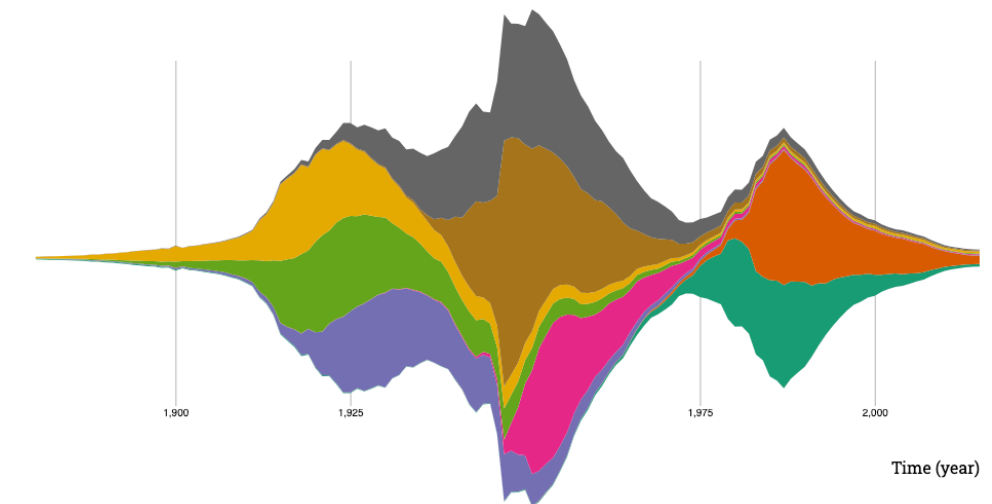
Brath. Visualizing with Text (Fig 7.14). CRC Press, 2020.

Line charts vs filled area charts

- should we reason differently about
 - line chart boundary vs filled area chart interior?
 - stacked area charts vs streamgraphs?
 - discrete stacked bar charts vs continuous streamgraphs?
- what matters?
 - boundary vs interior?
 - discrete vs continuous
 - occlusion?



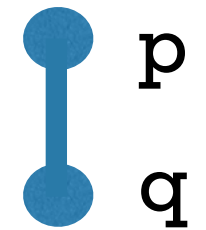
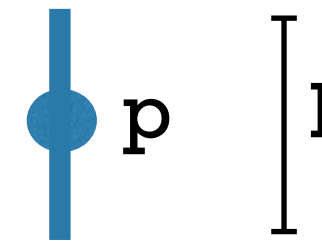
- Amanda
- Ashley
- Betty
- Deborah
- Dorothy
- Helen
- Linda
- Patricia



- <https://d3-graph-gallery.com/line.html>
- <https://d3-graph-gallery.com/area.html>
- <https://d3-graph-gallery.com/stackedarea.html>
- <https://d3-graph-gallery.com/streamgraph.html>

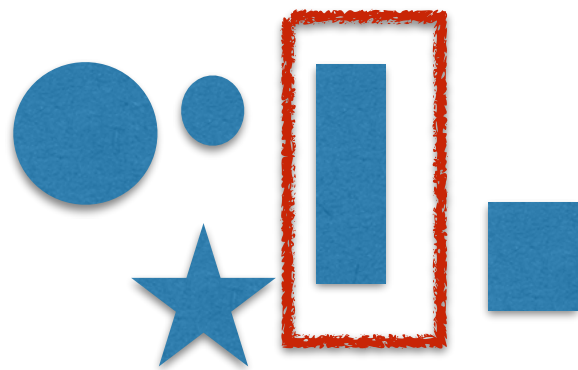
Line marks: what good are they?

- are line/segment marks an unnecessary construct?
 - how does segment differ from "length-coded point mark"?
 - is segment just two point marks in a trenchcoat???
 - two position-coded point marks, connected by segment

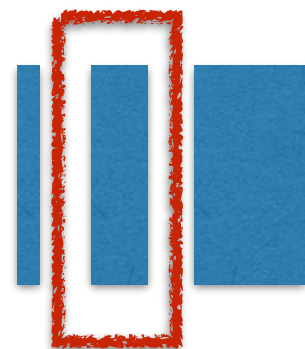


- radical proposal: eliminate line marks as a separate mark type!
 - merge lines and points together, into "unconstrained" marks

➔ Points



➔ Lines



Many, many questions

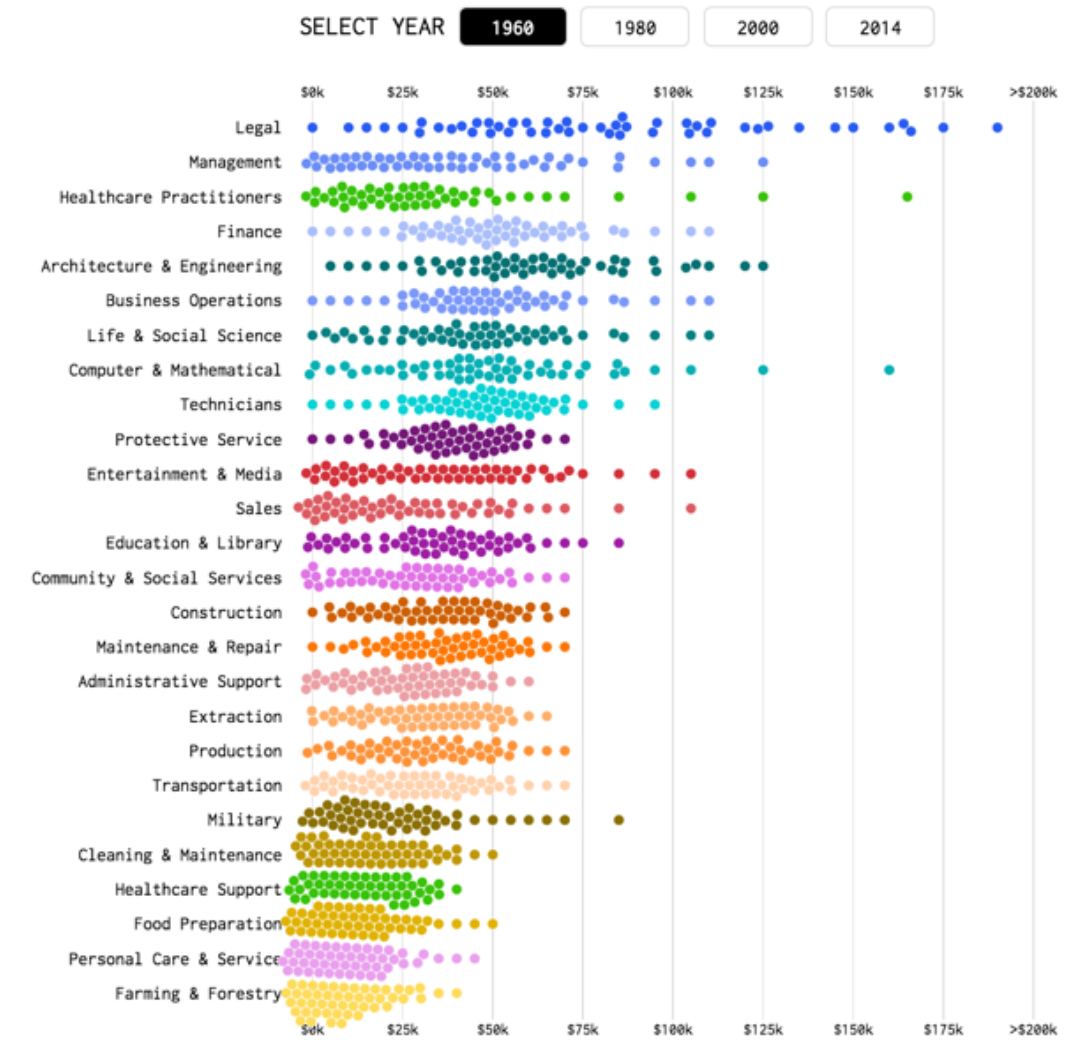
- including
 - Size-coded point marks vs area marks?
 - Area marks vs area channel?
 - What kind of marks are in
 - a tile heatmap?
 - a circle packing?
 - a cartogram?
 - a multi-level thingie with units?
 - Line mark: line segment vs curved path?
 - Do line charts use line marks?
 - Line chart boundaries vs filled area charts interiors?
 - Length-coded point marks vs line marks?

Alternative models: preliminary steps towards answers?

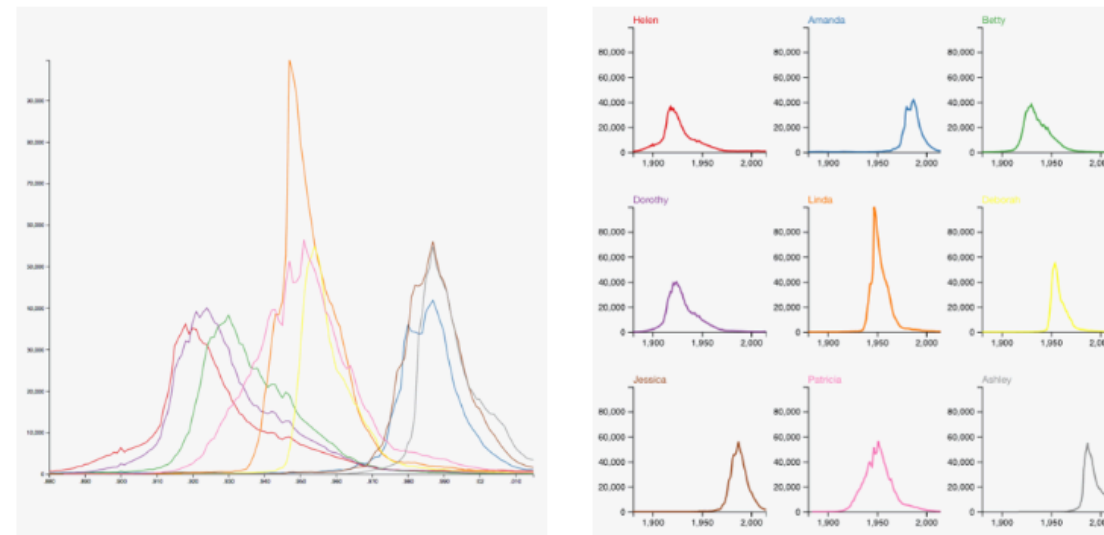
- Old mark/channel model: geometry-based marks
 - 0D points, 1D lines, 2D areas, 3D volumes
- Alternative mark/channel model
 - channel-based analysis: channel availability model
 - Encoded, Unavailable, Free
 - mark-based analysis: mark constraint model
 - Unconstrained (points), Interlocking (areas)
- what best helps us think and reason about design space of visual encoding?
 - combination of both? just one?
 - another alternative?
 - are there other interesting emergent properties arising from bottom-up channel analysis?

Beyond marks and channels

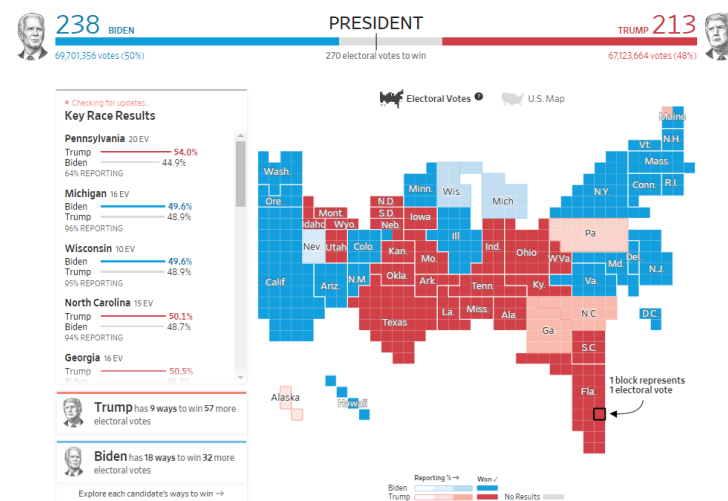
- multi-level analysis requires larger design space
 - small multiples: juxtaposed views
 - vertical position within row: algorithmic, avoid occlusion
 - vertical position across rows: encodes job type attribute
 - superposition:
layered views



<https://flowingdata.com/2016/06/28/distributions-of-annual-income/>



<https://d3-graph-gallery.com/line.html>



<https://www.anychart.com/blog/2020/11/06/election-maps-us-vote-live-results/>

- nesting:
multi-scale
views / glyphs

Rethinking book design space: Visualization Analysis & Design 2e

How?

Encode

➔ Arrange

➔ Express



➔ Order



➔ Use



➔ Separate



➔ Align



➔ Map

from **categorical** and **ordered** attributes

➔ Color

➔ Hue



➔ Saturation



➔ Luminance



➔ Size, Angle, Curvature, ...



➔ Shape



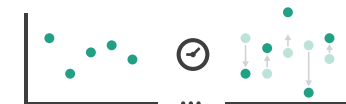
➔ Motion

Direction, Rate, Frequency, ...

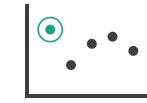


Manipulate

➔ Change



➔ Select



➔ Navigate

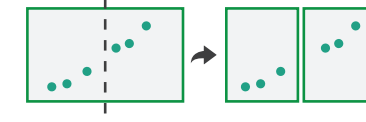


Facet

➔ Juxtapose



➔ Partition



➔ Superimpose

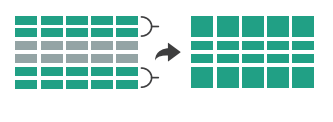


Reduce

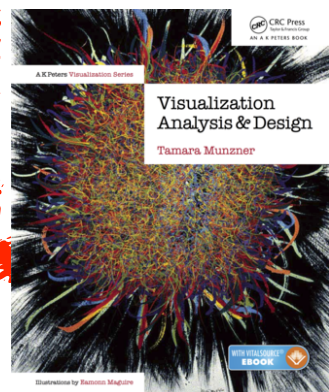
➔ Filter



➔ Aggregate



➔ Embed



More stuff

- this talk

<http://www.cs.ubc.ca/~tmm/talks.html#mit24>

–more questions? thoughts on answers??

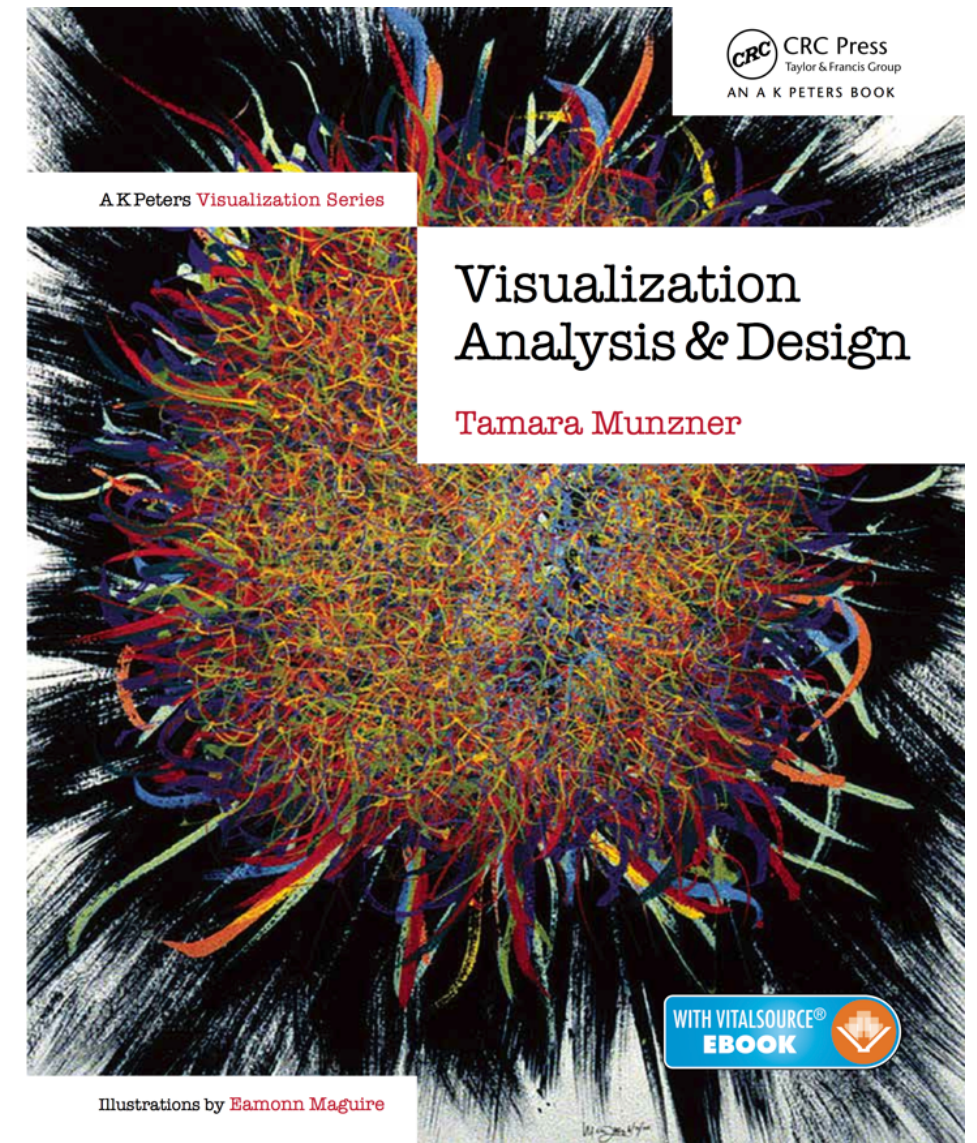
- book

<http://www.cs.ubc.ca/~tmm/vadbook>

- full courses, papers, videos, software, talks

<http://www.cs.ubc.ca/group/infovis>

<http://www.cs.ubc.ca/~tmm>



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

 [@tamara@vis.social](https://medium.com/@tamara@vis.social)

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