Visualization Analysis & Design **Full-Day Tutorial** Session 1

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Sanger Institute / European Bioinformatics Institute June 2014, Cambridge UK

http://www.cs.ubc.ca/~tmm/talks.html#minicourse14



Outline

- Visualization Analysis Framework Session | 9:30-10:45am – Introduction: Definitions
 - -Analysis: What, Why, How
 - Marks and Channels
- Idiom Design Choices, Part 2 Session 3 1:15pm-2:45pm
 - Manipulate: Change, Select, Navigate
 - Facet: Juxtapose, Partition, Superimpose
 - Reduce: Filter, Aggregate, Embed

Idiom Design Choices Session 2 11:00am-12:15pm

- -Arrange Tables
- -Arrange Spatial Data
- -Arrange Networks and Trees
- -Map Color
- Guidelines and Examples Session 4 3-4:30pm
 - -Rules of Thumb
 - -Validation
 - BioVis Analysis Example

http://www.cs.ubc.ca/~tmm/talks.html#minicoursel4

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Defining visualization (vis)

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

Why?...

Why have a human in the loop?

Computer-based visualization systems provide visual representations of datasets designed to help people arry out tasks more effectively. Visualization is suitable when there is a need to augment human capabilities rather than replace people with computational decision-making methods.

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

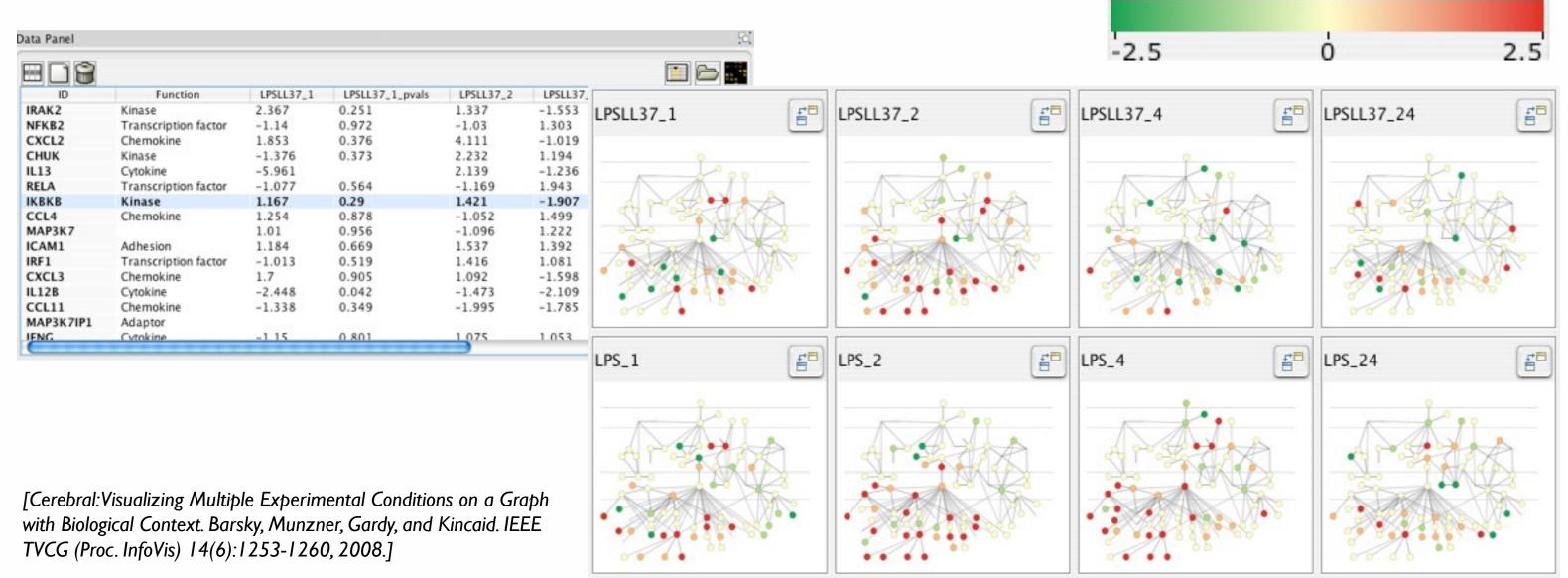




Why use an external representation?

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

• external representation: replace cognition with perception





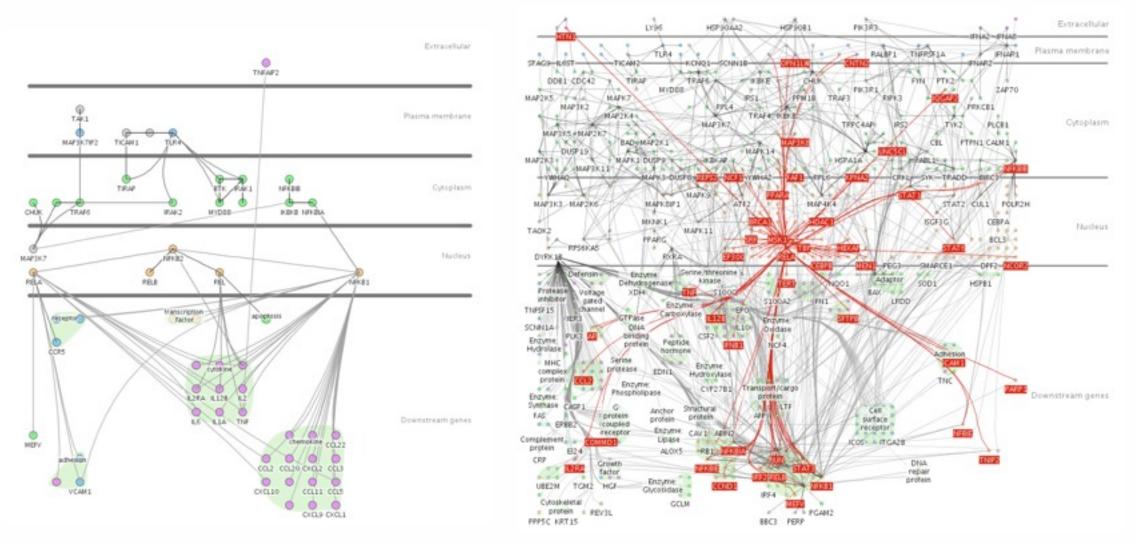
Expression color scale

Why have a computer in the loop?

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

beyond human patience: scale to large datasets, support interactivity

-consider: what aspects of hand-drawn diagrams are important?



[Cerebral: a Cytoscape plugin for layout of and interaction with biological networks using subcellular localization annotation. Barsky, Gardy, Hancock, and Munzner. Bioinformatics 23(8):1040-1042, 2007.] 7

Why depend on vision?

Computer-based visualization systems providevisual epresentations of datasets designed to help people carry out tasks more effectively.

- human visual system is high-bandwidth channel to brain
 - overview possible due to background processing
 - subjective experience of seeing everything simultaneously
 - significant processing occurs in parallel and pre-attentively
- sound: lower bandwidth and different semantics
 - overview not supported
 - subjective experience of sequential stream
- touch/haptics: impoverished record/replay capacity -only very low-bandwidth communication thus far
- taste, smell: no viable record/replay devices

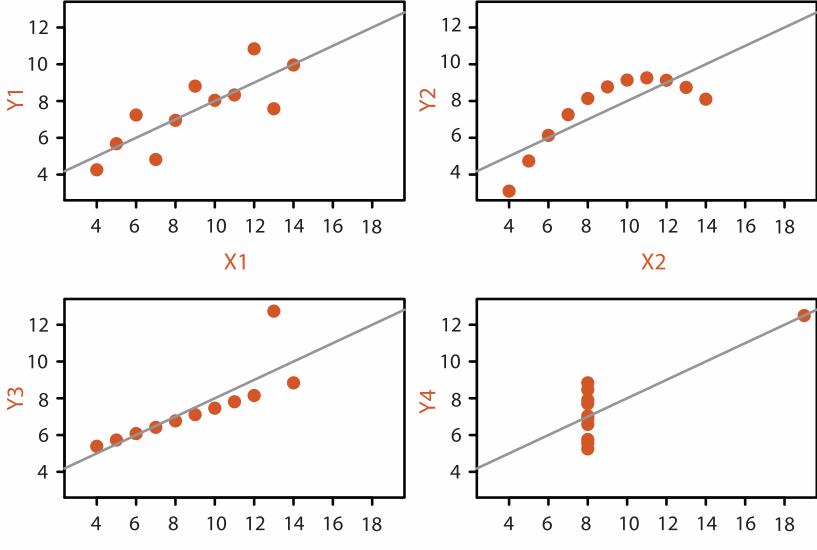
Why show the data in detail?

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

Anscombe's Quartet

Identical statistics

x mean	9
x variance	10
y mean	8
y variance	4
x/v correlation	1



9

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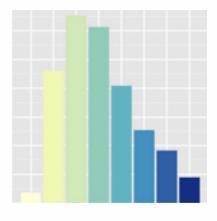
Idiom design space

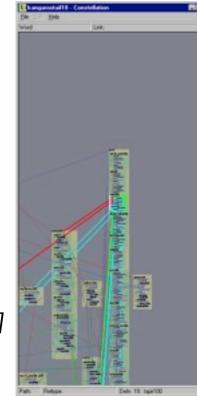
The design space of possible vis idioms is huge, and includes the considerations of both how to create and how to interact with visual representations.

- idiom: distinct approach to creating or manipulating visual representation
 - -how to draw it: **visual encoding** idiom
 - many possibilities for how to create
 - -how to manipulate it: **interaction** idiom
 - even more possibilities
 - make single idiom dynamic
 - link multiple idioms together through interaction

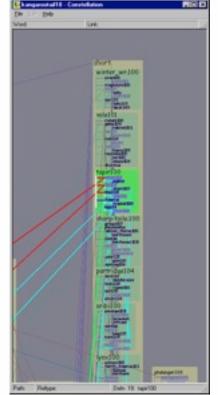
[A layered grammar of graphics. Wickham. Journal of Computational and Graphical Statistics 19:1 (2010), 3–28.]

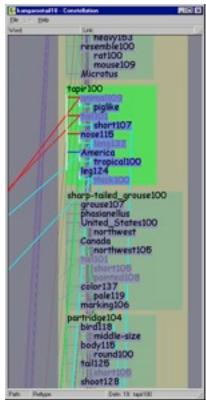
[Interactive Visualization of Large Graphs and Networks. Munzner. Ph.D. thesis, Stanford University Department of Computer Science, 2000.]











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Why focus on tasks and effectiveness?

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

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

Resource limitations

and States and Dag Distille States Vis designers must take into account three very different kinds of resource limitations: those of computers, of humans, and of displays.

- computational limits
 - -processing time
 - -system memory
- human limits
 - human attention and memory
- display limits
 - -pixels are precious resource, the most constrained resource
 - -information density: ratio of space used to encode info vs unused whitespace
 - tradeoff between clutter and wasting space, find sweet spot between dense and sparse



Further reading

• Visualization Analysis and Design. Munzner. AK Peters / CRC Press, Oct 2014. - Chap I: What's Vis, and Why Do It?

Outline

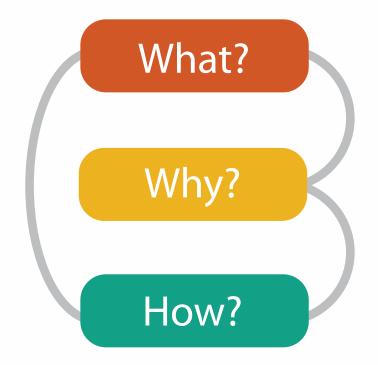
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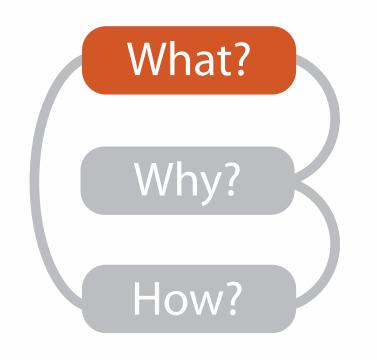
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Analysis: What, why, and how

- what is shown?
 - -data abstraction
- why is the user looking at it? **-task** abstraction
- how is it shown?
 - -idiom: visual encoding and interaction
- abstract vocabulary avoids domain-specific terms -translation process iterative, tricky
- what-why-how analysis framework as scaffold to think systematically about design space





				What?		
	D	atasets				At
 → Data Types → Items → → Data and Dat 	Attributes aset Types	→ Links	→	Positions	→ Grids	 → Attribut → Categ +
Tables	Networks & Trees	Fields		Geometry	Clusters, Sets, Lists	→ Orde
Items Attributes	Items (nodes) Links Attributes	Grids Positions Attributes		Items Positions	Items	 ★ Qua ►
Items (rows)	As (columns) As (columns) A a state of the state of		Node item)	Cell	es (columns)	 → Orderin → Seque → Diverg → Cyclic ↓
→ Geometry	Spatial) Position		(→ Dataset A→ Static	vailability	→ Dynamic

Attributes

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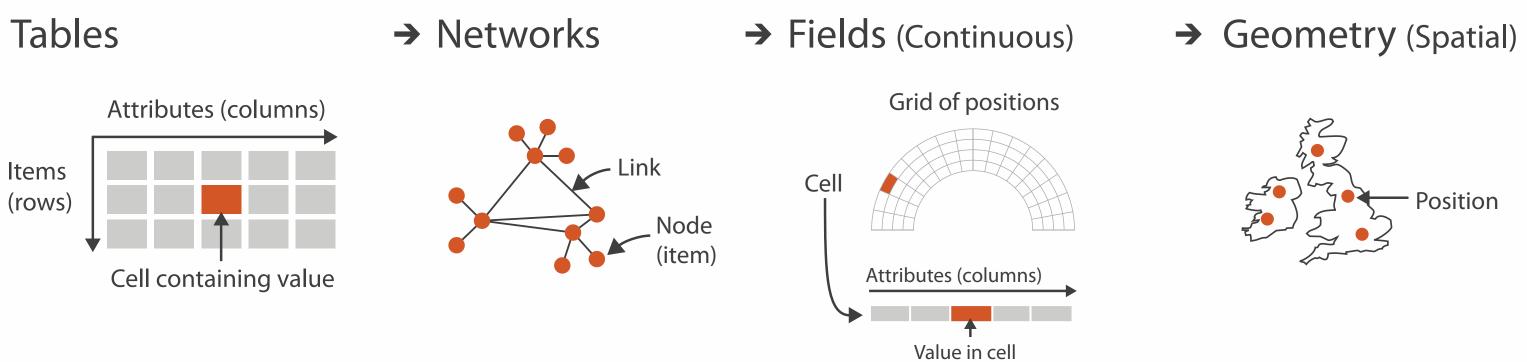
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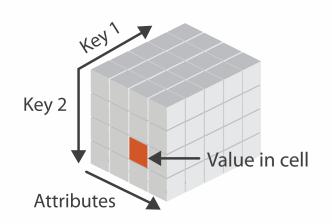


Dataset types

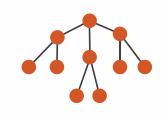
- **Dataset Types** \rightarrow
 - → Tables



→ Multidimensional Table



→ Trees



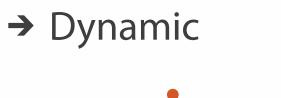
Dataset and data types

Data and Dataset Types (\rightarrow)

Tables	Networks & Trees	Fields	Geometry	Cluster Sets, Lis
ltems	ltems (nodes)	Grids	ltems	Items
Attributes	Links	Positions	Positions	
	Attributes	Attributes		
) Data Types				

- → Positions → Attributes → Links → Items → Grids
- **Dataset Availability** (\rightarrow)
 - → Static

(
ightarrow)





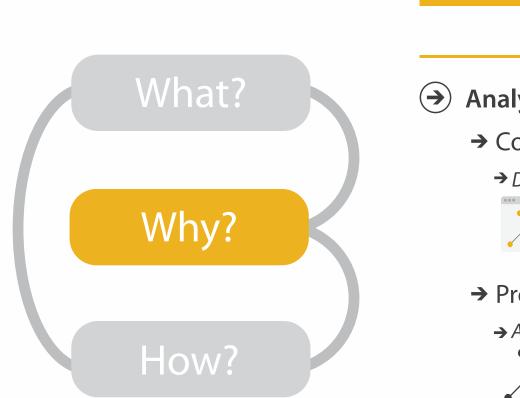


Attribute types

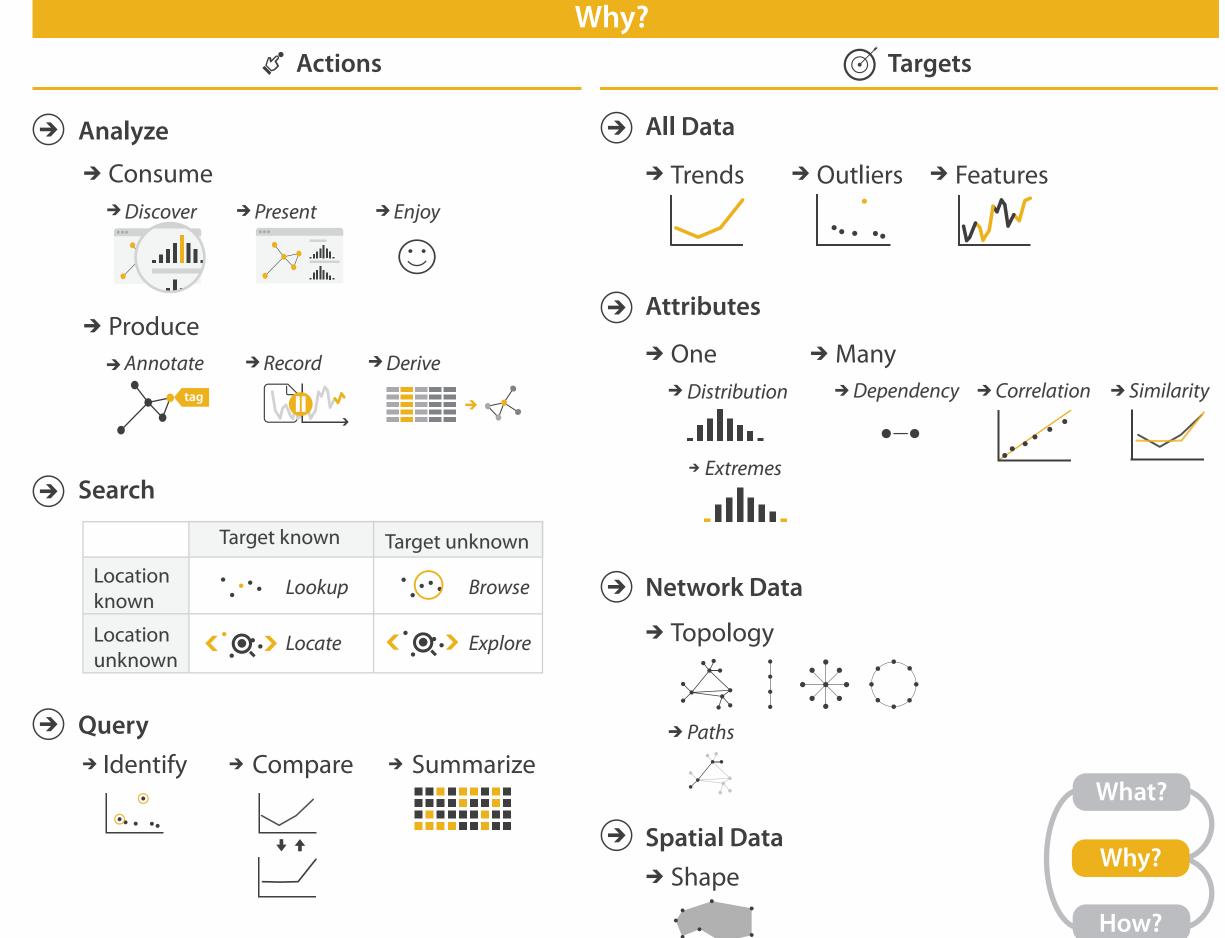








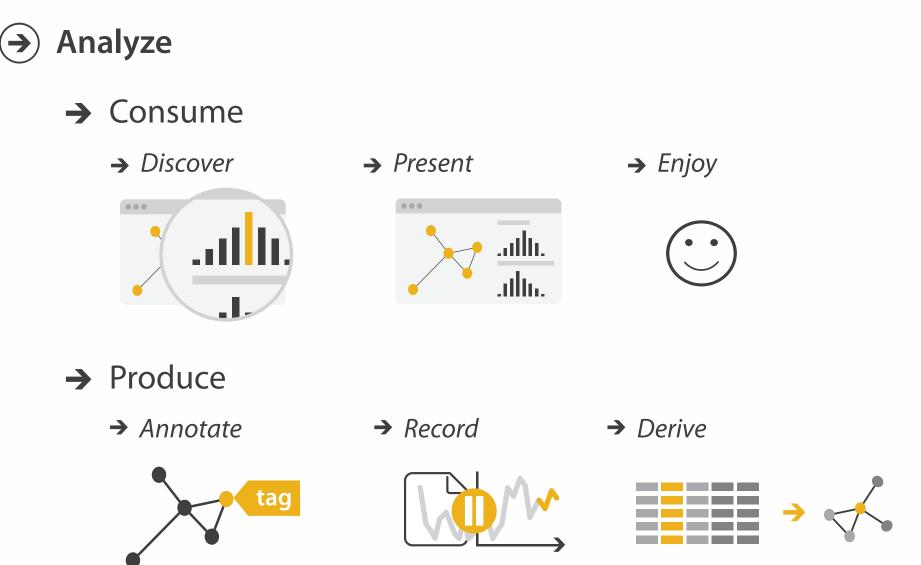
- {action, target} pairs
 - discover distribution
 - compare trends
 - locate outliers
 - browse topology





High-level actions: Analyze

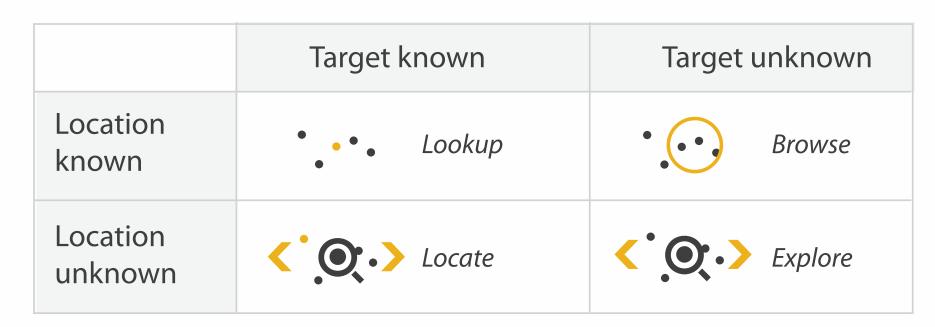
- consume
 - -discover vs present
 - classic split
 - aka explore vs explain
 - -enjoy
 - newcomer
 - aka casual, social
- produce
 - -annotate, record
 - -derive
 - crucial design choice

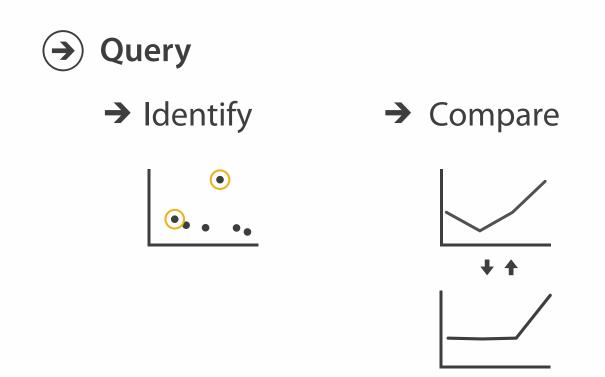


Actions: Mid-level search, low-level query

- what does user know?
 target, location
- how much of the data matters?
 - one, some, all

→ Search



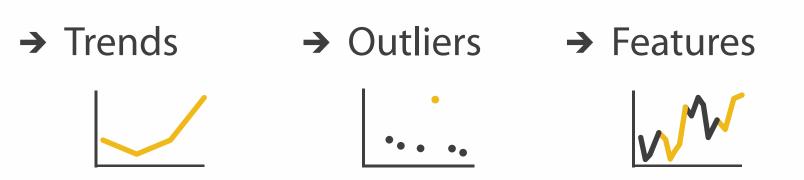






Why: Targets

ALL DATA (\rightarrow)

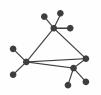


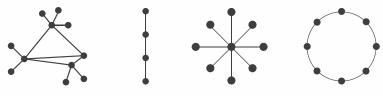
ATTRIBUTES (\rightarrow)

> → One → Many (
> ightarrow)→ Dependency → Correlation → Similarity → Distribution ,111. **↓** *Extremes*

NETWORK DATA (\rightarrow)

→ Topology



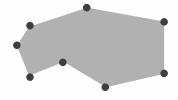


 \rightarrow Paths



SPATIAL DATA

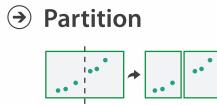
→ Shape



How?

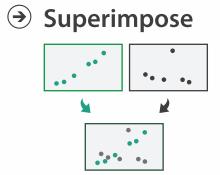
Encode		Manipulate
 → Arrange → Express → Separate 	 Map from categorical and ordered attributes 	Change
→ Order → Align	→ Color → Hue → Saturation → Luminance	→ Select
	Size, Angle, Curvature,	•••
→ Use	•■■ //_))) → Shape + ● ■ ▲	→ Navigate<>
What? Why?	Motion Direction, Rate, Frequency, • • • •	
How?		





→ Aggregate

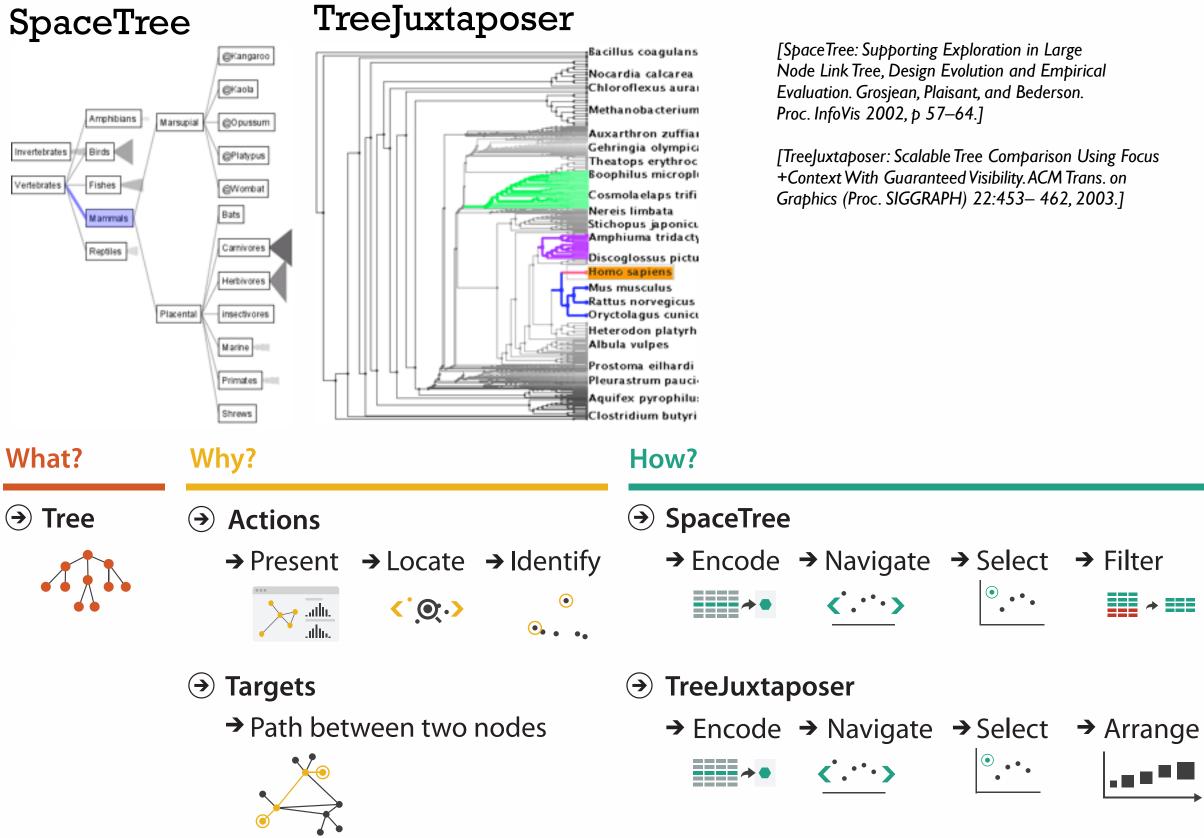
	



→ Embed



Analysis example: Compare idioms



[SpaceTree: Supporting Exploration in Large Node Link Tree, Design Evolution and Empirical Evaluation. Grosjean, Plaisant, and Bederson. Proc. InfoVis 2002, p 57-64.]

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

•••

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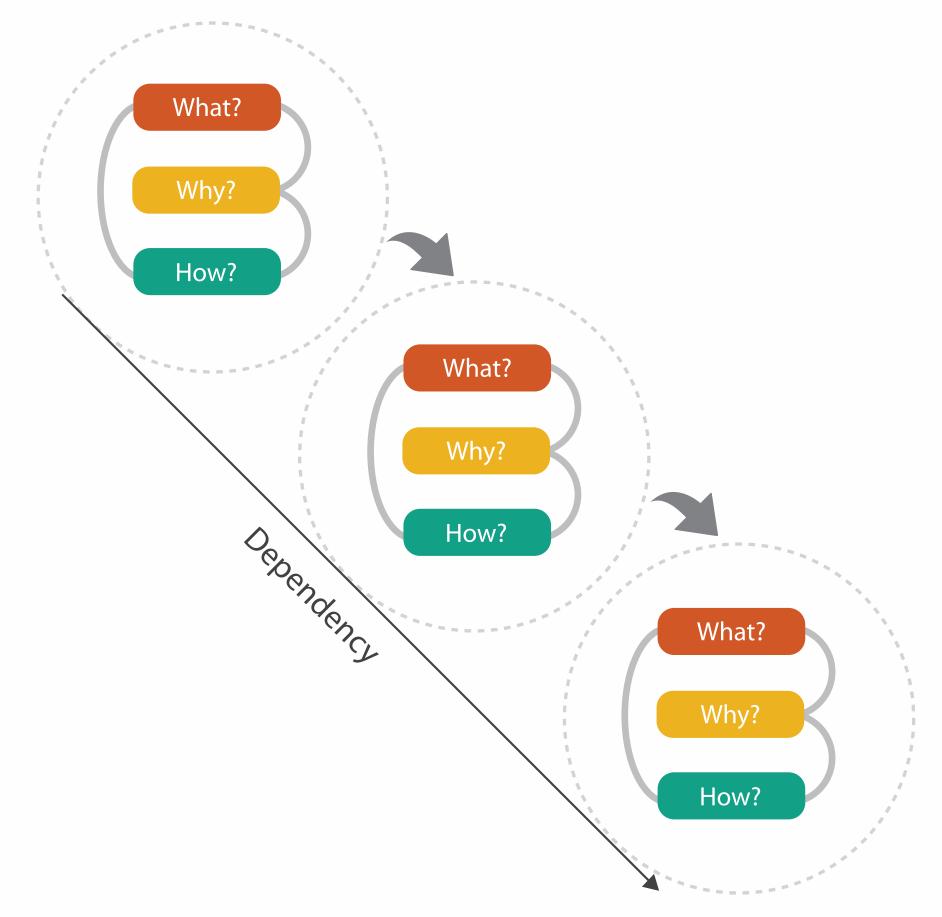


→ Aggregate



Chained sequences

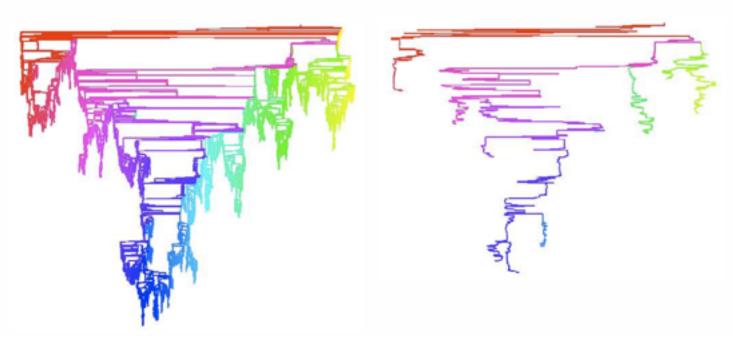
- output of one is input to next
 - -express dependencies
 - -separate means from ends

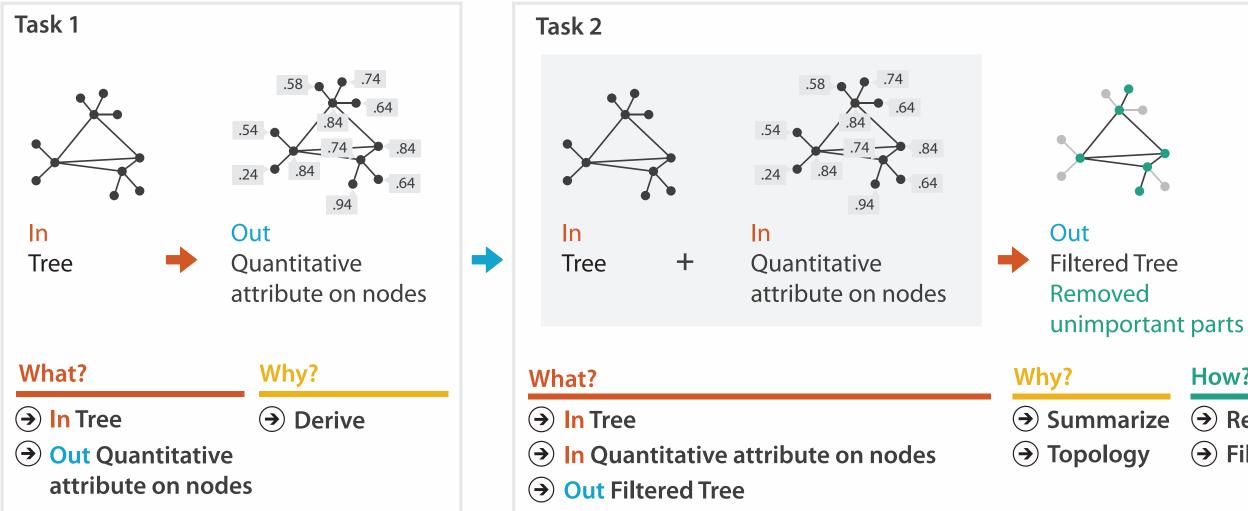


Analysis example: Derive one attribute

- Strahler number
 - centrality metric for trees/networks
 - derived quantitative attribute
 - draw top 5K of 500K for good skeleton

[Using Strahler numbers for real time visual exploration of huge graphs. Auber. Proc. Intl. Conf. Computer Vision and Graphics, pp. 56–69, 2002.]





How?

→ Reduce → Filter

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Further reading

- Visualization Analysis and Design. Munzner. AK Peters / CRC Press, Oct 2014.
 - Chap 2: What: Data Abstraction
 - Chap 3: Why: Task Abstraction
- A Multi-Level Typology of Abstract Visualization Tasks. Brehmer and Munzner. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis) 19:12 (2013), 2376–2385. Low-Level Components of Analytic Activity in Information Visualization. Amar, Eagan, and
- Stasko. Proc. IEEE InfoVis 2005, p 111–117.
- A taxonomy of tools that support the fluent and flexible use of visualizations. Heer and Shneiderman. Communications of the ACM 55:4 (2012), 45–54.
- Rethinking Visualization: A High-Level Taxonomy. Tory and Möller. Proc. IEEE InfoVis 2004, p 151–158.
- Visualization of Time-Oriented Data. Aigner, Miksch, Schumann, and Tominski. Springer, 2011.

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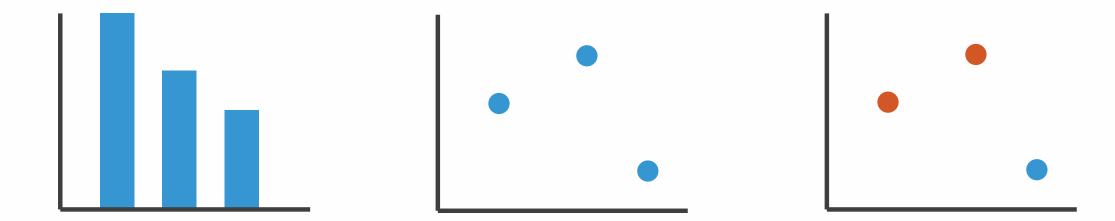
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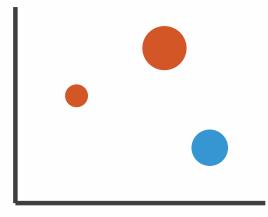
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Visual encoding

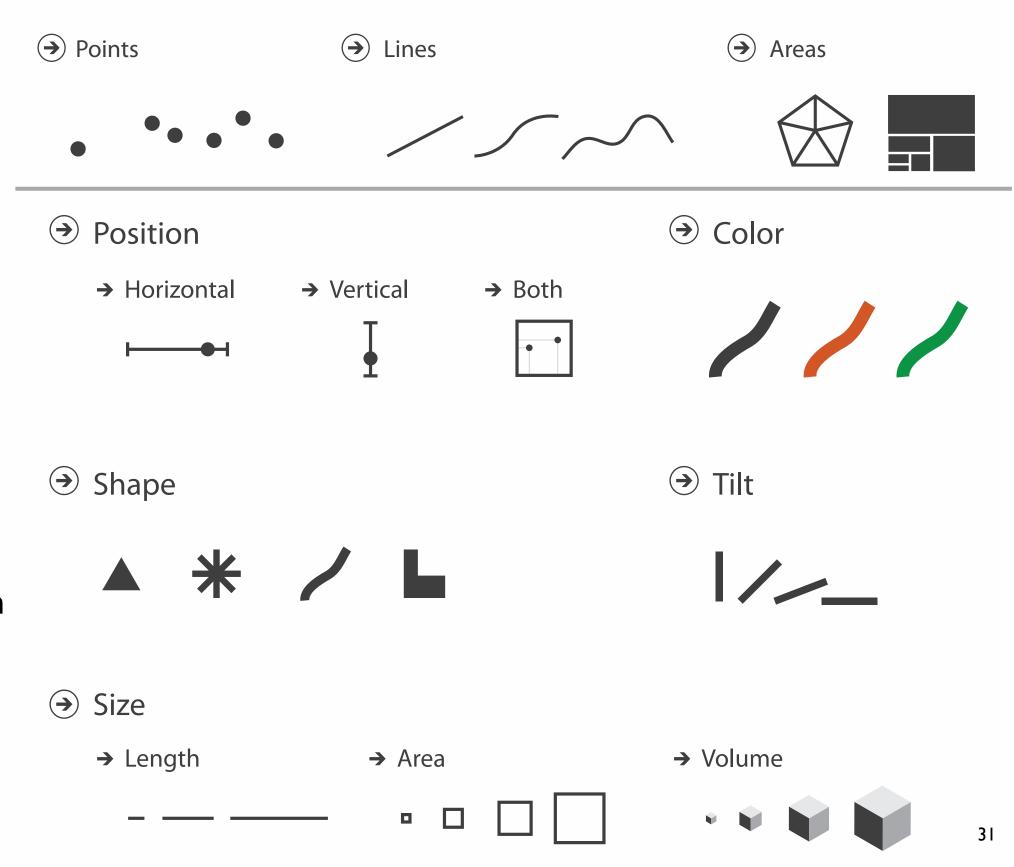
• analyze idiom structure





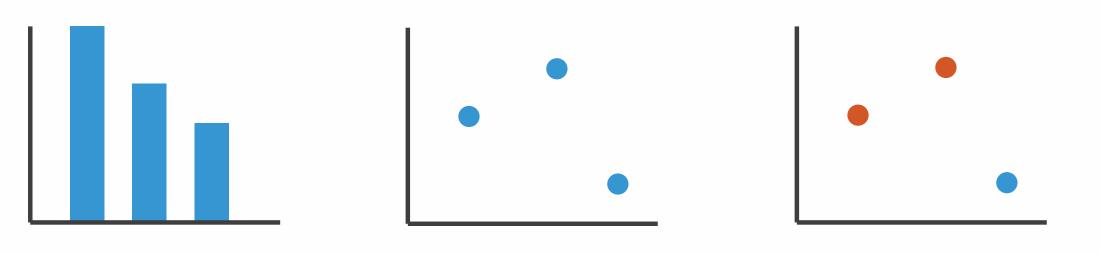
Definitions: Marks and channels

- marks
 - -geometric primitives
- channels
 - -control appearance of marks
 - can redundantly code with multiple channels
- interactions
 - point marks only convey position; no area constraints
 - can be size and shape coded
 - -line marks convey position and length
 - can only be size coded in ID (width)
 - -area marks fully constrained
 - cannot be size or shape coded



Visual encoding

- analyze idiom structure
 - -as combination of marks and channels



1: vertical position

2: vertical position horizontal position 3:

vertical position horizontal position color hue

mark: line

mark: point

mark: point



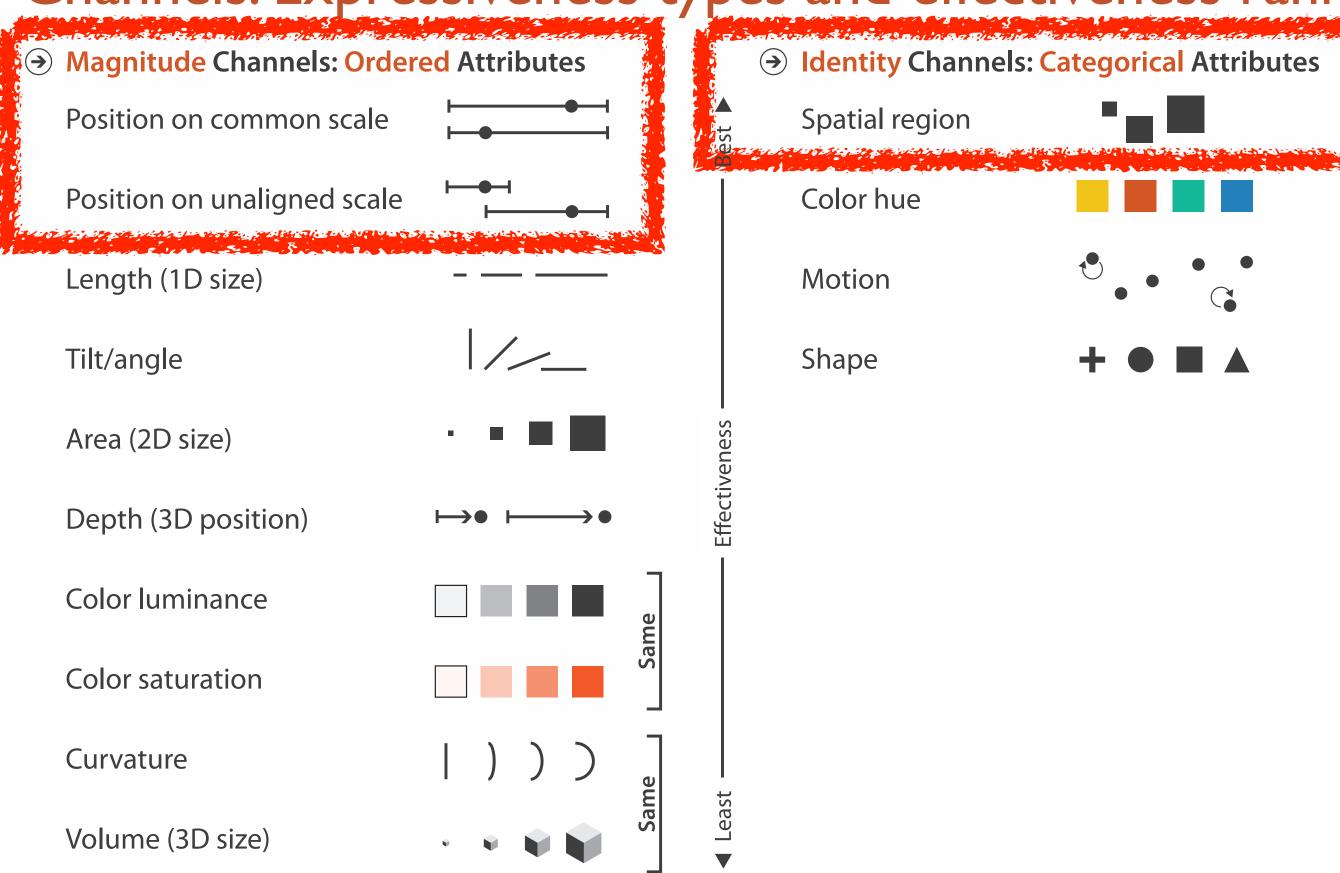


4:

vertical position horizontal position color hue size (area)

mark: point

Channels: Expressiveness types and effectiveness rankings









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Effectiveness and expressiveness principles

• effectiveness principle

-encode most important attributes with highest ranked channels

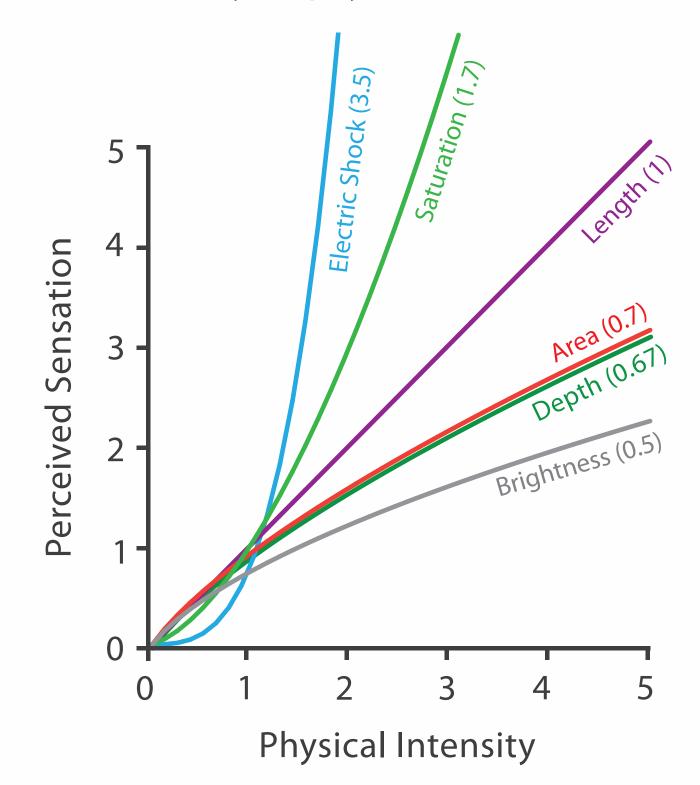
- expressiveness principle
 - -match channel and data characteristics

[Automating the Design of Graphical Presentations of Relational Information. Mackinlay. ACM Trans. on Graphics (TOG) 5:2 (1986), 110–141.]

- rankings: where do they come from?
 - -accuracy
 - discriminability
 - separability
 - -popout

Accuracy: Fundamental Theory

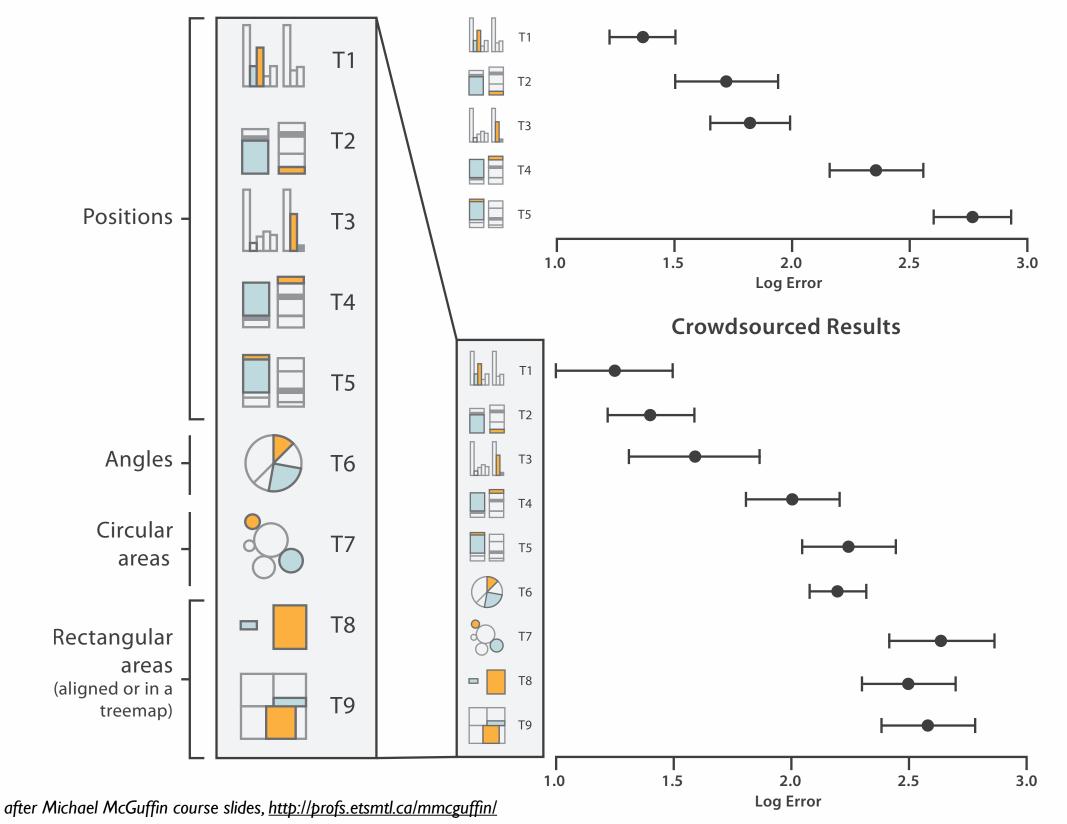
Steven's Psychophysical Power Law: S= I^N



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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?

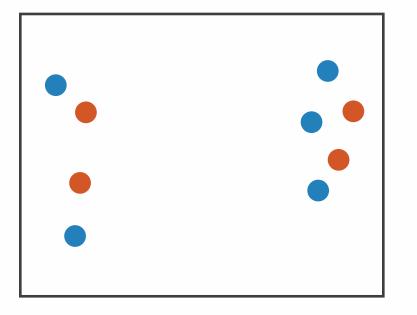
• linewidth: only a few



[mappa.mundi.net/maps/maps 014/telegeography.html]

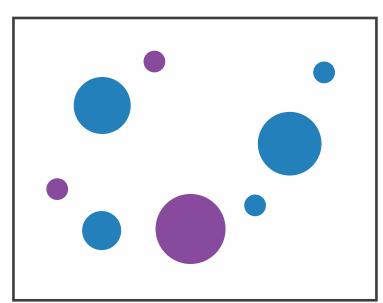
Separability vs. Integrality

Position + Hue (Color)



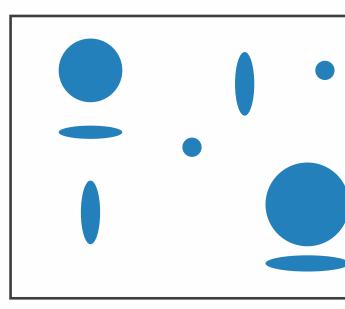
Fully separable

Size + Hue (Color)



Some interference

Width + Height



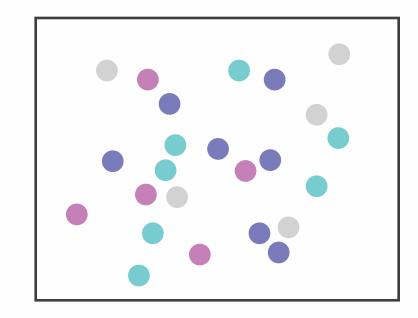
Some/significant interference

2 groups each

2 groups each

3 groups total: integral area

Red + Green

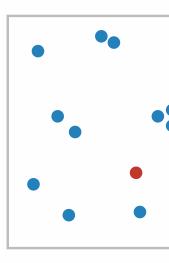


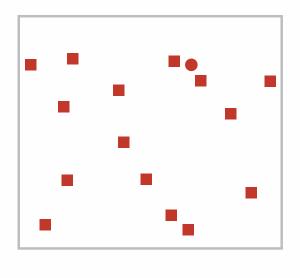
Major interference

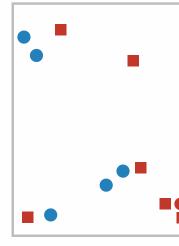
4 groups total: integral hue

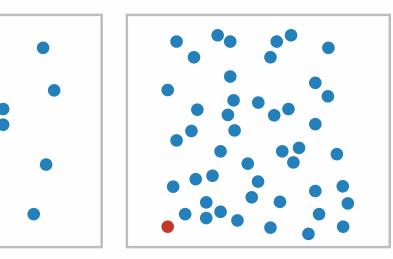
Popout

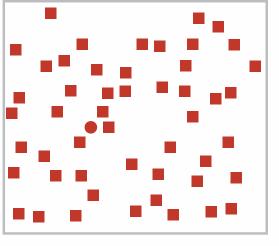
- find the red dot
 how long does it take?
- parallel processing on many individual channels
 - -speed independent of distractor count
 - speed depends on channel and amount of difference from distractors
- serial search for (almost all) combinations
 speed depends on number of distractors

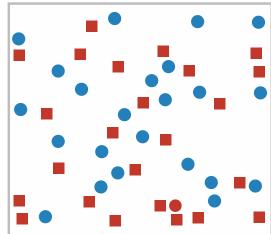




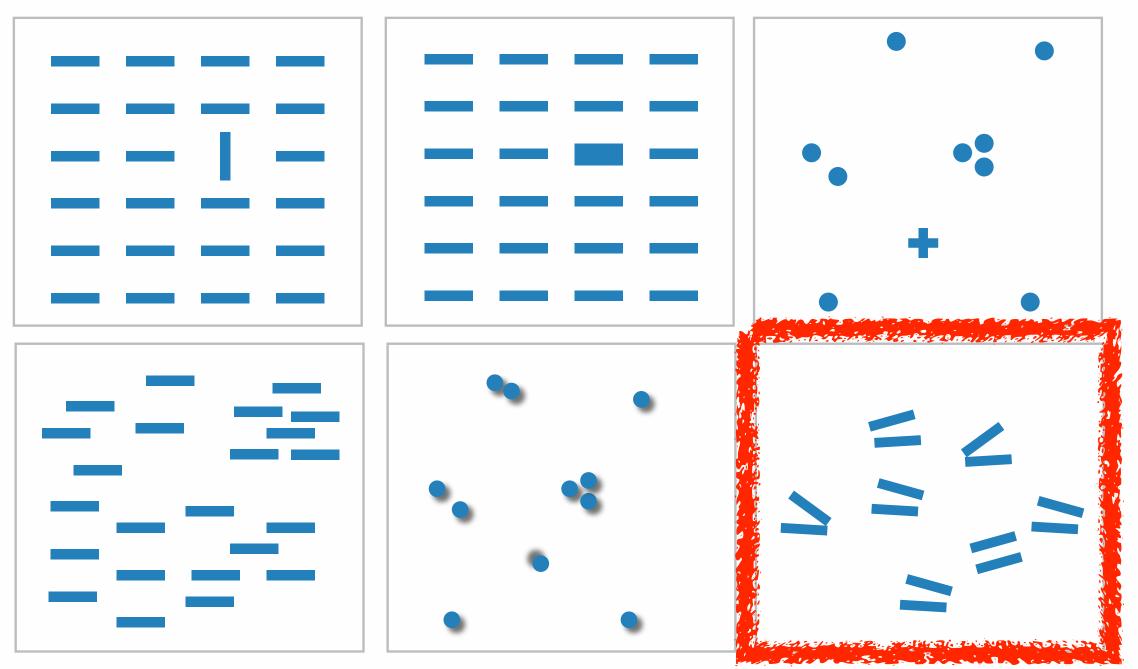








Popout

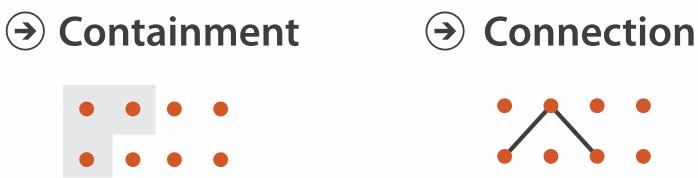


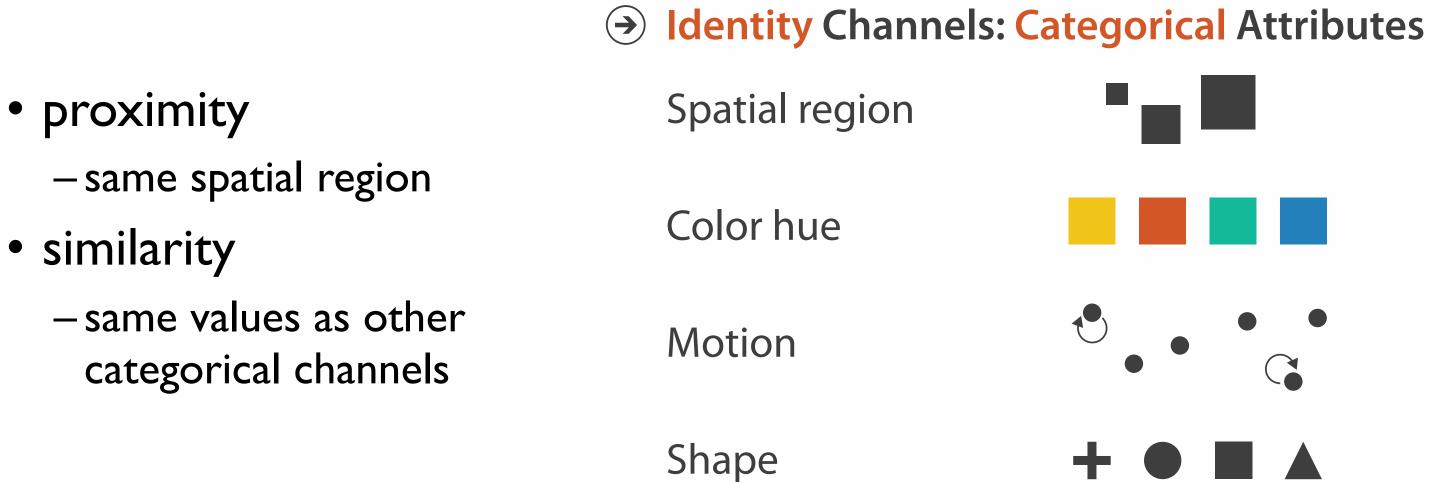
- many channels: tilt, size, shape, proximity, shadow direction, ...
- but not all! parallel line pairs do not pop out from tilted pairs

Grouping

- containment
- connection

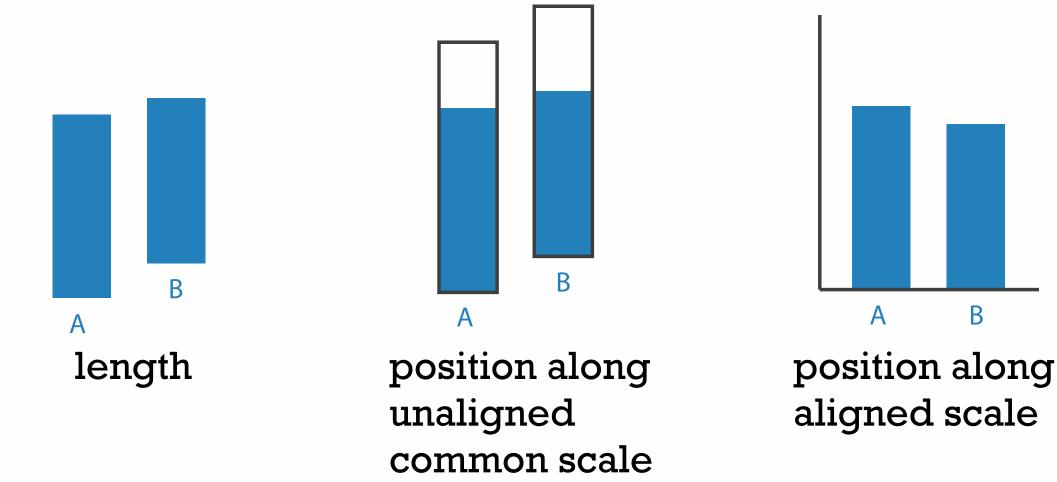
Marks as Links





Relative vs. absolute judgements

- 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



42 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.]

B

Further reading

- Visualization Analysis and Design. Munzner. AK Peters / CRC Press, Oct 2014. - Chap 5: Marks and Channels
- On the Theory of Scales of Measurement. Stevens. Science 103:2684 (1946), 677–680.
- Psychophysics: Introduction to its Perceptual, Neural, and Social Prospects. Stevens. Wiley, 1975.
- Graphical Perception: Theory, Experimentation, and Application to the Development of Graphical Methods. Cleveland and McGill. Journ. American Statistical Association 79:387 (1984), 531–554.
- Perception in Vision. Healey. <u>http://www.csc.ncsu.edu/faculty/healey/PP</u>
- Visual Thinking for Design. Ware. Morgan Kaufmann, 2008.
- Information Visualization: Perception for Design, 3rd edition. Ware. Morgan Kaufmann / Academic Press, 2004.