Information Visualization Data, Tasks, Nested Model *Ex: Abstractions*

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Week 2: 14 September 2022

http://www.cs.ubc.ca/~tmm/courses/547-22

Course Logistics

Async so far

- last week
 - -async read only
 - Course Logistics (no comments, no responses)
 - -async read & comment
 - VAD Ch I: Why Visualization? (comments only, no responses)
 - -async discuss
 - self-intros
- this week
 - -async read & comment & respond
 - VAD Ch 2: Data Abstraction
 - VAD Ch 3: Task Abstraction
 - paper: Nested Model [basis for VAD Ch 4]

Updates

- All students moved from waitlist to registered – official enrolment now 33
- Canvas link added
 - -future: assignment handin

-soon: marks for sync & async participation (posted weekly)

Mini-Lecture, Q&A: Round 1

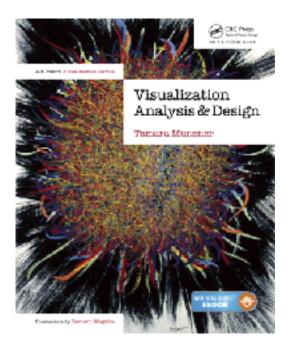
Visualization Analysis & Design

Data Abstraction (Ch 2)

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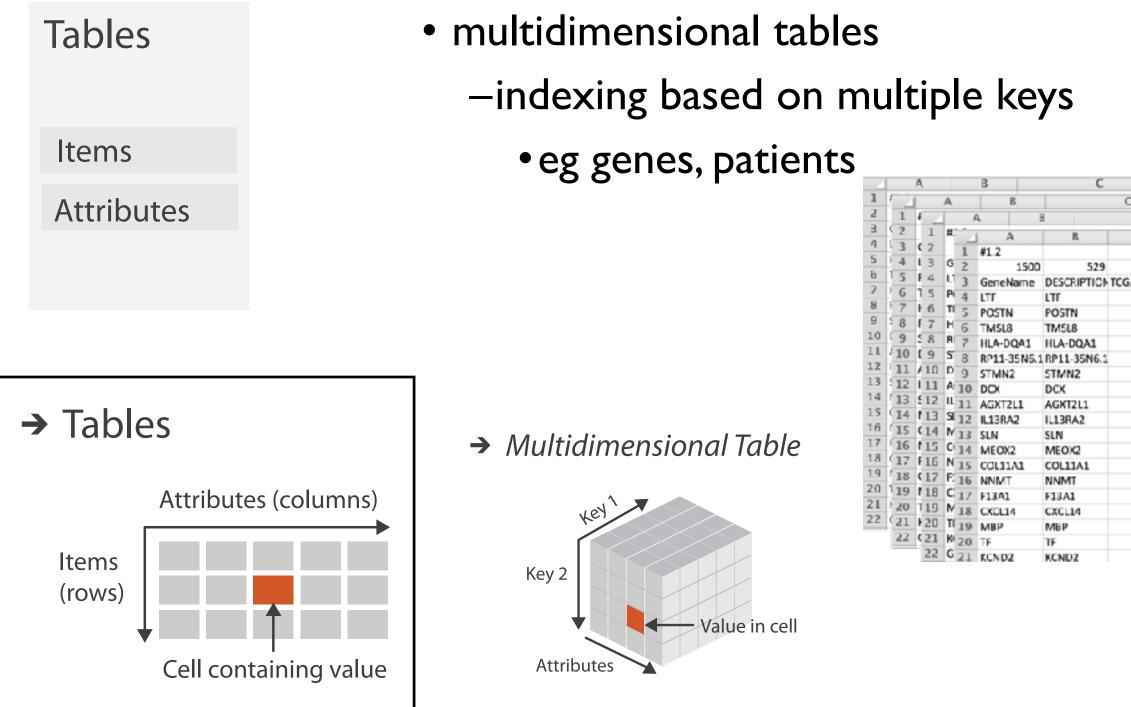


Data abstraction: Three operations

- translate from domain-specific language to generic visualization language
- identify dataset type(s), attribute types
- identify cardinality
 - -how many items in the dataset?
 - -what is cardinality of each attribute?
 - number of levels for categorical data
 - range for quantitative data
- consider whether to transform data
 - -guided by understanding of task

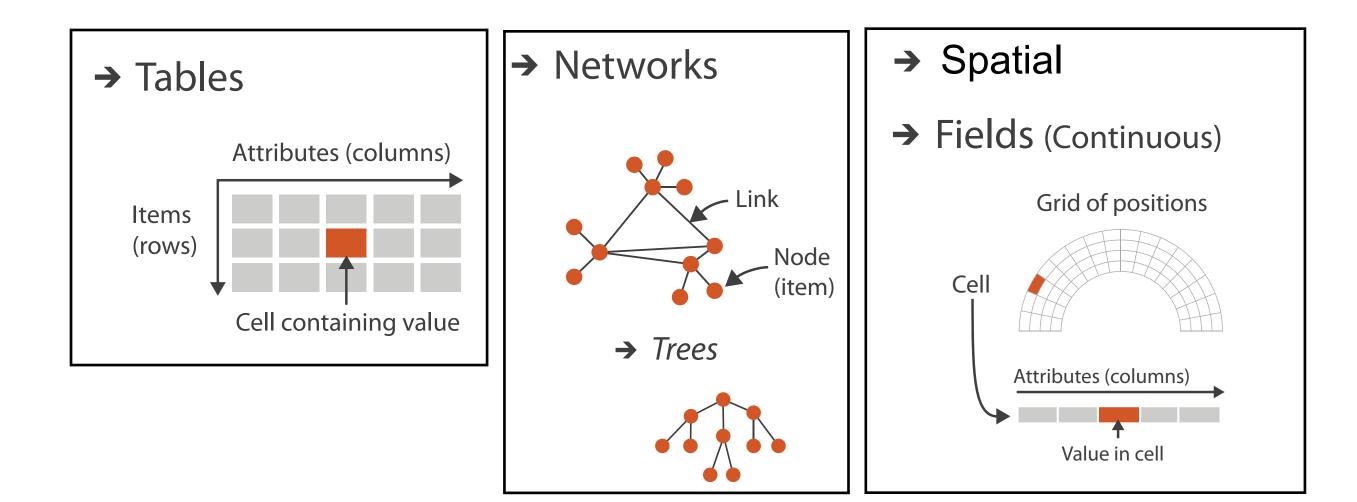
attributes: name, age, shirt size, fave fruit

Tables	 flat table 			,		,
	-one item per row				¥ 1	
	–each column is attribute	ID	Name	Age	Shirt Size	Favorite Fruit
Items	-cell holds value for	1	Amy	8	S	Apple
Attributes	item-attribute pair	2	Basil	7	S	Pear
	-unique key	3	Clara	9	М	Durian
	(could be implicit)	4	Desmond	13	L	Elderberry
(could be implicit)			Ernest	12	L	Peach
→ Tables		6	Fanny	10	S	Lychee
		7	George	9	М	Orange
Attributes (columns) Items (rows)			Hector	8	L	Loquat
			Ida	10	М	Pear
			Amy	12	М	Orange
\downarrow						//////////////////////////////////////
Cell con	i	item: person				



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	2.6624	11805		3.	932400	324		5.031585	377	
	-3.0822	17838		-2.	243148	513		-0.02313	581	
	-1.7396	64398		4.	.577962	344		3.127744	964	
	-3.3463	52968		-2.	895400	157		-3.473035	067	
	-2.5785	11106		-3.	051605	144		-1.729892	386	
	-2.260	78975		-2.	529795	801		-2.844966	278	
	-2.6394	93611		-3.	113204	863		-0.403975	027	
	-2.935	96915		-1.	873600	916		2.976256	911	
	-2.4657	18221		-2.	208406	749		1.025827	904	
	-2.3950	54056		-1.	062676	046		1.783235	317	
	1.2119	34832		-0.	399392	588		4.733608	974	
	0.7037	45154		0.	664082	419		3.069030	715	
	-0.2240	94042		2.	222197	544		1.171354	775	
	-8.18	09694		-1	395056	071		2.569540	659	
	-1.9053	90566		-2.	037626	447		-2.935744	906	
	-4.3341	23292		-4.	680680	246		-2.975788	866	T.
	-1.7776	92395		-2.	100352	021		-1.996306	032	

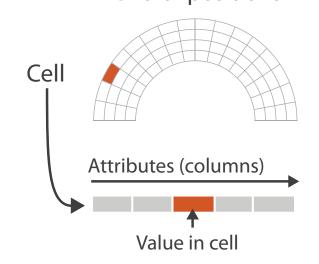
Tables	Networks & Trees	Fields
ltems	Items (nodes)	Grids
Attributes	Links	Positions
	Attributes	Attributes

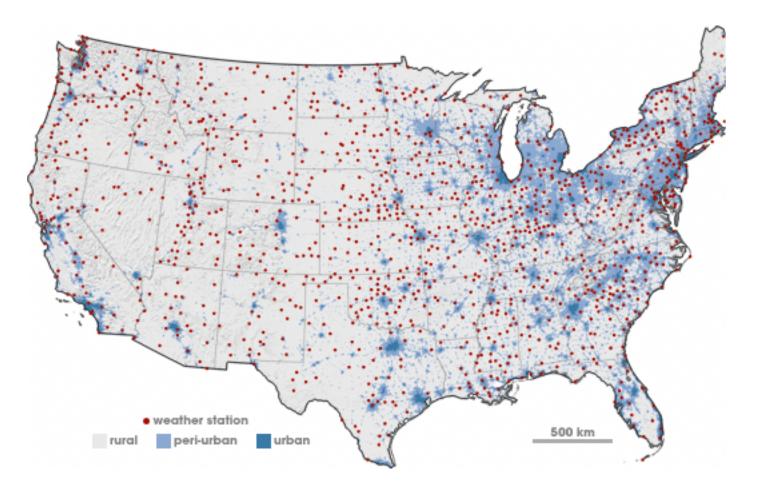


Spatial fields

- attribute values associated w/ cells
- cell contains value from continuous domain
 - eg temperature, pressure, wind velocity
- measured or simulated
 - → Spatial
 - → Fields (Continuous)







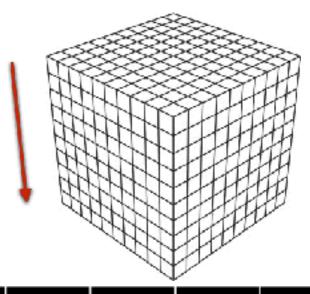
Spatial fields

- attribute values associated w/ cells
- cell contains value from continuous domain
 - eg temperature, pressure, wind velocity
- measured or simulated
- major concerns
 - sampling:
 where attributes are measured
 - interpolation:

how to model attributes elsewhere

- grid types

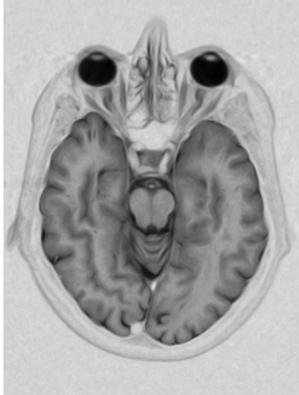




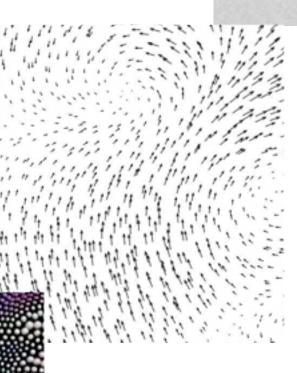
S			

Spatial fields

- attribute values associated w/ cells
- cell contains value from continuous domain
 - eg temperature, pressure, wind velocity
- measured or simulated
- major concerns
 - sampling:
 where attributes are measured
 - interpolation:
 how to model attributes elsewhere
 - -grid types
- major divisions
 - attributes per cell: scalar (1), vector (2), tensor (many)

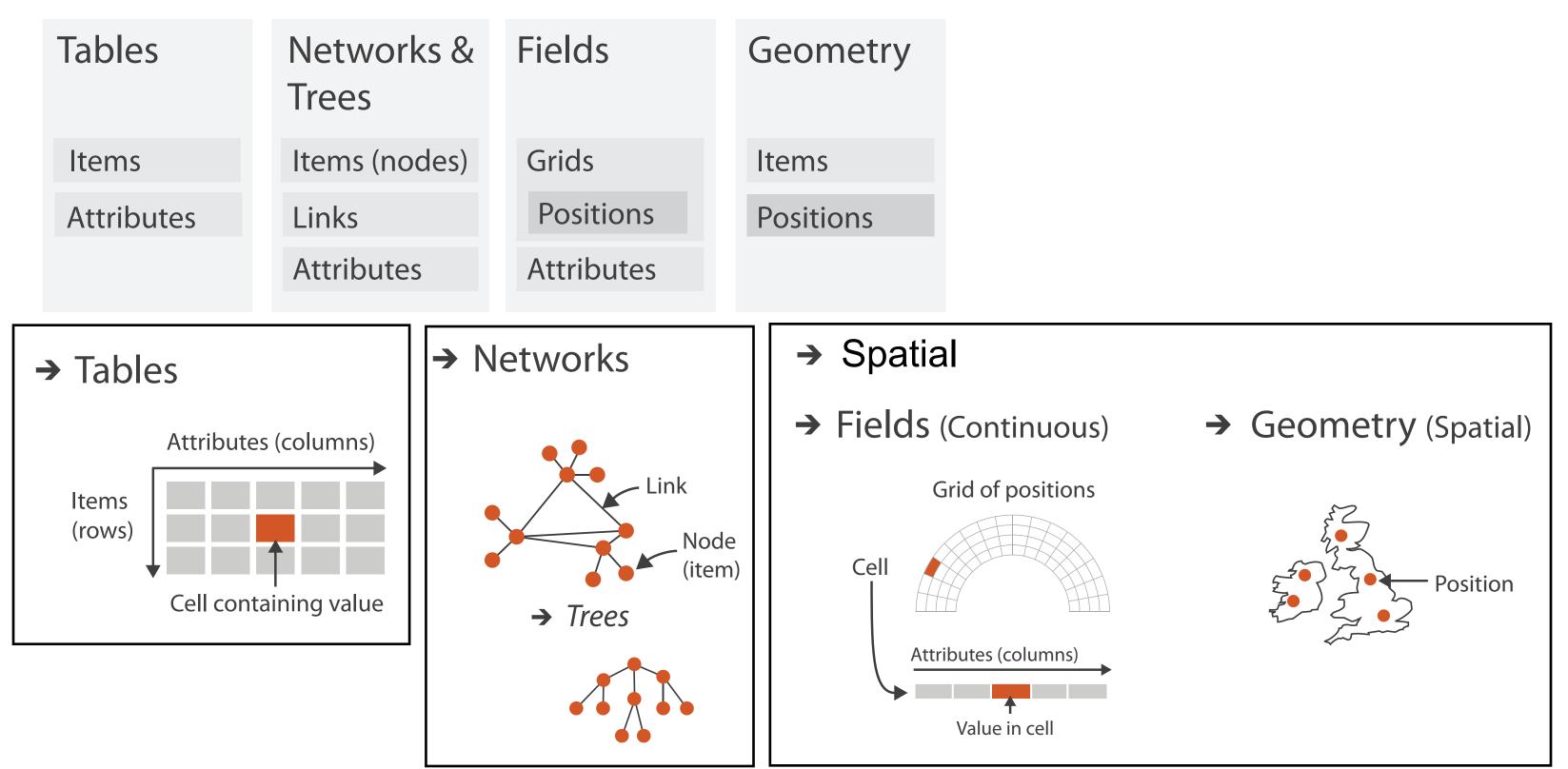


scalar



vector

tensor



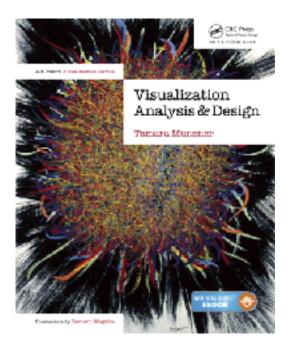
Visualization Analysis & Design

Analysis: Nested Model (Ch 4)

Tamara Munzner

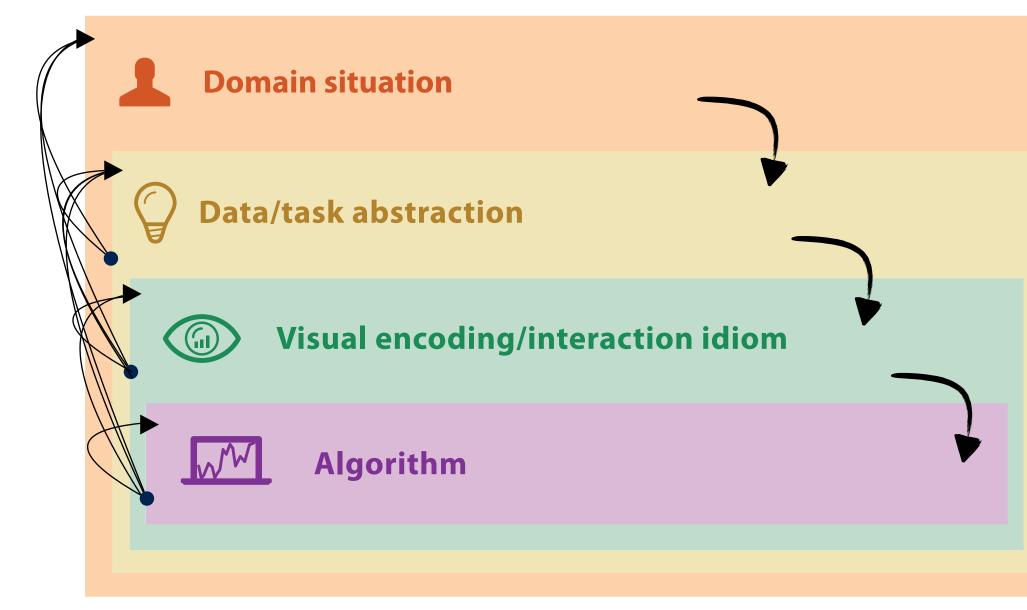
Department of Computer Science University of British Columbia

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Nested model

- downstream: cascading effects
- upstream: iterative refinement





How to evaluate a visualization: So many methods, how to pick?

- Computational benchmarks?
 - -quant: system performance, memory
- User study in lab setting?
 - -quant: (human) time and error rates, preferences
 - -qual: behavior/strategy observations
- Field study of deployed system?
 - -quant: usage logs
 - -qual: interviews with users, case studies, observations
- Analysis of results?
 - -quant: metrics computed on result images
 - -qual: consider what structure is visible in result images
- Justification of choices?
 - -qual: perceptual principles, best practices

Analysis examples: Single paper includes only subset of methods

MatrixExplorer. Henry and Fekete. InfoVis 2006.

observe and interview target users

justify encoding/interaction design

measure system time/memory

qualitative result image analysis

LiveRAC. McLachlan, Munzner, Koutsofios, and North. CHI 2008.

observe and interview target users

justify encoding/interaction design

qualitative result image analysis

field study, document deployed usage

An energy model for visual graph clustering. (LinLog) Noack. Graph Drawing 2003

qualitative/quantitative image analysis

Effectiveness of animation in trend visualization. Robertson et al. InfoVis 2008.

lab study, measure time/errors for operation

Interactive visualization of genealogical graphs.

McGuffin and Balakrishnan. InfoVis 2005.

justify encoding/interaction design

qualitative result image analysis test on target users, get utility anecdotes

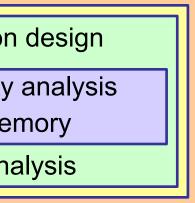
Flow map layout. Phan et al. InfoVis 2005.

justify encoding/interaction design

computational complexity analysis

measure system time/memory

qualitative result image analysis



Paper Types

Paper types

- each has different contributions, validation methods, structure
 - design studies
 - -technique/algorithm
 - -evaluation
 - model/taxonomy
 - system

http://ieeevis.org/year/2017/info/call-participation/infovis-paper-types

Paper types: Validation

- design studies
 - -qualitative discussion of result images/videos
 - -abstraction & idiom validation: case studies, field studies, design justification
- technique/algorithm
 - qualitative discussion of result images/videos
 - -algorithm validation for algorithm papers: computational benchmarks
 - -idiom validation for technique papers: controlled experiments
- evaluation
 - -(controlled experiment as primary contribution)
- theory/model/taxonomy
 - -show power: descriptive, generative, evaluative, (predictive)
- system

-show power for developer using system

Paper structures

- typical research paper vs expectations for this course final report - more on implementation
 - novel research contribution not required

<u>http://www.cs.ubc.ca/~tmm/courses/547-17/projectdesc.html#outlines</u>

Reading visualization papers

• one strategy: multiple passes

-title

- -abstract, authors/affiliation
- -flip through, glance at figures, notice structure from section titles
- skim intro, results/discussion (maybe conclusion)
- -fast read to get big ideas
 - if you don't get something, just keep going
- second pass to work through details
 - later parts may cast light on earlier parts for badly structured papers
- -third pass to dig deep
 - if it's highly relevant, or you're presenting it to class
- literature search

-decide when to stop reading: is this relevant to my current concerns?

Literature search

- this course: I will give you seed papers during our I on I meetings
- forwards vs backwards search
 - Google Scholar forward citations!
 - -only a subset of forwards & backwards citations will be what you need
- building up landscape
 - -authors/affiliations will have more signal as you develop expertise

Exercise: Abstractions



Data abstraction: Three operations

- translate from domain-specific language to generic visualization language
- identify dataset type(s), attribute types
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 - -how many items in the dataset?
 - -what is cardinality of each attribute?
 - number of levels for categorical data
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- consider whether to transform data
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Now: In-class design exercise, in small groups

Abstractions

- -practice with data & task abstractions, on concrete example: Aid to Countries
- -crucial ideas: determine cardinalities/ranges
 - -precondition for all decisions about visual encoding
- Small-group exercise: 60-ish min
 - -breakout groups (4 people/group)
 - -googledoc worksheets, as before
 - -document in your group's googledoc w/ text as you go!
 - -reportbacks, as before (intermediate and final)
 - -I'll flip through googledocs, some questions for group spokesperson

Next week

- to read & discuss (async, before next class)
 VAD book, Ch 5: Marks & Channels
 - –VAD book, Ch 6: Rules of Thumb
 - -paper: Design Study Methodology

Backup/Reference Slides



Ch 1. What's Vis, and Why Do It?

Defining visualization (vis)

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

Why?...

Visualization (vis) defined & motivated

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.

- human in the loop needs the details & no trusted automatic solution exists
 - -doesn't know exactly what questions to ask in advance
 - -exploratory data analysis
 - **speed up** through human-in-the-loop visual data analysis
 - -present known results to others
 - -stepping stone towards automation
 - -before model creation to provide understanding
 - -during algorithm creation to refine, debug, set parameters
 - -before or during deployment to build trust and monitor

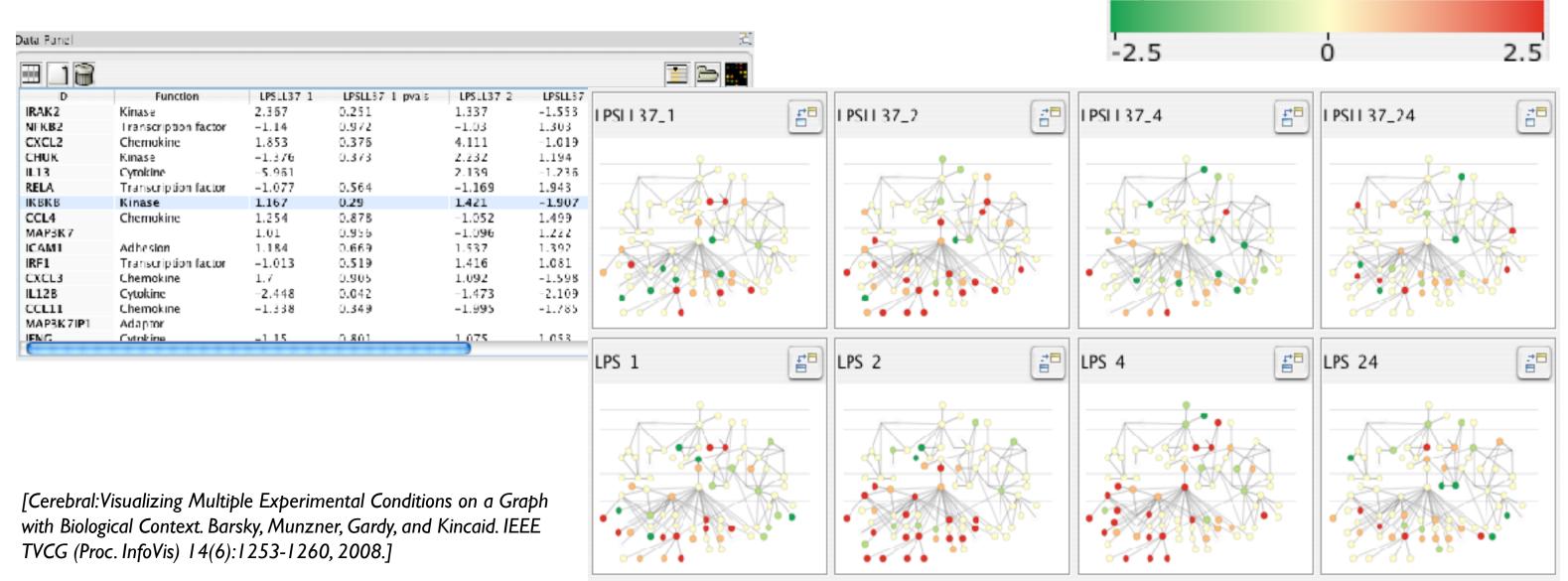




Why use an external representation?

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

external representation: replace cognition with perception





Expression color scale

Why depend on vision?

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

- 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 represent all the data?

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

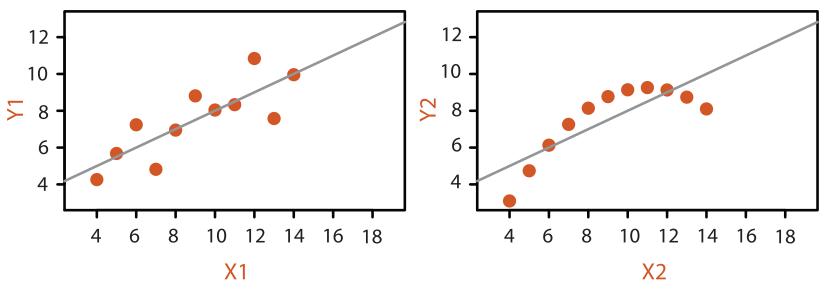
- summaries lose information, details matter
 - confirm expected and find unexpected patterns
 - -assess validity of statistical model

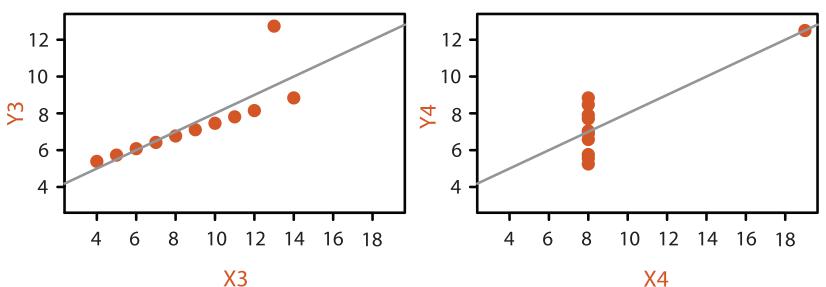
Anscombe's Quartet

Identical statistics			
x mean	9		
x variance	10		
y mean	7.5		
y variance	3.75		
x/y correlation	0.816		

https://www.youtube.com/watch?v=DbJyPELmhJc

Same Stats, Different Graphs



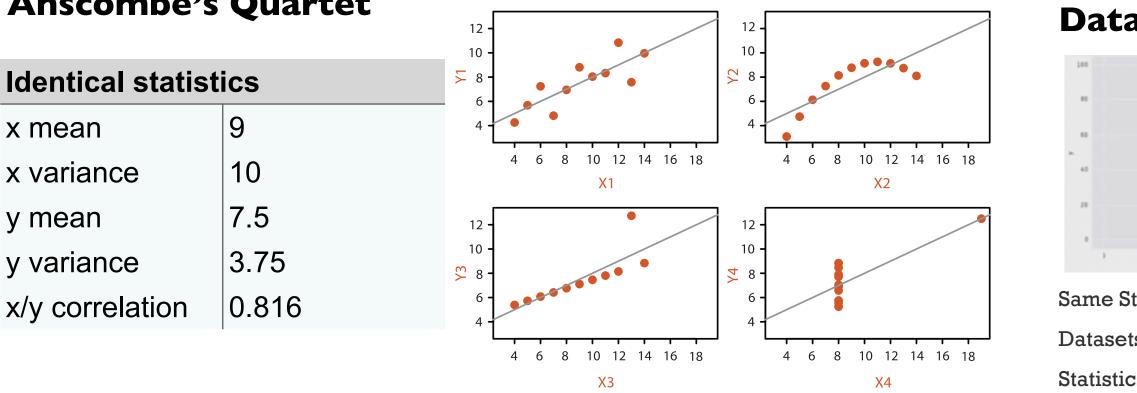




Visualization defined & motivated

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

- suitable when human in the loop needs details
 - interplay between human judgement and automatic computation



Anscombe's Quartet

Datasaurus Dozen

	ř	•		54.2600597 47.8349704
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20 4	10 10 X	80 100		

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

Why focus on tasks and effectiveness?

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

- effectiveness requires match between data/task and representation - set of representations is huge
 - -many are ineffective mismatch for specific data/task combo
 - -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
- how to validate effectiveness
 - -many methods, must pick appropriate one for your context



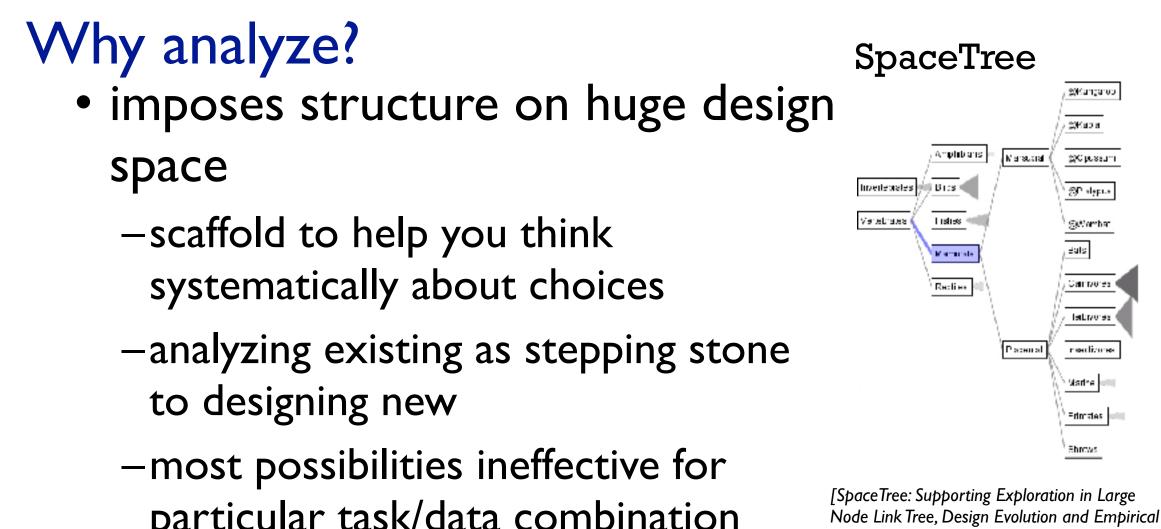
37

What resource limitations are we faced with?

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

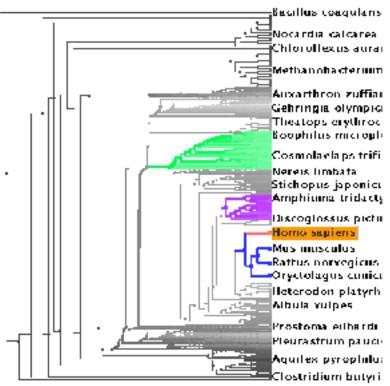
- computational limits
 - -processing time
 - -system memory
- human limits
 - human attention and memory
- display limits
 - -pixels are precious resource, the most constrained resource
 - -information density: ratio of space used to encode info vs unused whitespace
 - tradeoff between clutter and wasting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



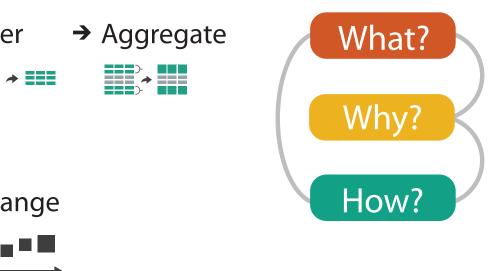


What?	Why?	How?	Evaluation. Grosjean, Plaisant, and Be Proc. InfoVis 2002, p 57–64.]	derson.
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	→ Targets	→ TreeJuxta	poser	
	Path between two nodes	→ Encode	→ Navigate → Select	→ Arran
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TreeJuxtaposer



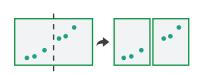
[TreeJuxtaposer: Scalable Tree Comparison Using Focus+Context With Guaranteed Visibility. ACM Trans. on Graphics (Proc. SIGGRAPH) 22:453–462, 2003.]



How?

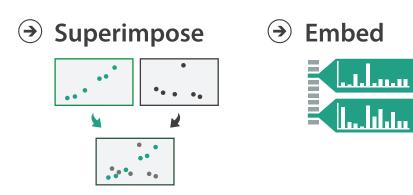
Encode		Manipulate
 → Arrange → Express → Separate 	 Map from categorical and ordered attributes 	 Change Chang
→ Order → Align	$\rightarrow Color$ $\rightarrow Hue \qquad \Rightarrow Saturation \qquad \Rightarrow Luminance$	→ Select
	→ Size, Angle, Curvature,	
→ Use	• ■ ■ //_)))) → Shape + ● ■ ▲	→ Navigate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





→ Aggregate





Further reading

- Visualization Analysis and Design. Munzner. AK Peters Visualization Series, CRC Press, 2014. - Chap I: What's Vis, and Why Do It?
- The Nature of External Representations in Problem Solving. Jiajie Zhang. Cognitive Science 21:2 (1997), 179-217.
- A Representational Analysis of Numeration Systems. Jiajie Zhang and Donald A. Norman. Cognition 57 (1995), 271-295.
- Why a Diagram Is (Sometimes) Worth Ten Thousand Words.. Jill H. Larkin and Herbert A. Simon. Cognitive Science 11:1 (1987), 65-99.
- Graphs in Statistical Analysis.F.J. Anscombe. American Statistician 27 (1973), 17-21.
- Design Study Methodology: Reflections from the Trenches and the Stacks. Michael SedImair, Miriah Meyer, and Tamara Munzner. IEEE Trans. Visualization and Computer Graphics (Proc. InfoVis 2012), **18(12):2431-2440, 2012**.
- Information Visualization: Perception for Design, 3rd edition, Colin Ware, Morgan Kaufmann, 2013.
- Current approaches to change blindness Daniel J. Simons. Visual Cognition 7, 1/2/3 (2000), 1-15.
- Semiology of Graphics, Jacques Bertin, Gauthier-Villars 1967, EHESS 1998
- The Visual Display of Quantitative Information. Edward R. Tufte. Graphics Press, 1983.

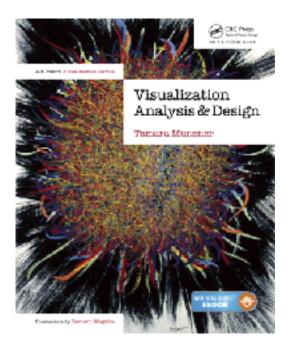
Visualization Analysis & Design

Data Abstraction (Ch 2)

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- 14, 2.6, 30, 30, 15, 100001
- What does this sequence of six numbers mean?

44

14, 2.6, 30, 30, 15, 100001

- What does this sequence of six numbers mean?
 - two points far from each other in 3D space?

45

14, 2.6, 30, 30, 15, 100001

- What does this sequence of six numbers mean?
 - two points far from each other in 3D space?
 - two points close to each other in 2D space, with 15 links between them, and a weight of 100001 for the link?

14, 2.6, 30, 30, 15, 100001

- What does this sequence of six numbers mean?
 - two points far from each other in 3D space?
 - two points close to each other in 2D space, with 15 links between them, and a weight of 100001 for the link?
 - something else??

14, 2.6, 30, 30, 15, 100001

- What does this sequence of six numbers mean?
 - two points far from each other in 3D space?
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 - something else??

Basil, 7, S, Pear

• What about this data?

14, 2.6, 30, 30, 15, 100001

- What does this sequence of six numbers mean?
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 - two points close to each other in 2D space, with 15 links between them, and a weight of 100001 for the link?
 - something else??

- What about this data?
 - food shipment of produce (basil & pear) arrived in satisfactory condition on 7th day of month

14, 2.6, 30, 30, 15, 100001

- What does this sequence of six numbers mean?
 - two points far from each other in 3D space?
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 - something else??

- What about this data?
 - food shipment of produce (basil & pear) arrived in satisfactory condition on 7th day of month
 - Basil Point neighborhood of city had 7 inches of snow cleared by the Pear Creek Limited snow removal service

14, 2.6, 30, 30, 15, 100001

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 - something else??

- What about this data?
 - food shipment of produce (basil & pear) arrived in satisfactory condition on 7th day of month
 - Basil Point neighborhood of city had 7 inches of snow cleared by the Pear Creek Limited snow removal service
 - lab rat Basil made 7 attempts to find way through south section of maze, these trials used pear as reward food

Now what?

• semantics: real-world meaning

Amy	8
Basil	7
Clara	9
Desmond	13
Ernest	12
Fanny	10
George	9
Hector	8
Ida	10
Amy	12

S Apple S Pear Durian М Elderberry L Peach L S Lychee М Orange Loquat L М Pear М Orange

Now what?

semantics: real-world meaning

Name	Age
Amy	8
Basil	7
Clara	9
Desmond	13
Ernest	12
Fanny	10
George	9
Hector	8
Ida	10
Amy	12

Shirt Size	Favorite Fruit
S	Apple
S	Pear
М	Durian
L	Elderberry
L	Peach
S	Lychee
М	Orange
L	Loquat
Μ	Pear
М	Orange

Now what?

- semantics: real-world meaning
- data types: structural or mathematical interpretation of data
 - item, link, attribute, position, (grid)
 - different from data types in programming!

Name	Age
Amy	8
Basil	7
Clara	9
Desmond	13
Ernest	12
Fanny	10
George	9
Hector	8
Ida	10
Amy	12

Shirt Size	Favorite Fruit
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S	Pear
М	Durian
L	Elderberry
L	Peach
S	Lychee
М	Orange
L	Loquat
Μ	Pear
М	Orange

- item: individual entity, discrete
 - eg patient, car, stock, city
 - -"independent variable"

Name	Age
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Shirt Size	Favorite Fruit
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Name	Age	Shirt Size	Favorite Fruit
Amy	8	S	Apple
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Ernest	12	L	Peach
Fanny	10	S	Lychee
George	9	М	Orange
Hector	8	L	Loquat
Ida	10	М	Pear
Amy	12	М	Orange

item: person

- item: individual entity, discrete
 - eg patient, car, stock, city
 - "independent variable"
- attribute: property that is measured, observed, logged...
 - -eg height, blood pressure for patient
 - -eg horsepower, make for car
 - -"dependent variable"

Name	Age	Shirt Size	Favorite Fruit
Amy	8	S	Apple
Basil	7	S	Pear
Clara	9	М	Durian
Desmond	13	L	Elderberry
Ernest	12	L	Peach
Fanny	10	S	Lychee
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Hector	8	L	Loquat
Ida	10	М	Pear
Amy	12	М	Orange

item: person

- item: individual entity, discrete
 - eg patient, car, stock, city
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 - -eg height, blood pressure for patient
 - -eg horsepower, make for car
 - -"dependent variable"

attributes:	name,	ag
-------------	-------	----

	Current and a second and a seco
Name	Age
Amy	8
Basil	7
Clara	9
Desmond	13
Ernest	12
Fanny	10
George	9
Hector	8
Ida	10
Amy	12

item: person

ge, shirt size, fave fruit

Shirt Size	Favorite Fruit		
S	Apple		
S	Pear		
М	Durian		
L	Elderberry		
L	Peach		
S	Lychee		
М	Orange		
L	Loquat		
М	Pear		
М	Orange		

Other data types

- links
 - -express relationship between two items
 - -eg friendship on facebook, interaction between proteins
- positions
 - -spatial data: location in 2D or 3D
 - -pixels in photo, voxels in MRI scan, latitude/longitude
- grids

- sampling strategy for continuous data

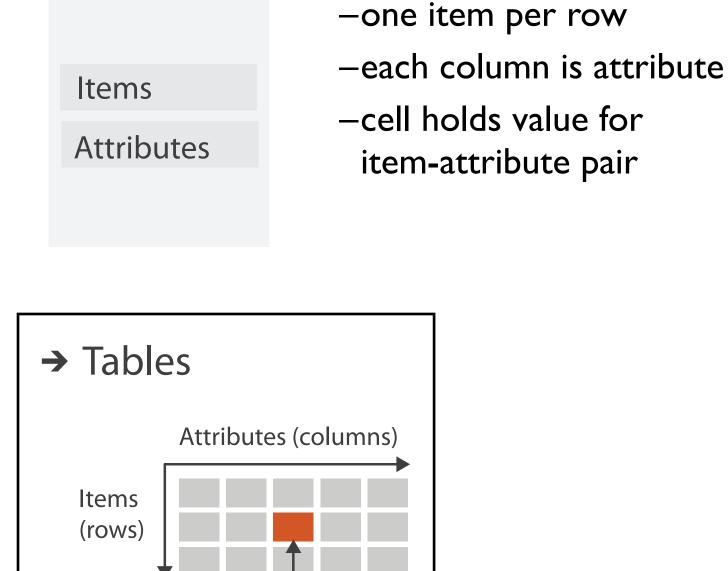
60

Tables

attributes: name, a

Name	Age	Shirt Size	Favorite Fruit
Amy	8	S	Apple
Basil	7	S	Pear
Clara	9	М	Durian
Desmond	13	L	Elderberry
Ernest	12	L	Peach
Fanny	10	S	Lychee
George	9	М	Orange
Hector	8	L	Loquat
Ida	10	М	Pear
Amy	12	М	Orange

item: person



Cell containing value

• flat table

attributes: name, age, shirt size, fave fruit

attributes: name, age, shirt size, fave fruit

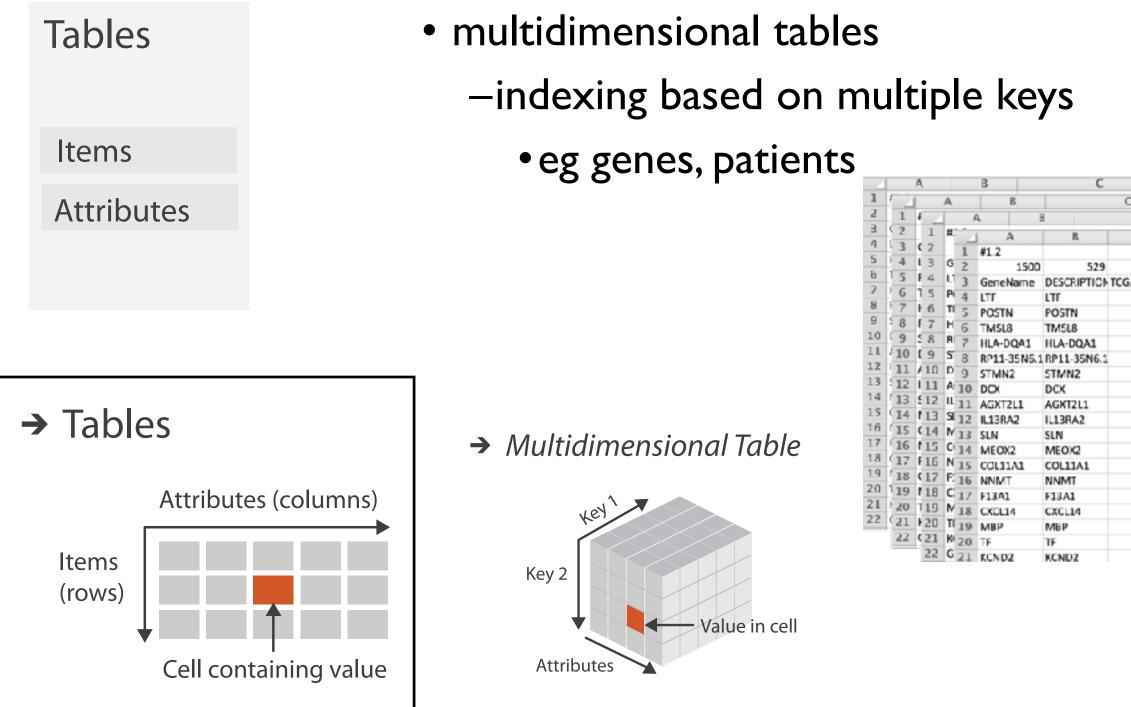
Tables	 flat table 			,		,
	-one item per row				K	
	–each column is attribute	ID	Name	Age	Shirt Size	Favorite Fruit
Items	-cell holds value for	1	Amy	8	S	Apple
Attributes	item-attribute pair	2	Basil	7	S	Pear
	-unique key	3	Clara	9	М	Durian
	(could be implicit)	4	Desmond	13	L	Elderberry
		5	Ernest	12	L	Peach
		6	Fanny	10	S	Lychee
→ Tables		7	George	9	М	Orange
Attribute	es (columns)	8	Hector	8	L	Loquat
Attributes (columns) Items (rows) Cell containing value			Ida	10	М	Pear
			Amy	12	М	Orange
						~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~
			tem: persor	ו		

Α	В	С	S	Т	U
Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
3	10/14/06	5-Low	Large Box	0.8	10/21/06
6	2/21/08	4-Not Specified	Small Pack	0.55	2/22/08
32	7/16/07	2-High	Small Pack	0.79	7/17/07
32	7/16/07	2-High	Jumbo Box	0.72	7/17/07
32	7/16/07	2-High	Medium Box	0.6	7/18/07
32	7/16/07	2-High	Medium Box	0.65	7/18/07
35	10/23/07	4-Not Specified	Wrap Bag	0.52	10/24/07
35	10/23/07	4-Not Specified	Small Box	0.58	10/25/07
36	11/3/07	1-Urgent	Small Box	0.55	11/3/07
65	3/18/07	1-Urgent	Small Pack	0.49	3/19/07
66	1/20/05	5-Low	Wrap Bag	0.56	1/20/05
69	6/4/05	4-Not Specified	Small Pack	0.44	6/6/05
69	6/4/05	4-Not Specified	Wrap Bag	0.6	6/6/05
70	12/18/06	5-Low	Small Box	0.59	12/23/06
70	12/18/06	5-Low	Wrap Bag	0.82	12/23/06
96	4/17/05	2-High	Small Box	0.55	4/19/05
97	1/29/06	3-Medium	Small Box	0.38	1/30/06
129	11/19/08	5-Low	Small Box	0.37	11/28/08
130	5/8/08	2-High	Small Box	0.37	5/9/08
130	5/8/08	2-High	Medium Box	0.38	5/10/08
130	5/8/08	2-High	Small Box	0.6	5/11/08
132	6/11/06	3-Medium	Medium Box	0.6	6/12/06
132	6/11/06	3-Medium	Jumbo Box	0.69	6/14/06
134	5/1/08	4-Not Specified	Large Box	0.82	5/3/08
135	10/21/07	4-Not Specified	Small Pack	0.64	10/23/07
166	9/12/07	2-High	Small Box	0.55	9/14/07
193		1-Urgent	Medium Box	0.57	8/10/06
194	4/5/08	3-Medium	Wrap Bag	0.42	4/7/08

	Α	В	С	S	Т	U
	Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
	3	10/14/06	5-Low	Large Box	0.8	10/21/06
	6	2/21/08	4-Not Specified	Small Pack	0.55	2/22/08
	32	7/16/07	2-High	Small Pack	0.79	7/17/07
	32	7/16/07	2-High	Jumbo Box	0.72	7/17/07
	32	7/16/07	2-High	Medium Box	0.6	7/18/07
item	32	7/16/07	2-High	Medium Box	0.65	7/18/07
	35		4-Not Specified	wrap Bag	0.52	10/24/07
	35	10/23/07	4-Not Specified	Small Box	0.58	10/25/07
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	69	6/4/05	4-Not Specified	Small Pack	0.44	6/6/05
	69	6/4/05	4-Not Specified	Wrap Bag	0.6	6/6/05
	70	12/18/06	5-Low	Small Box	0.59	12/23/06
	70	12/18/06	5-Low	Wrap Bag	0.82	12/23/06
	96	4/17/05	2-High	Small Box	0.55	4/19/05
	97	1/29/06	3-Medium	Small Box	0.38	1/30/06
	129	11/19/08	5-Low	Small Box	0.37	11/28/08
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	130	5/8/08	2-High	Medium Box	0.38	5/10/08
	130	5/8/08	2-High	Small Box	0.6	5/11/08
	132	6/11/06	3-Medium	Medium Box	0.6	6/12/06
	132	6/11/06	3-Medium	Jumbo Box	0.69	6/14/06
	134	5/1/08	4-Not Specified	Large Box	0.82	5/3/08
	135	10/21/07	4-Not Specified	Small Pack	0.64	10/23/07
	166	9/12/07	2-High	Small Box	0.55	9/14/07
	193	8/8/06	1-Urgent	Medium Box	0.57	8/10/06
	194	4/5/08	3-Medium	Wrap Bag	0.42	4/7/08

	Α	В	С	5	Т	U
	Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
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	32	7/16/07	2-High	Small Pack	0.79	7/17/07
	32	7/16/07	2-High	Jumbo Box	0.72	7/17/07
	32	7/16/07	2-Hiah	Medium Box	0.6	7/18/07
item	32	7/16/07	2-High	Medium Box	0.65	7/18/07
	35			wrap Bag	0.52	
	35	10/23/07	4-Not Specified	Small Box	0.58	10/25/07
	36	11/3/07	1-Urgent	Small Box	0.55	11/3/07
	65	3/18/07	1-Urgent	Small Pack	0.49	3/19/07
	66	1/20/05	5-Low	Wrap Bag	0.56	1/20/05
	69	6/4/05	4-Not Specified	Small Pack	0.44	6/6/05
	69	6/4/05	4-Not Specified	Wrap Bag	0.6	6/6/05
	70	12/18/06	5-Low	Small Box	0.59	12/23/06
	70	12/18/06	5-Low	Wrap Bag	0.82	12/23/06
	96	4/17/05	2-High		<b>DUTE</b> 0.55	4/19/05
	97	1/29/06	3-Medium	Small Box	0.38	1/30/06
	129	11/19/08	5-Low	Small Box	0.37	11/28/08
	130	5/8/08	2-High	Small Box	0.37	5/9/08
	130	5/8/08	2-High	Medium Box	0.38	5/10/08
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	132	6/11/06	3-Medium	Jumbo Box	0.69	6/14/06
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	193	8/8/06	1-Urgent	Medium Box	0.57	8/10/06
	194	4/5/08	3-Medium	Wrap Bag	0.42	4/7/08

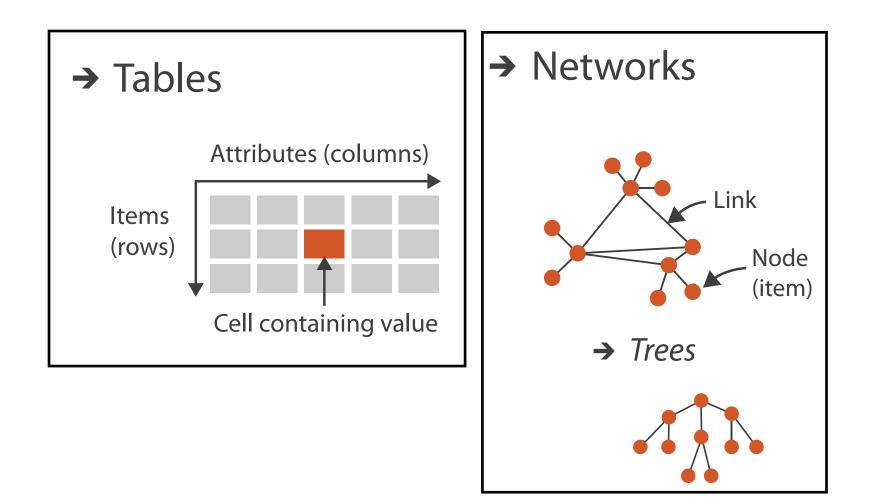
	Α	В	С	5	́Т ́	U
	Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
	3	10/14/06	5-Low	Large Box	0.8	10/21/06
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item	32	7/16/07	2-High	Medium Box	0.65	7/18/07
	35		4-Not Specified	wrap Bag	0.52	10/24/07
	35	10/23/07	4-Not Specified	Small Box	0.58	10/25/07
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	65	3/18/07	1-Urgent	Small Pack	0.49	3/19/07
	66	1/20/05	5-Low	Wrap Bag	0.56	1/20/05
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	69	6/4/05	4-Not Specified	Wrap Bag	0.6	6/6/05
	70	12/18/06	5-Low	Small Box	0.59	12/23/06
	70	12/18/06	5-Low	Wrap Bag	0.82	12/23/06
	96	4/17/05	2-High	Small Box att		
	97	1/29/06	3-Medium	Small Box	0.38	1/30/06
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	130	5/8/08	2-High	Small Box	0.37	5/9/08
	130	5/8/08	2-High	Medium Box	0.38	5/10/08
	130	5/8/08	2-High	Small Box	0.6	5/11/08
	132	6/11/06	3-Medium	Medium Box	0.6	6/12/06
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	135	10/21/07	4-Not Specified	Small Pack	0.64	10/23/07
	166	9/12/07	2-High	Small Box	0.55	9/14/07
	193	8/8/06	1-Urgent	Medium Box	0.57	8/10/06
	194	4/5/08	3-Medium	Wrap Bag	0.42	4/7/08



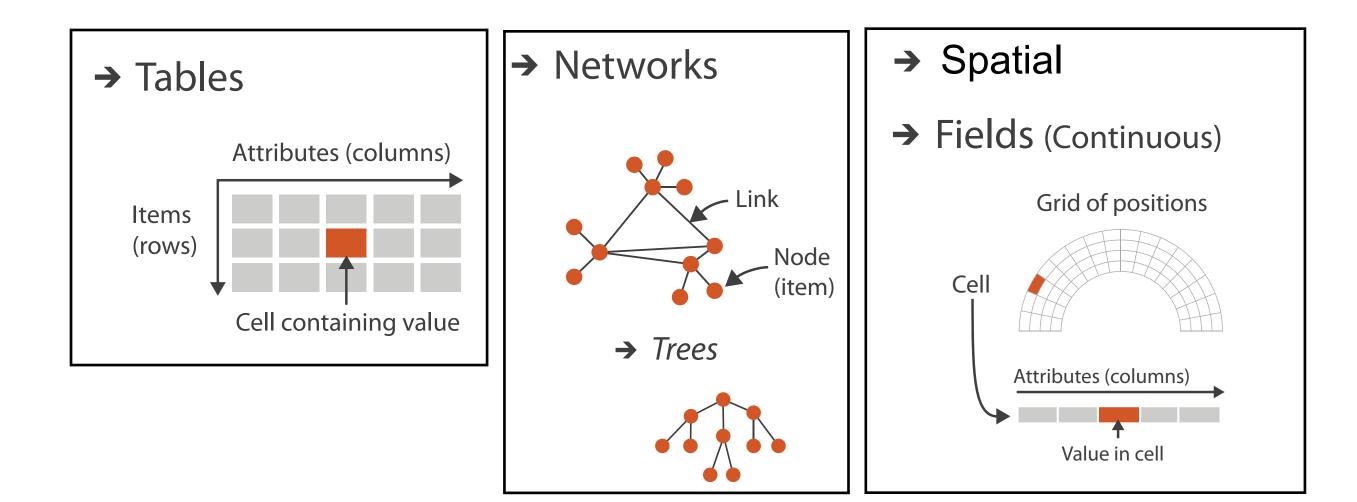
		D			E				
С			D			E			
C			D			E			
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GA-02-0000	1-01C-01R-01	77-01	TCGA-02-0003-01	A-01R-0	0177-01	TCGA-02-0004-	01A-01R-02	98-01	1
	-1.2657	28057		2.37	7012066		4.1239	979585	í -
	2.6624	11805		3.932	2400324		5.031	585377	1
	-3.0822	17838		-2.243	3148513		-0.023	313581	
	-1.7390	64398		4.573	7962344		3.1277	744964	T.
	-3.3463	52968		-2.899	\$400157		-3.4730	035067	
	-2.5785	11106		-3.05:	1605144		-1.7298	92386	T.
	-2.260	78976		-2.529	9795801		-2.8449	66278	T.
	-2.6394	93611		-3.113	3204863		-0.4039	975027	
	-2.935	96915		-1.873	8600916		2.9762	256911	T.
	-2.4657	18221		-2.208	8406749		1.0258	827904	T.
	-2.3950	54056		-1.063	2676046		1.7833	235317	Ť.
	1.2119	34832		-0.399	9392588		4,733(	608974	T.
	0.7037	45154		0.66	4082419		3.0690	030715	t.
	-0.2240	94042		2.223	2197544		1.1713	354775	T.
	-8.18	09694		-1.395	5056071		2.5695	540659	Ť.
	-1.9053	90566		-2.031	626447		-2.935	744906	t.
	-4.3341	23292		-4.680	0680246		-2.9757	788866	t.
	-1.7776	92395		-2.100	352021		-1.9963	06032	Ť

Tables	Networks & Trees
ltems	Items (nodes)
Attributes	Links
	Attributes

 network/graph -nodes (vertices) connected by links (edges) -tree is special case: no cycles • often have roots and are directed



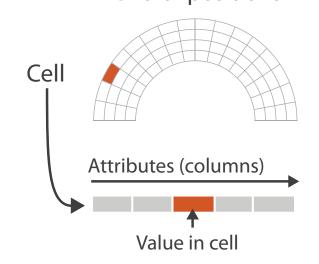
Tables	Networks & Trees	Fields
ltems	Items (nodes)	Grids
Attributes	Links	Positions
	Attributes	Attributes

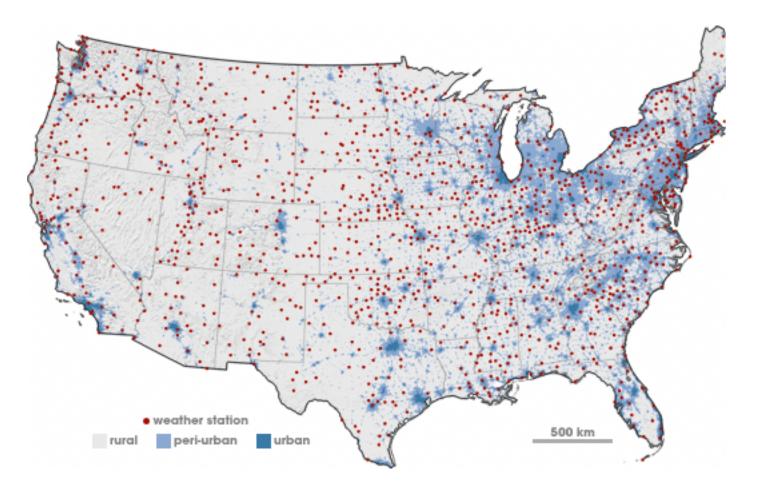


## Spatial fields

- attribute values associated w/ cells
- cell contains value from continuous domain
  - eg temperature, pressure, wind velocity
- measured or simulated
  - → Spatial
  - → Fields (Continuous)







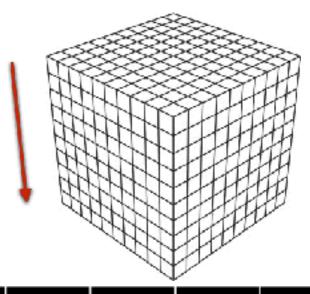
## Spatial fields

- attribute values associated w/ cells
- cell contains value from continuous domain
  - eg temperature, pressure, wind velocity
- measured or simulated
- major concerns
  - sampling:
     where attributes are measured
  - interpolation:

how to model attributes elsewhere

- grid types

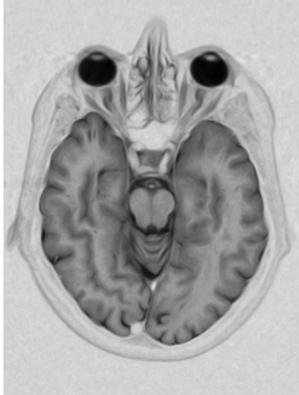




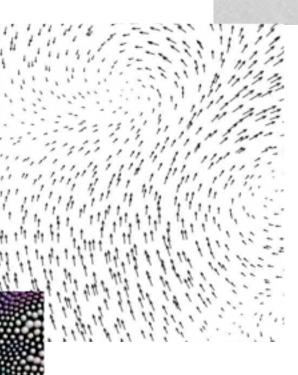
S			

## Spatial fields

- attribute values associated w/ cells
- cell contains value from continuous domain
  - eg temperature, pressure, wind velocity
- measured or simulated
- major concerns
  - sampling:
     where attributes are measured
  - interpolation:
     how to model attributes elsewhere
  - -grid types
- major divisions
  - attributes per cell: scalar (1), vector (2), tensor (many)



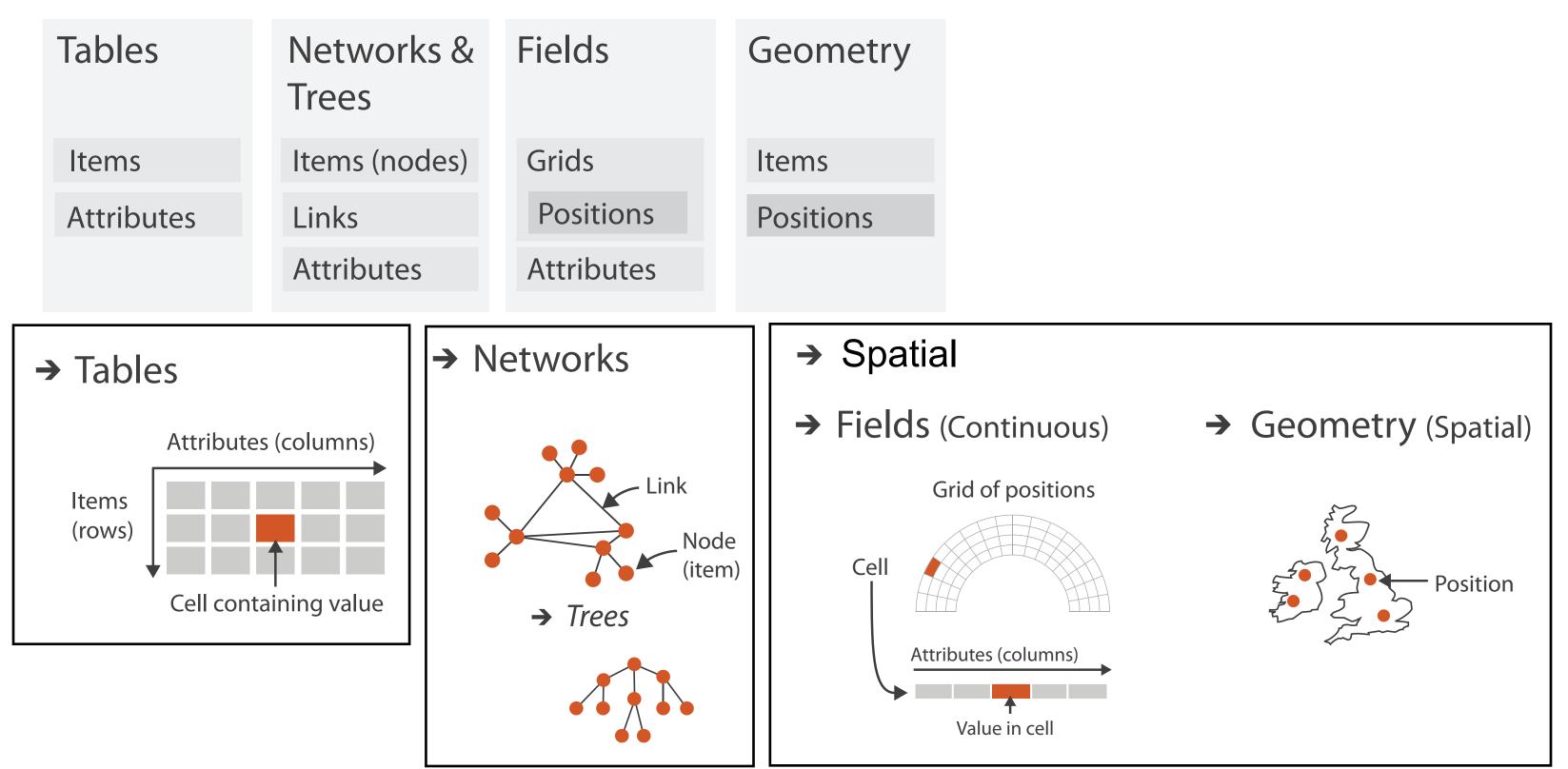
### scalar



vector

tensor

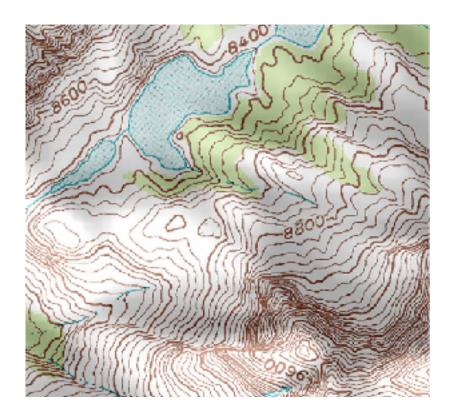
### Dataset types



## Geometry

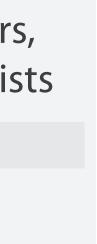
- shape of items
- explicit spatial positions / regions
   points, lines, curves, surfaces, volumes
- boundary between computer graphics and visualization
  - -graphics: geometry taken as given
  - -vis: geometry is result of a design decision





### Dataset types

Tables	Networks & Trees	Fields	Geometry	Clusters, Sets, Lists
Items	Items (nodes)	Grids	Items	Items
Attributes	Links	Positions	Positions	
	Attributes	Attributes		
Items (rows)	s (columns) aining value	Ietworks	de m) Cell	Continuous) rid of positions

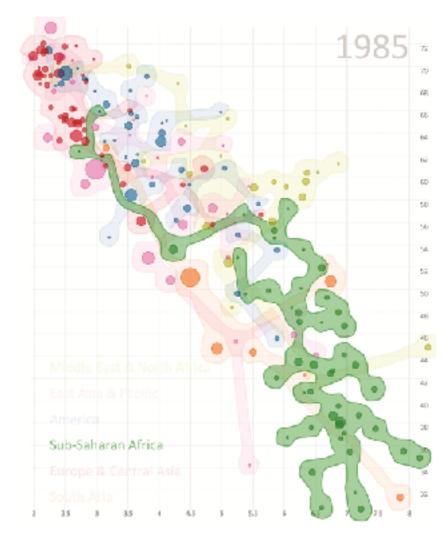


### → Geometry (Spatial)



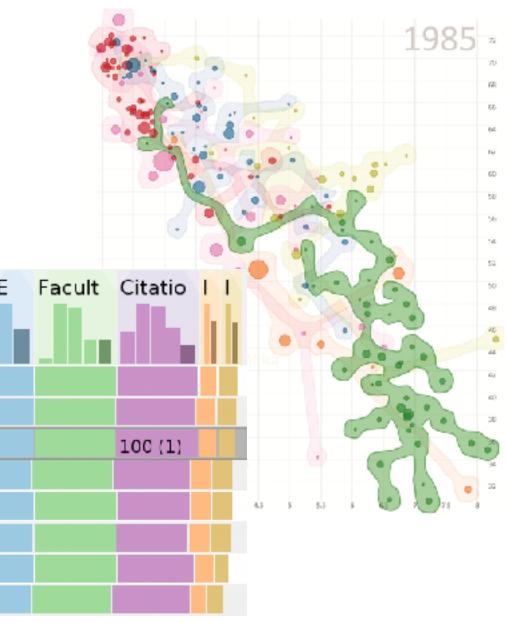
how we group items

- how we group items
- sets
  - unique items, unordered

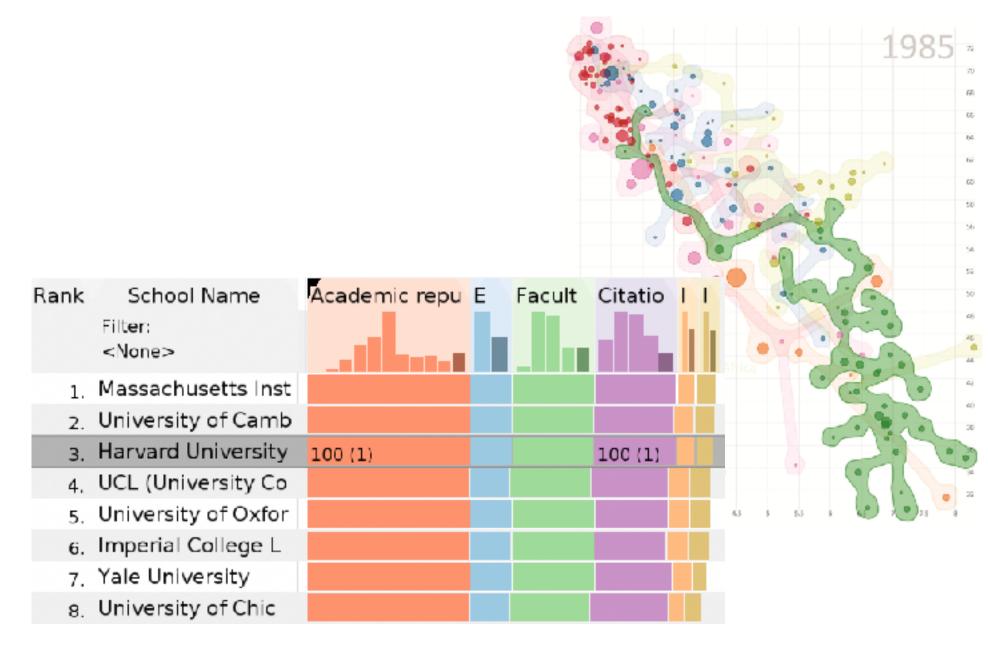


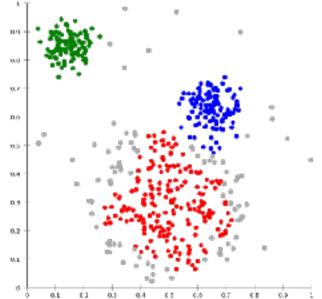
- how we group items
- sets
  - unique items, unordered
- lists
  - ordered, duplicates possible

Rank	School Name Filter: <none></none>	Academic repu	E
1.	Massachusetts Inst		
2.	University of Camb		
З.	Harvard University	100(1)	
4.	UCL (University Co		
5.	University of Oxfor		
6.	Imperial College L		
7.	Yale University		
8.	University of Chic		



- how we group items
- sets
  - unique items, unordered
- lists
  - -ordered, duplicates possible
- clusters
  - -groups of similar items





### Dataset and data types

### Data and Dataset Types

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

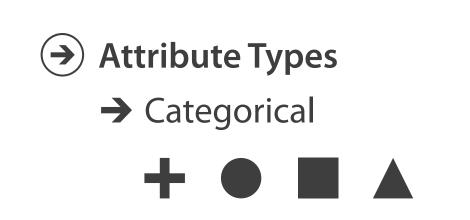
→ Items → Attribu	tes $\rightarrow$ Links	→ Positions	→ Grid
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### ds

## Attribute types

- which classes of values & measurements?
- categorical (nominal)
  - -compare equality
  - -no implicit ordering
- ordered
  - -ordinal
    - less/greater than defined
  - quantitative
    - meaningful magnitude
    - arithmetic possible



### ➔ Ordered

### → Ordinal



→ Quantitative

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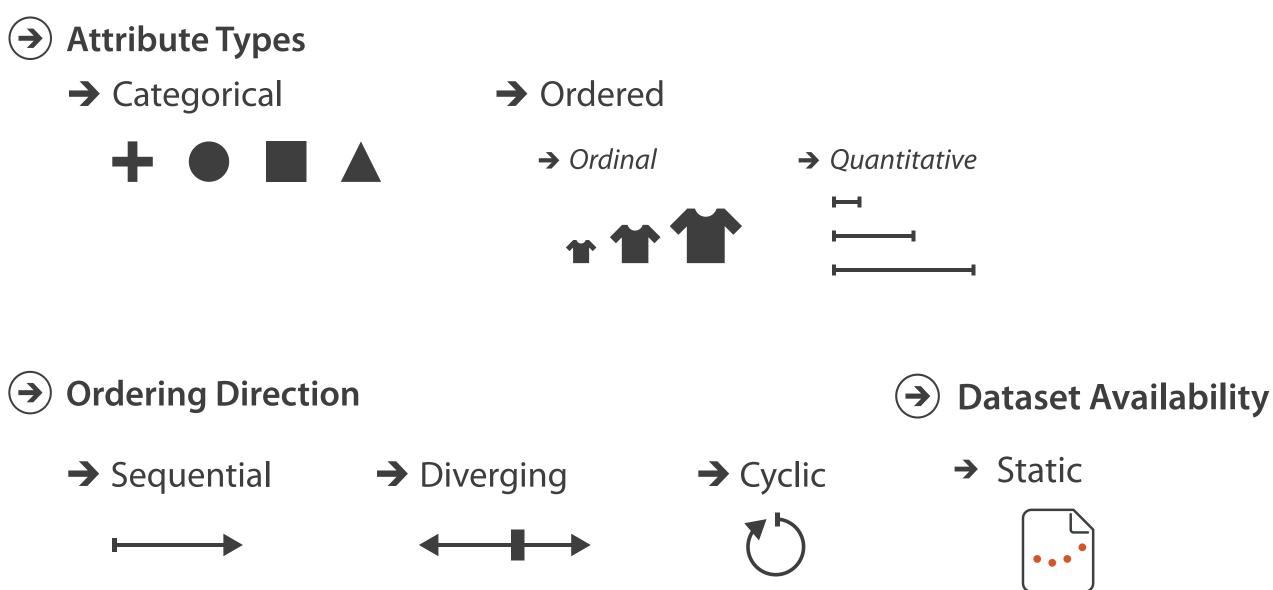
### Table

Α	В	С	S	Т	U
Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
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32	7/16/07	2-High	Medium Box	0.65	7/18/07
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69	6/4/05	4-Not Specified	Wrap Bag	0.6	6/6/05
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70	12/18/06	5-Low	Wrap Bag	0.82	12/23/06
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97	1/29/06	3-Medium	Small Box	0.38	1/30/06
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130	5/8/08	2-High	Medium Box	0.38	5/10/08
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132	6/11/06	3-Medium	Jumbo Box	0.69	6/14/06
134	5/1/08	4-Not Specified	Large Box	0.82	5/3/08
135	10/21/07	4-Not Specified	Small Pack	0.64	10/23/07
166	9/12/07	2-High	Small Box	0.55	9/14/07
193	8/8/06	1-Urgent	Medium Box	0.57	8/10/06
194	4/5/08	3-Medium	Wrap Bag	0.42	4/7/08

### categorical ordinal quantitative

A	В	C	2	5	
Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
3	10/14/06	5-Low	Large Box	0.8	10/21/06
6	2/21/08	4-Not Specified	Small Pack	0.55	2/22/08
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32	7/16/07	2-High	Jumbo Box	0.72	7/17/07
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70	12/18/06	5-Low	Wrap Bag	0.82	12/23/06
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130	5/8/08	2-High	Small Box	0.6	5/11/08
132	6/11/06	3-Medium	Medium Box	0.6	6/12/06
132	6/11/06	3-Medium	Jumbo Box	0.69	6/14/06
134	5/1/08	4-Not Specified	Large Box	0.82	5/3/08
135	10/21/07	4-Not Specified	Small Pack	0.64	10/23/07
166	9/12/07	2-High	Small Box	0.55	9/14/07
193	8/8/06	1-Urgent	Medium Box	0.57	8/10/06

## Other data concerns







### Data abstraction: Three operations

- translate from domain-specific language to generic visualization language
- identify dataset type(s), attribute types
- identify cardinality
  - -how many items in the dataset?
  - -what is cardinality of each attribute?
    - number of levels for categorical data
    - range for quantitative data
- consider whether to transform data
  - -guided by understanding of task

### Data vs conceptual models

- data model
  - -mathematical abstraction
    - sets with operations, eg floats with * / +
    - variable data types in programming languages
- conceptual model
  - -mental construction (semantics)
  - -supports reasoning
  - -typically based on understanding of tasks [stay tuned!]
- data abstraction process relies on conceptual model
   for transforming data if needed

- data model: floats
  - -32.52, 54.06, -14.35, ...

- data model: floats
   32.52, 54.06, -14.35, ...
- conceptual model
  - -temperature

- data model: floats
   32.52, 54.06, -14.35, ...
- conceptual model
  - -temperature
- multiple possible data abstractions

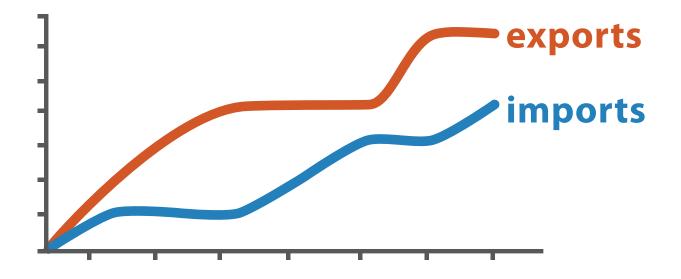
- data model: floats
   32.52, 54.06, -14.35, ...
- conceptual model
  - -temperature
- multiple possible data abstractions
  - continuous to 2 significant figures: quantitative
    - task: forecasting the weather

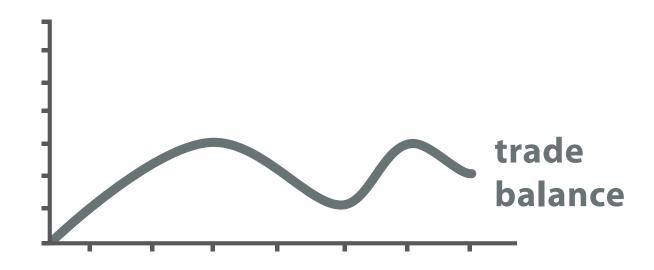
- data model: floats
   32.52, 54.06, -14.35, ...
- conceptual model
  - -temperature
- multiple possible data abstractions
  - continuous to 2 significant figures: quantitative
    - task: forecasting the weather
  - -hot, warm, cold: ordinal
    - task: deciding if bath water is ready

- data model: floats
   32.52, 54.06, -14.35, ...
- conceptual model
  - -temperature
- multiple possible data abstractions
  - continuous to 2 significant figures: quantitative
    - task: forecasting the weather
  - -hot, warm, cold: ordinal
    - task: deciding if bath water is ready
  - -above freezing, below freezing: categorical
    - task: decide if I should leave the house today

### **Derived** attributes

- derived attribute: compute from originals
  - simple change of type
  - -acquire additional data
  - complex transformation





trade balance = exports – imports

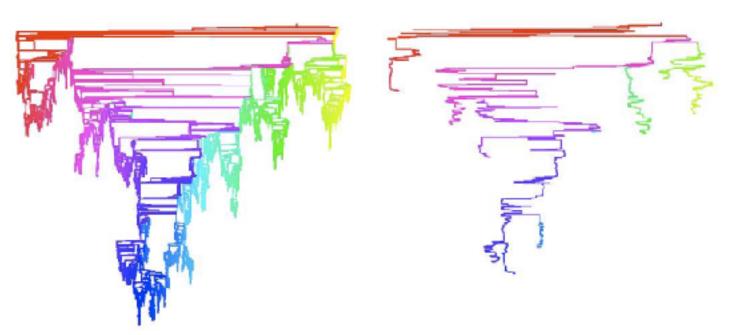
**Derived** Data

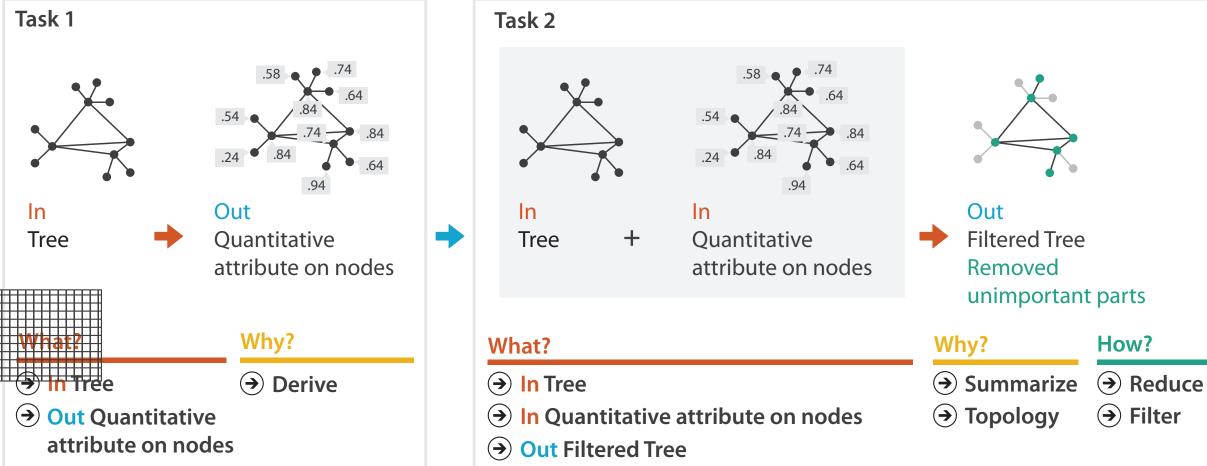
### **Original Data**

### Analysis example: Derive one attribute

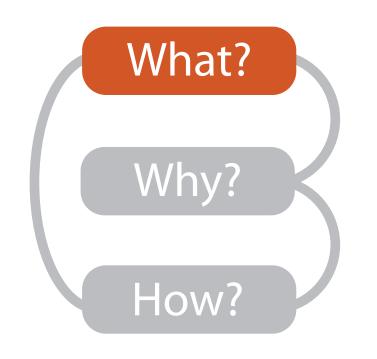
- Strahler number
  - centrality metric for trees/networks
  - derived quantitative attribute
  - draw top 5K of 500K for good skeleton

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









What?					
	D	atasets			At
<ul> <li>→ Data Type</li> <li>→ Items</li> <li>→ Data and I</li> </ul>	s → Attributes Dataset Types	→ Links	→ Positions	→ Grids	<ul> <li>→ Attribut</li> <li>→ Cate</li> <li>↓</li> </ul>
Tables Items Attributes	Networks & Trees Items (nodes) Links Attributes	Fields Grids Positions Attributes	Geometry Items Positions	Clusters, Sets, Lists Items	<ul> <li>→ Orde</li> <li>→ Orde</li> <li>→ Orde</li> <li>→ Que</li> </ul>
Dataset Ty → Tables Att Items (rows)		Vetworks	→ Fields (Co Grid Node	ontinuous) of positions	→ Orderin → Seque → Diverg
→ Multidi	Value in cell		item)	tes (columns)	→ Cyclic
→ Geome	try (Spatial)		<ul> <li>→ Dataset A</li> <li>→ Static</li> <li>→</li> </ul>	vailability	→ Dynamic

### Attributes

ute Types

tegorical



dered

rdinal



uantitative

### ing Direction

uential



erging



ic





## Further reading, full Ch 2

- Readings in Information Visualization: Using Vision To Think, Chapter 1. Stuart K. Card, Jock Mackinlay, and Ben Shneiderman. Morgan Kaufmann, 1999.
- Rethinking Visualization: A High-Level Taxonomy. InfoVis 2004, p 151-158, 2004.
- The Eyes Have It: A Task by Data Type Taxonomy for Information Visualizations Ben Shneiderman, Proc. **1996** IEEE Visual Languages
- Data Visualization: Principles and Practice, 2nd ed. Alexandru Telea, CRC Press, 2014.
- Interactive Data Visualization: Foundations, Techniques, and Applications, 2nd ed. Matthew O. Ward, Georges Grinstein, Daniel Keim. CRC Press, 2015.
- The Visualization Handbook. Charles Hansen and Chris Johnson, eds. Academic Press, 2004.
- Visualization Toolkit: An Object-Oriented Approach to 3D Graphics, 4th ed. Will Schroeder, Ken Martin, and Bill Lorensen. Kitware 2006.
- The Structure of the Information Visualization Design Space. Stuart Card and Jock Mackinlay, Proc. InfoVis 97.
- Polaris: A System for Query, Analysis and Visualization of Multi-dimensional Relational Databases (extended paper) Chris Stolte, Diane Tang and Pat Hanrahan. IEEE TVCG 8(1):52-65 2002. • Visualization of Time-Oriented Data. Wolfgang Aigner, Silvia Miksch, Heidrun Schumann, Chris Tominski.
- Springer 2011.

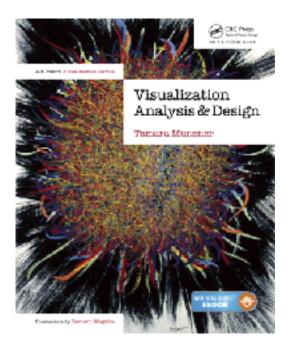
## Visualization Analysis & Design

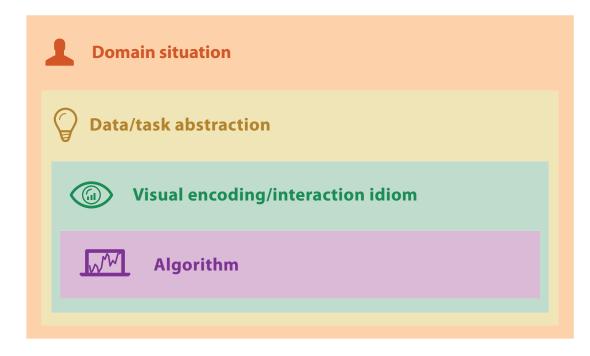
## Task Abstraction (Ch 3)

### Tamara Munzner

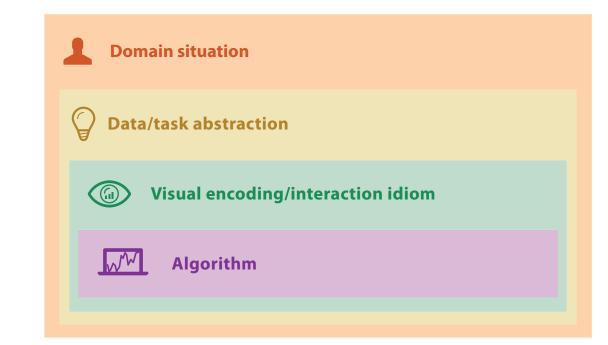
Department of Computer Science University of British Columbia

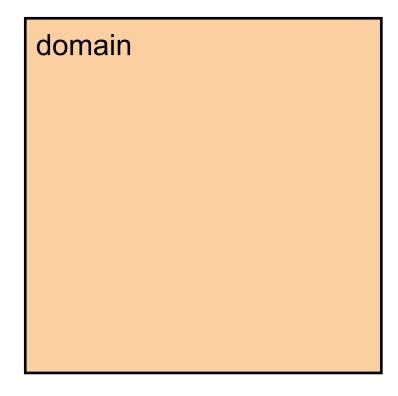
<u>@tamaramunzner</u>



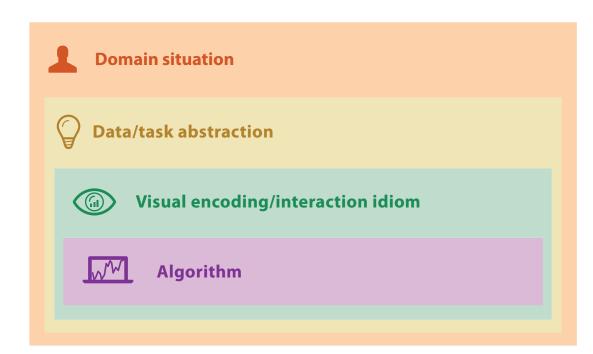


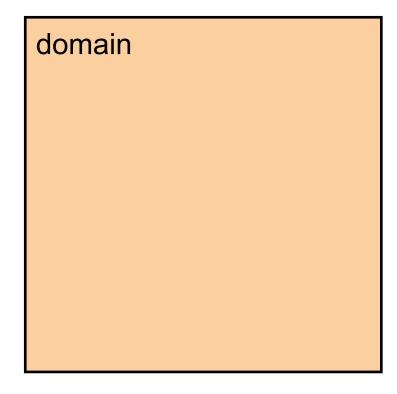
• domain characterization: details of application domain



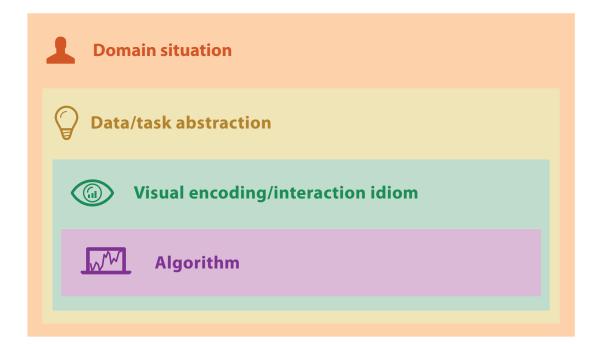


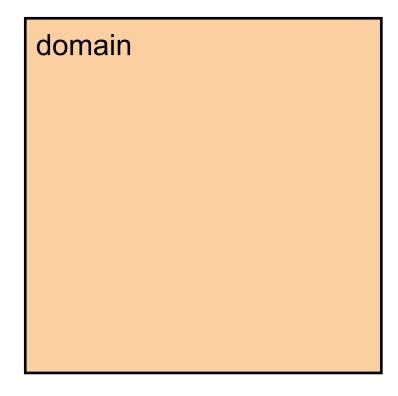
- domain characterization: details of application domain
  - -group of users, target domain, their questions & data
    - varies wildly by domain
    - must be specific enough to get traction



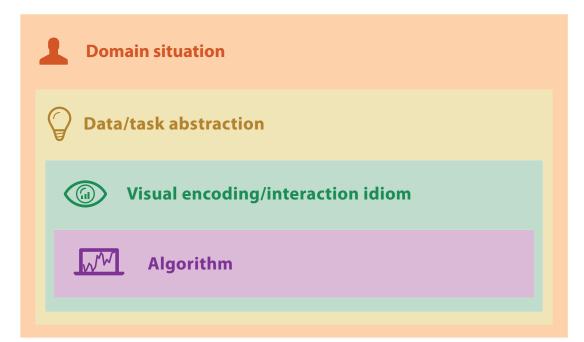


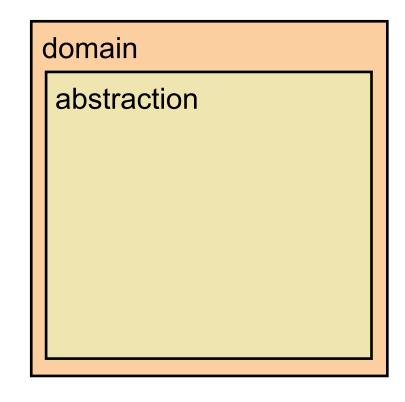
- domain characterization: details of application domain
  - -group of users, target domain, their questions & data
    - varies wildly by domain
    - must be specific enough to get traction
  - -domain questions/problems
    - break down into simpler abstract tasks



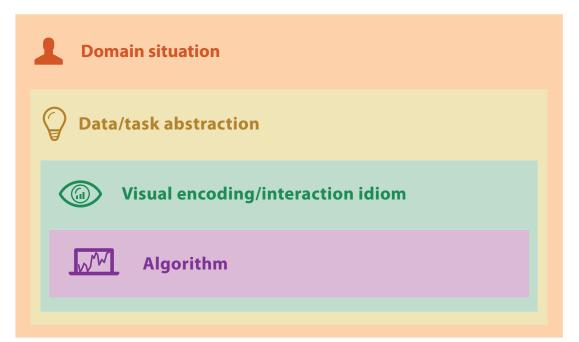


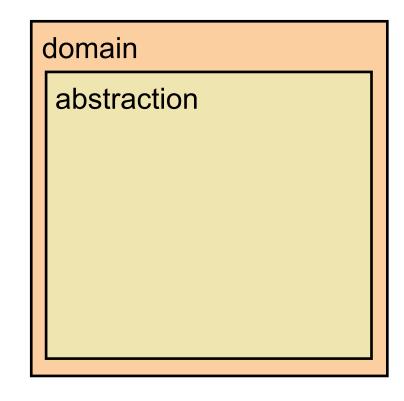
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  - -domain questions/problems
    - break down into simpler abstract tasks
- abstraction: data & task
  - -map what and why into generalized terms





- domain characterization: details of application domain
  - -group of users, target domain, their questions & data
    - varies wildly by domain
    - must be specific enough to get traction
  - -domain questions/problems
    - break down into simpler abstract tasks
- 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





### Design process



Map Domain-Language Data Description to **Data Abstraction** 



Identify/Create Suitable Idiom/Technique

Identify/Create Suitable Algorithm

# Map Domain-Language Task

### Task abstraction: Actions and targets

• very high-level pattern

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

### Task abstraction: Actions and targets

• very high-level pattern

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- actions
  - -analyze
    - high-level choices
  - -search
    - find a known/unknown item
  - -query
    - find out about characteristics of item

### Task abstraction: Actions and targets

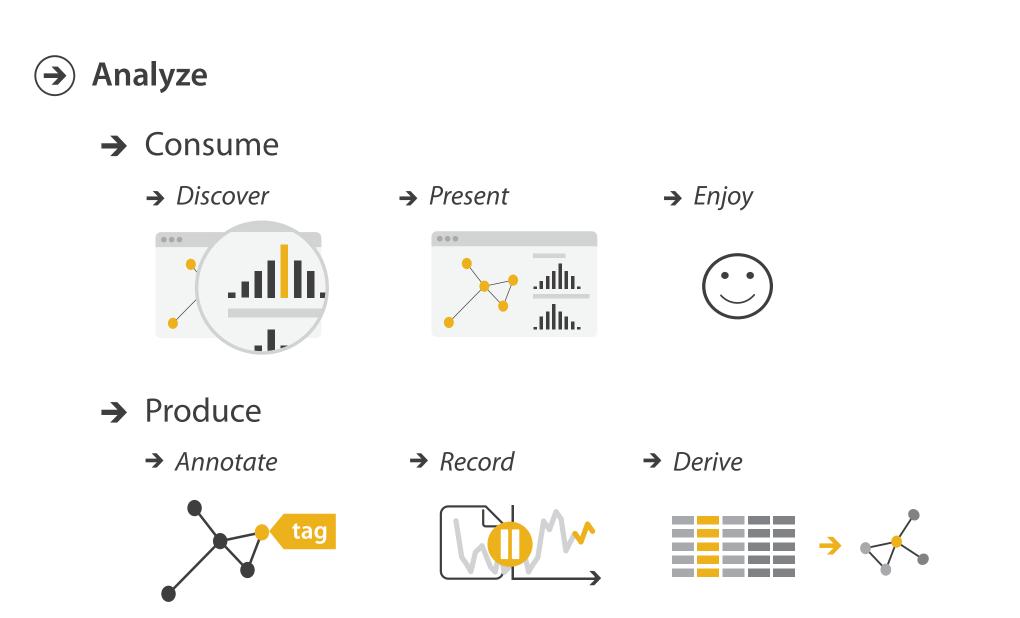
• very high-level pattern

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  - -browse topology

- actions
  - -analyze
    - high-level choices
  - -search
    - find a known/unknown item
  - -query
    - find out about characteristics of item
- targets
  - what is being acted on

## Actions: Analyze

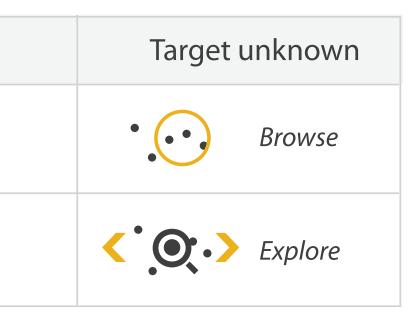
- consume
  - -discover vs present
    - classic split
    - aka explore vs explain
  - -enjoy
    - newcomer
    - aka casual, social
- produce
  - -annotate, record
  - -derive
    - crucial design choice



- what does user know?
  - target, location



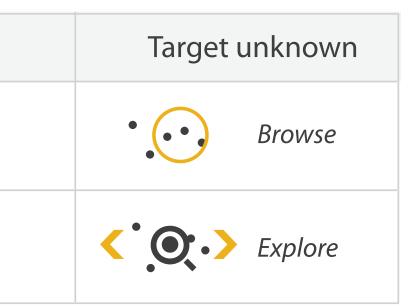
	Target known		
Location known	• • • Lookup		
Location unknown	<b>C O C</b> <i>Locate</i>		



- what does user know?
   target, location
- lookup
  - ex: word in dictionary
    - alphabetical order

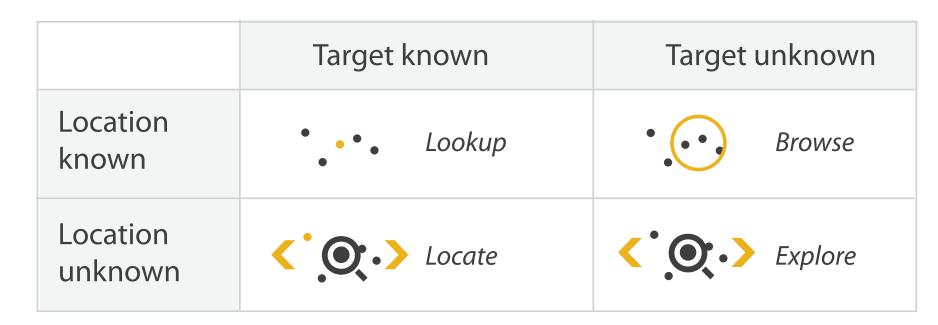


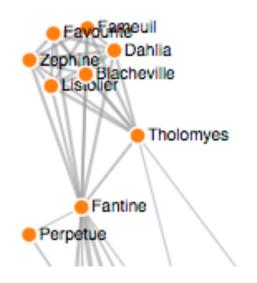
	Target known		
Location known	• • • •	Lookup	
Location unknown	< <u>(</u> )	Locate	



- what does user know?
   target, location
- lookup
  - ex: word in dictionary
    - alphabetical order
- locate
  - ex: keys in your house
  - ex: node in network

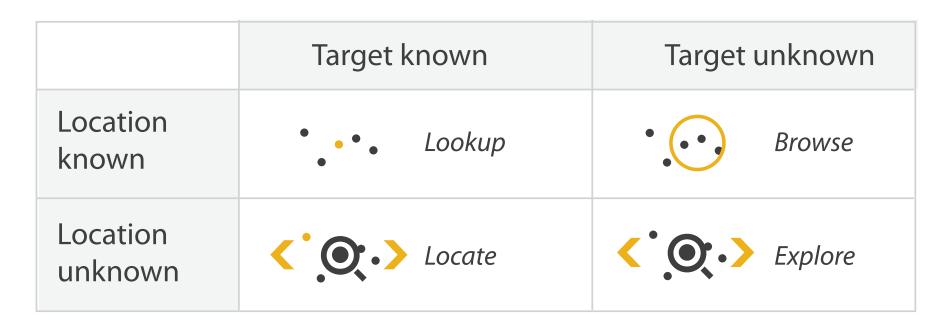


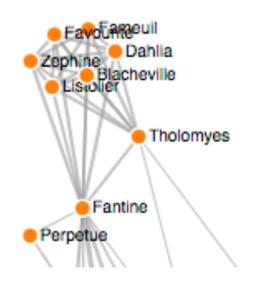




- what does user know?
   target, location
- lookup
  - ex: word in dictionary
    - alphabetical order
- locate
  - ex: keys in your house
  - ex: node in network
- browse
  - ex: books in bookstore

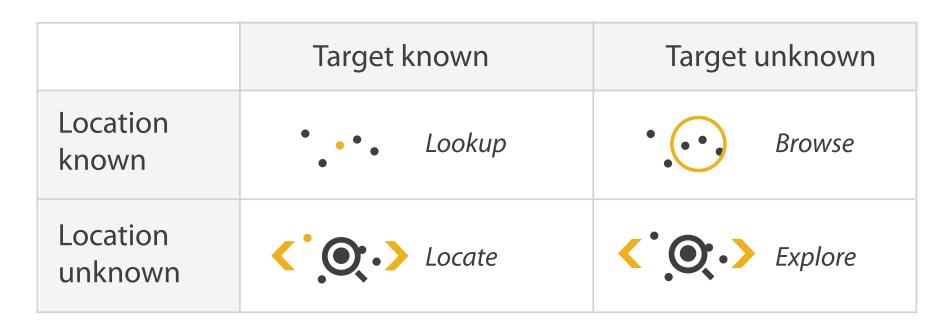


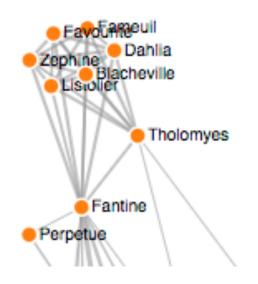




- what does user know?
   target, location
- lookup
  - ex: word in dictionary
    - alphabetical order
- locate
  - ex: keys in your house
  - ex: node in network
- browse
  - ex: books in bookstore
- explore
  - ex: find cool neighborhood in new city



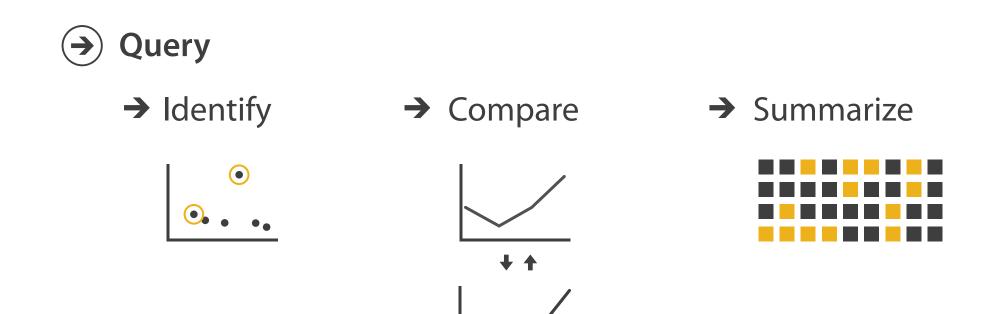




https://bl.ocks.org/heybignick/3faf257bbbbc7743bb72310d03b86ee8

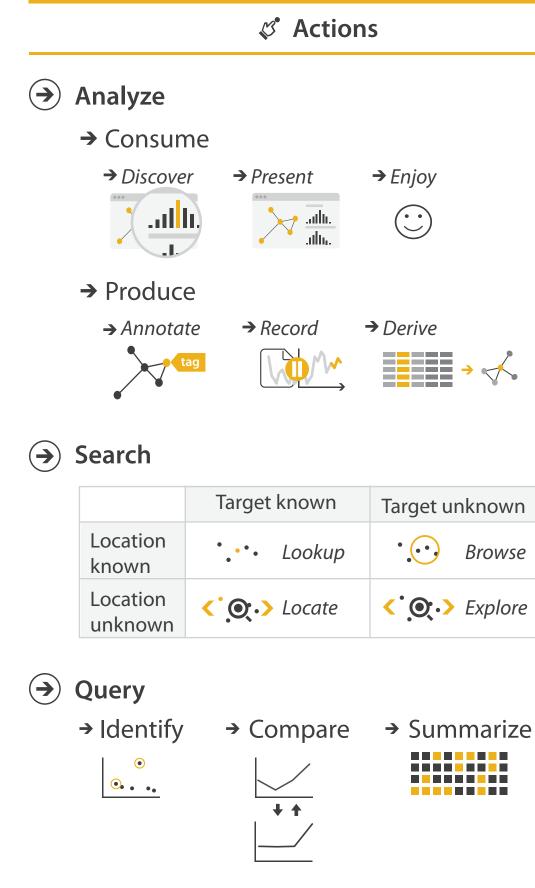
## Actions: Query

- how much of the data matters?
  - one: identify
  - some: compare
  - -all: summarize

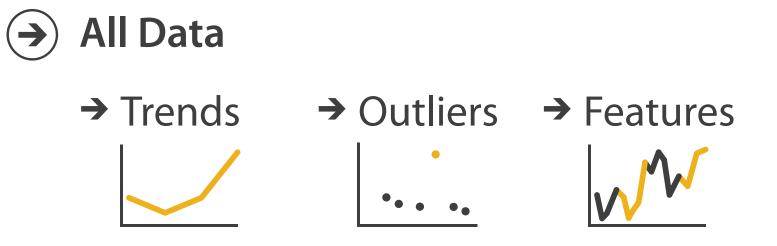


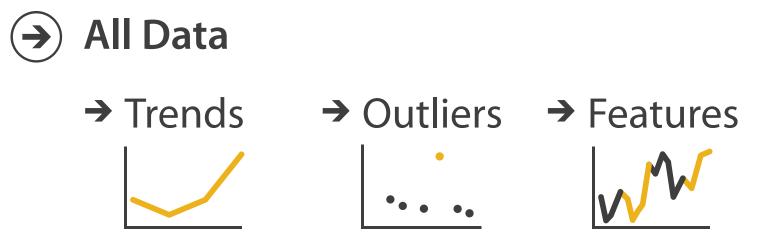
### Actions

- independent choices for each of these three levels
  - -analyze, search, query
  - -mix and match

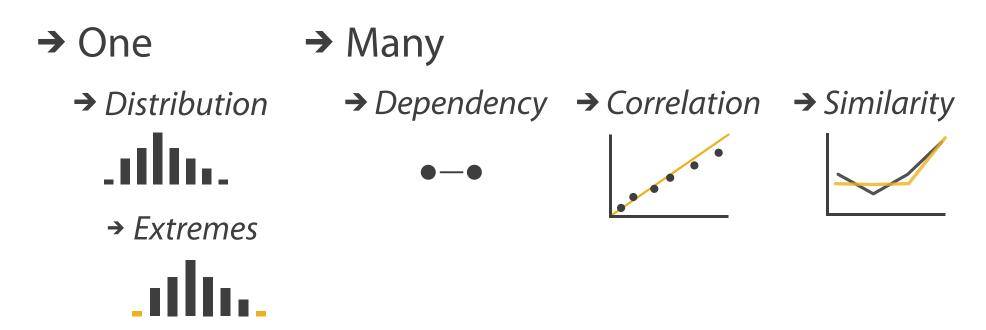






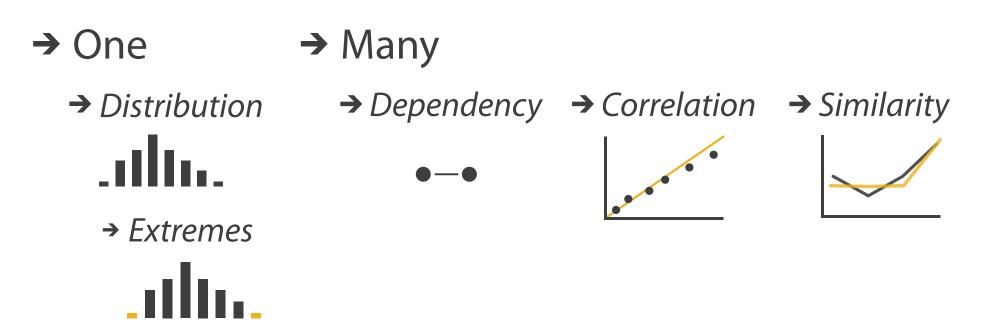


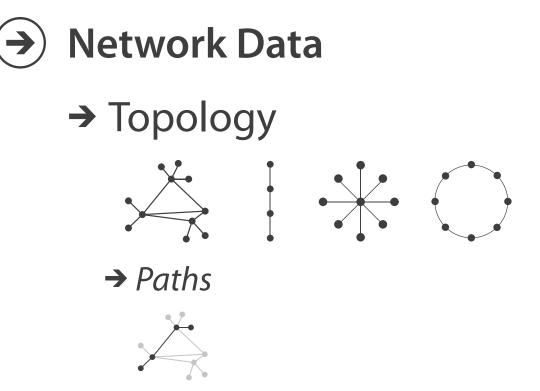
→ Attributes





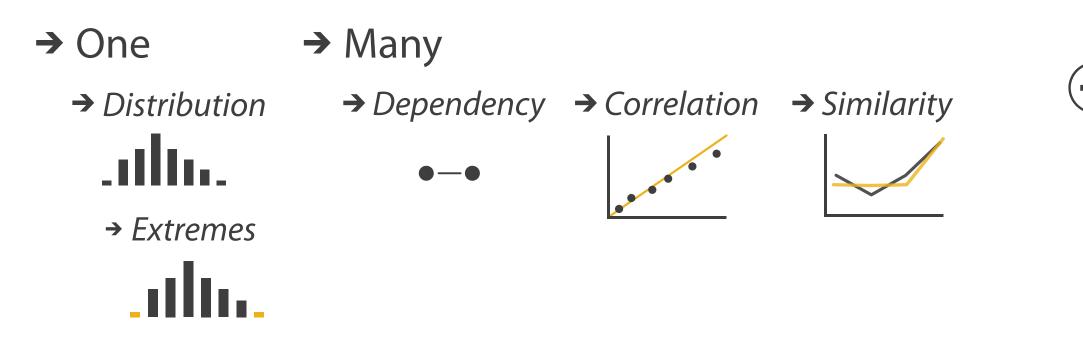
→ Attributes

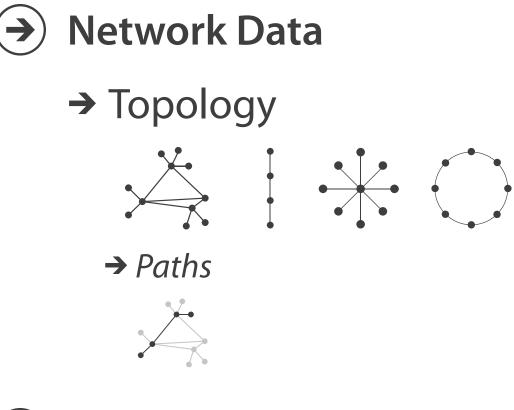


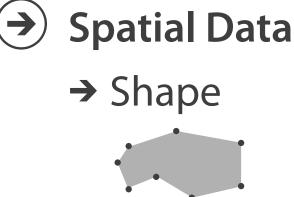




→ Attributes

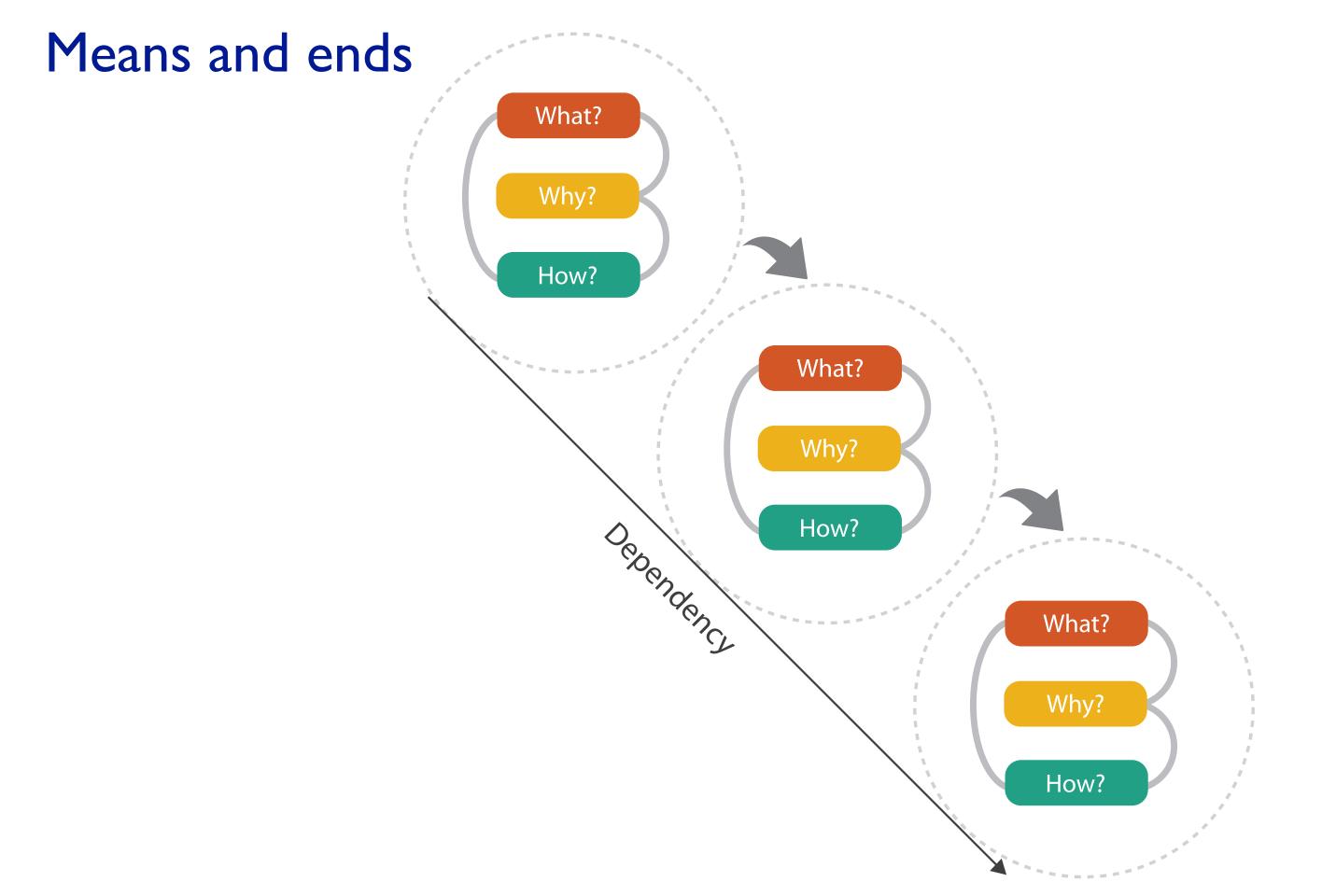


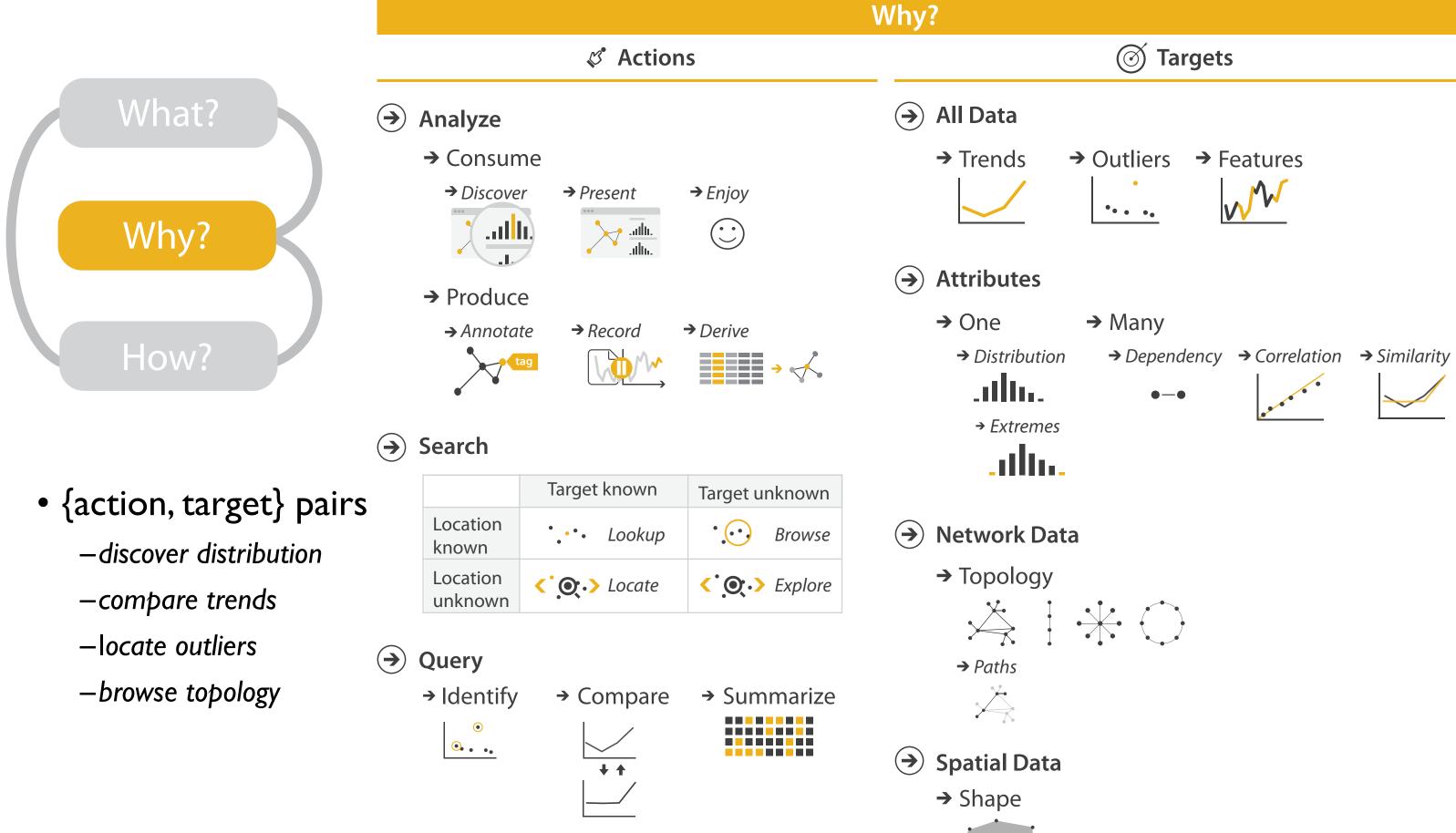




### Abstraction

- these {action, target} pairs are good starting point for vocabulary -but sometimes you'll need more precision!
- rule of thumb
  - systematically remove all domain jargon
- interplay: task and data abstraction
  - -need to use data abstraction within task abstraction
    - to specify your targets!
    - but task abstraction can lead you to transform the data
  - -iterate back and forth
    - first pass data, first pass task, second pass data, ...







## Further reading

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

## Further reading, full Ch 3

- A Multi-Level Typology of Abstract Visualization Tasks.. Matthew Brehmer and Tamara Munzner. IEEE Transactions on Visualization and Computer Graphics (Proc. InfoVis 13) 19:12 (2013), 2376-2385.
- A characterization of the scientific data analysis process. Rebecca R. Springmeyer, Meera M. Blattner, and Nelson M. Max. Proc. Vis 1992, p 235-252.
- Low-Level Components of Analytic Activity in Information Visualization. Robert Amar, James Eagan, and John Stasko. Proc. InfoVis 05, pp. |||-||7.
- Task taxonomy for graph visualization. Bongshin Lee, Catherine Plaisant, Cynthia Sims Parr, Jean-Daniel Fekete, and Nathalie Henry. Proc. BELIV 2006.
- Interactive Dynamics for Visual Analysis. Jeffrey Heer and Ben Shneiderman. Communications of the ACM, 55(4), pp. 45-54, 2012.
- What does the user want to see?: what do the data want to be? A. Johannes Pretorius and Jarke J. van Wijk. Information Visualization 8(3):153-166, 2009.
- Chapter I, Readings in Information Visualization: Using Vision to Think. Stuart Card, Jock Mackinlay, and Ben Shneiderman, Morgan Kaufmann 1999.
- An Operator Interaction Framework for Visualization Systems. Ed H. Chi and John T. Riedl. Proc. InfoVis 1998, p 63-70.
- Nominal, Ordinal, Interval, and Ratio Typologies are Misleading. Paul F. Velleman and Leland Wilkinson. The American Statistician 47(1):65-72, 1993.
- Rethinking Visualization: A High-Level Taxonomy. Melanie Tory and Torsten Möller, Proc. InfoVis 2004, pp. 151-158.
- SpaceTree: Supporting Exploration in Large Node Link Tree, Design Evolution and Empirical Evaluation. Catherine Plaisant, Jesse Grosjean, and Ben B. Bederson. Proc. InfoVis 2002.
- TreeJuxtaposer: Scalable Tree Comparison using Focus+Context with Guaranteed Visibility Tamara Munzner, Francois Guimbretiere, Serdar Tasiran, Li Zhang, and Yunhong Zhou. SIGGRAPH 2003.
- Feature detection in linked derived spaces. Chris Henze. Proc. Visualization (Vis) 1998, p 87-94.
- Using Strahler numbers for real time visual exploration of huge graphs. David Auber. Intl Conf. Computer Vision and Graphics, 2002, p 56-69?

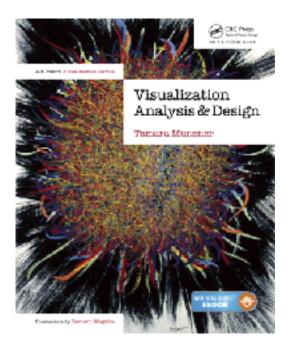
# Visualization Analysis & Design

# Analysis: Nested Model (Ch 4)

### Tamara Munzner

Department of Computer Science University of British Columbia

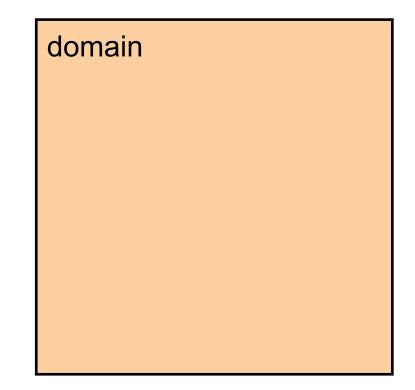
<u>@tamaramunzner</u>



### How to evaluate a visualization: So many methods, how to pick?

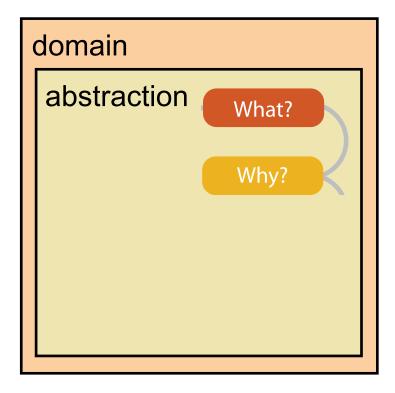
- Computational benchmarks?
  - -quant: system performance, memory
- User study in lab setting?
  - -quant: (human) time and error rates, preferences
  - -qual: behavior/strategy observations
- Field study of deployed system?
  - -quant: usage logs
  - -qual: interviews with users, case studies, observations
- Analysis of results?
  - -quant: metrics computed on result images
  - -qual: consider what structure is visible in result images
- Justification of choices?
  - -qual: perceptual principles, best practices

- domain situation
  - -who are the target users?



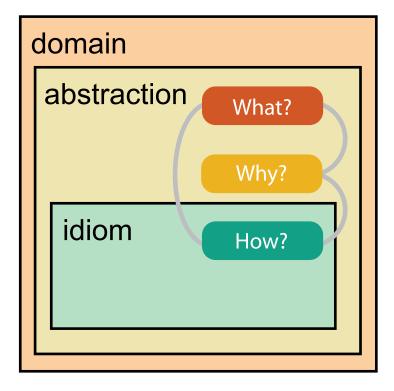
- domain situation
  - -who are the target users?
- abstraction
  - translate from specifics of domain to vocabulary of vis
    - what is shown? data abstraction
    - why is the user looking at it? **task** abstraction

[A Multi-Level Typology of Abstract Visualization Tasks. Brehmer and Munzner. IEEE TVCG 19(12):2376-2385, 2013 (Proc. InfoVis 2013).] [A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]



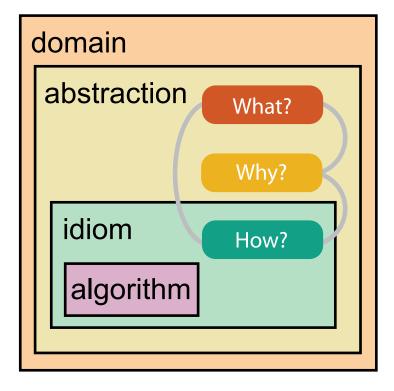
- domain situation
  - -who are the target users?
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  - translate from specifics of domain to vocabulary of vis
    - what is shown? data abstraction
    - why is the user looking at it? **task** abstraction
- idiom
  - -**how** is it shown?
    - visual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

[A Multi-Level Typology of Abstract Visualization Tasks. Brehmer and Munzner. IEEE TVCG 19(12):2376-2385, 2013 (Proc. InfoVis 2013).] [A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]



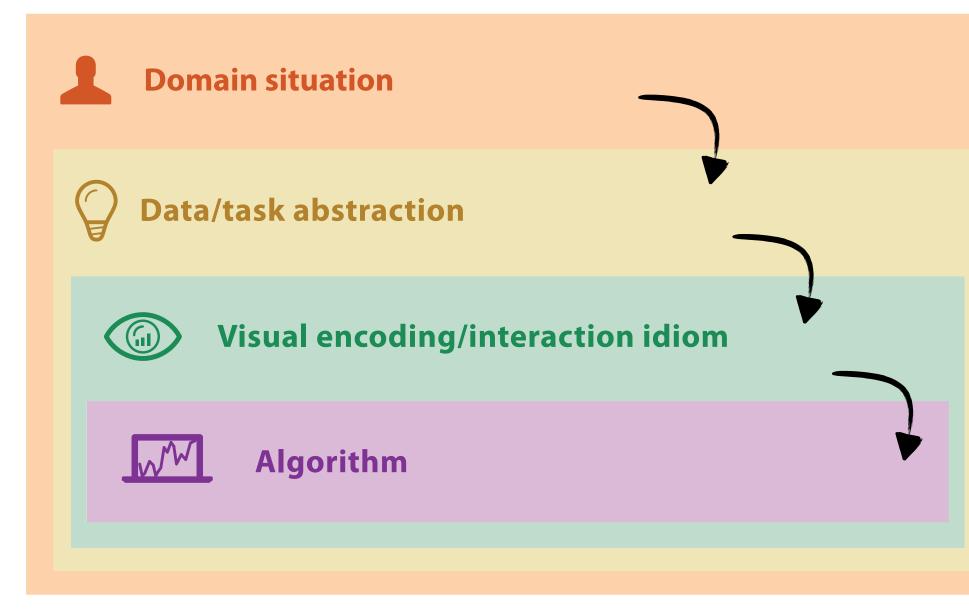
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- abstraction
  - translate from specifics of domain to vocabulary of vis
    - what is shown? data abstraction
    - why is the user looking at it? **task** abstraction
- idiom
  - -**how** is it shown?
    - visual encoding idiom: how to draw
    - **interaction** idiom: how to manipulate
- algorithm
  - efficient computation

[A Multi-Level Typology of Abstract Visualization Tasks. Brehmer and Munzner. IEEE TVCG 19(12):2376-2385, 2013 (Proc. InfoVis 2013).] [A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]



### Nested model

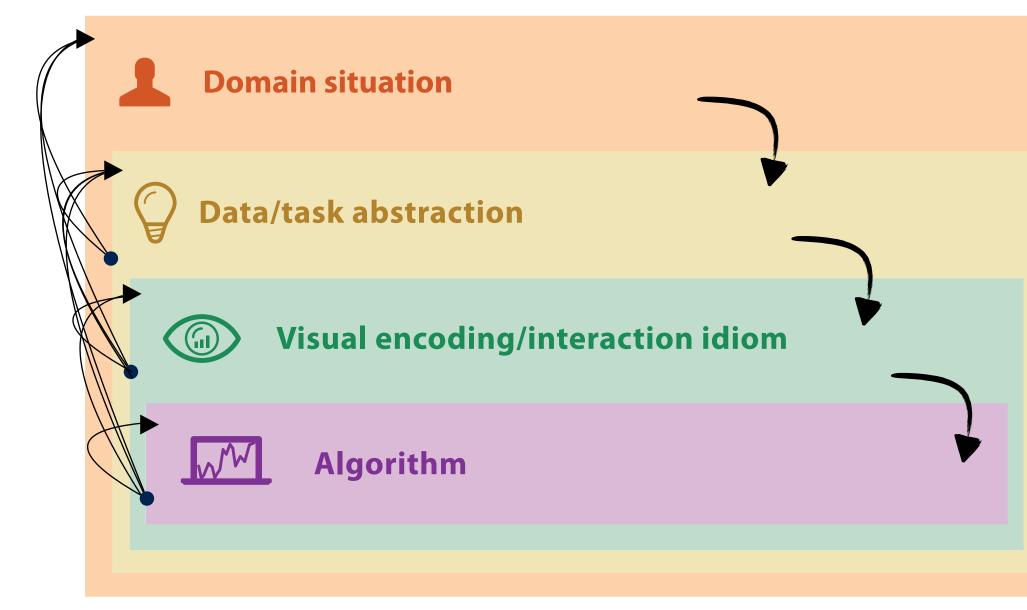
downstream: cascading effects





### Nested model

- downstream: cascading effects
- upstream: iterative refinement





• different ways to get it wrong at each level

**L** Domain situation You misunderstood their needs Data/task abstraction  $\bigcirc$ You're showing them the wrong thing **Wisual encoding/interaction idiom** The way you show it doesn't work Algorithm WW Your code is too slow

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

solution: use methods from different fields at each level

Algorithm Measure system time/memory Analyze computational complexity

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

solution: use methods from different fields at each level

computer science

**Algorithm** Measure system time/memory Analyze computational complexity



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

# 💫 technique-driven

solution: use methods from different fields at each level

design

computer science

cognitive psychology

Visual encoding/interaction idiom Justify design with respect to alternatives	
Algorithm Measure system time/memory Analyze computational complexity	▶ techni work
Analyze results qualitatively Measure human time with lab experiment ( <i>lab study</i> )	

# ique-driven

solution: use methods from different fields at each level

anthropology/ ethnography

design

computer science

cognitive psychology

anthropology/ ethnography

	<b>Domain situation</b> Observe target users using existing tools	
	Data/task abstraction	
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[A Nested Model of Visualization Design and Validation. Munzner. IEEE TVCG 15(6):921-928, 2009 (Proc. InfoVis 2009).]

### inique-driven k

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### problem-driven work (design study)

hnique-driven rk

### Avoid mismatches

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lab studies do not confirm task abstraction

computational benchmarks do not confirm idiom design

### Analysis examples: Single paper includes only subset of methods

MatrixExplorer. Henry and Fekete. InfoVis 2006.

observe and interview target users

justify encoding/interaction design

measure system time/memory

qualitative result image analysis

LiveRAC. McLachlan, Munzner, Koutsofios, and North. CHI 2008.

observe and interview target users

justify encoding/interaction design

qualitative result image analysis

field study, document deployed usage

An energy model for visual graph clustering. (LinLog) Noack. Graph Drawing 2003

qualitative/quantitative image analysis

Effectiveness of animation in trend visualization. Robertson et al. InfoVis 2008.

lab study, measure time/errors for operation

Interactive visualization of genealogical graphs.

McGuffin and Balakrishnan. InfoVis 2005.

justify encoding/interaction design

qualitative result image analysis test on target users, get utility anecdotes

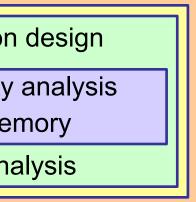
Flow map layout. Phan et al. InfoVis 2005.

justify encoding/interaction design

computational complexity analysis

measure system time/memory

qualitative result image analysis



## Further reading

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

- Chap 4: Analysis: Four Levels for Validation

- Storks Deliver Babies (p= 0.008). Robert Matthews. Teaching Statistics 22(2):36-38, 2000.
- The Earth is spherical (p < 0.05): alternative methods of statistical inference. Kim J.Vicente and Gerard L. Torenvliet. Theoretical Issues in Ergonomics Science, 1(3):248-271, 2000.
- The Prospects for Psychological Science in Human-Computer Interaction. Allen Newell and Stuart K. Card. Journal Human-Computer Interaction 1(3):209-242, 1985.
- How to do good research, get it published in SIGKDD and get it cited!, Eamonn Keogh, SIGKDD Tutorial 2009.
- False-Positive Psychology: Undisclosed Flexibility in Data Collection and Analysis Allows Presenting Anything as Significant. Joseph P. Simmons, Leif D. Nelson and Uri Simonsohn. Psychological Science 22(11):1359-1366, 2011.
- Externalisation how writing changes thinking. Alan Dix. Interfaces, Autumn 2008.



### Guerilla/Discount Usability

- 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
- The Art of Guerrilla Usability Testing, David Peter Simon
  - <u>http://www.uxbooth.com/articles/the-art-of-guerrilla-usability-testing/</u>
- Discount Usability: 20 Years, Jakob Nielsen
  - <u>https://www.nngroup.com/articles/discount-usability-20-years/</u>
- Interaction Design: Beyond Human-Computer Interaction
  - Preece, Sharp, Rogers. Wiley, 4th edition, 2015.
- About Face: The Essentials of Interaction Design
  - Cooper, Reimann, Cronin, Noessel. Wiley, 4th edition, 2014.
- Task-Centered User Interface Design. Lewis & Rieman, 1994 – <u>http://hcibib.org/tcuid/</u>
- Designing with the Mind in Mind. Jeff Johnson. Morgan Kaufmann, 2nd, 2014.