Visualization & Visual Data Science: A Primer

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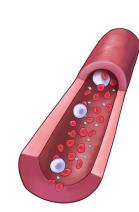
ICTP Workshop 2022 2. December 2022



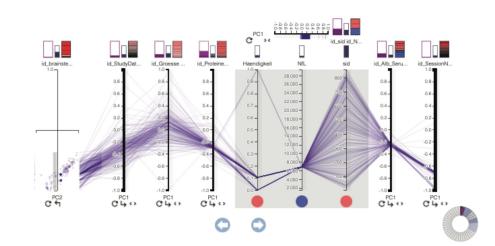
A bit about me...

Source: The Oatmeal



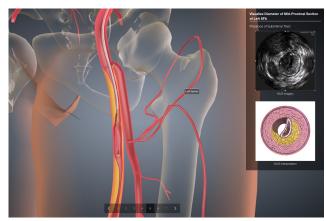


MS Biomedical Visualization



PhD Visualization,University of Bergen (UiB)

BA Biology/Physiology, Art, & Chemistry

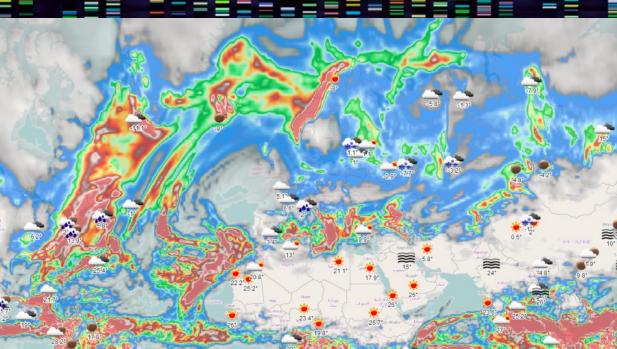


Research fellow, UiB/MMIV Consultent, Bouvet

Medical & Health Tech Start-ups



Source: BioDigital



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Data Science

- Extract knowledge and insights from data, often using advanced analytics methods from:
 - Mathematics
 - Statistics
 - Algorithms
 - Machine learning



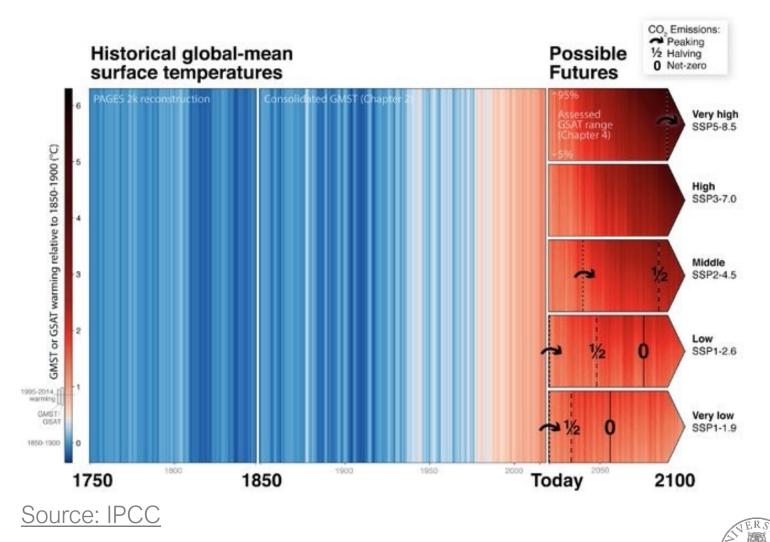
Src: Kiranshastry - Flaticon



Dhar, V. (2013). Data science and prediction. *Communications of the ACM*, 56(12), 64-73.

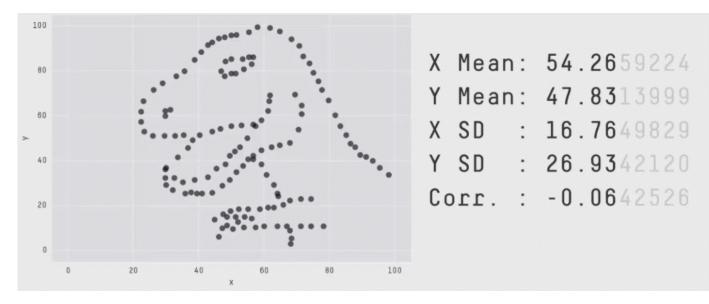
Visualization

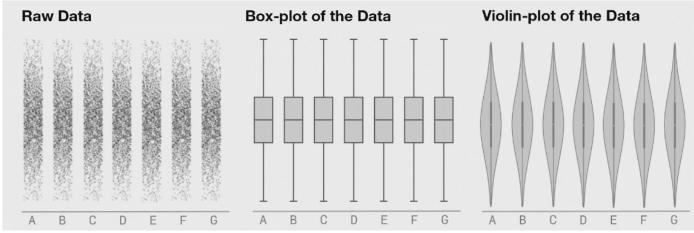
Computer-based visualization systems provide <u>visual</u> <u>representations</u> of <u>datasets</u> designed to help <u>people</u> carry out tasks more effectively.



Why visualize?

- "visual representation"
 - replace cognition with perception
- "representations of datasets"
 - details matter, summaries can lose information



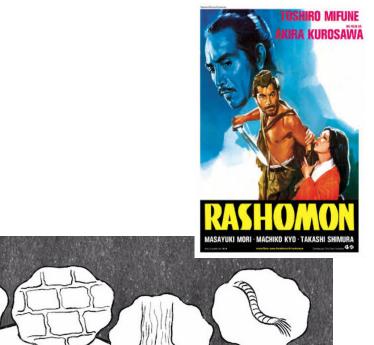


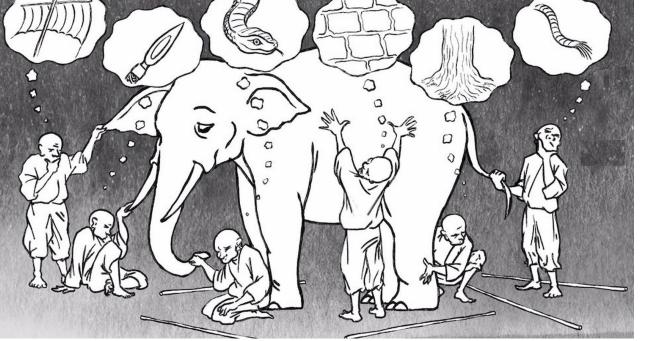


Matejka, J., & Fitzmaurice, G. (2017). Same stats, different graphs: generating datasets with varied appearance and identical statistics through simulated annealing. In Proceedings of the 2017 CHI conference on human factors in computing systems (pp. 1290-1294).

Why visualize?

- Rashomon Effect
 - Different models, parameters, representations, etc. can tell different stories
- Visualization can help us spot and understand reasons for these differences

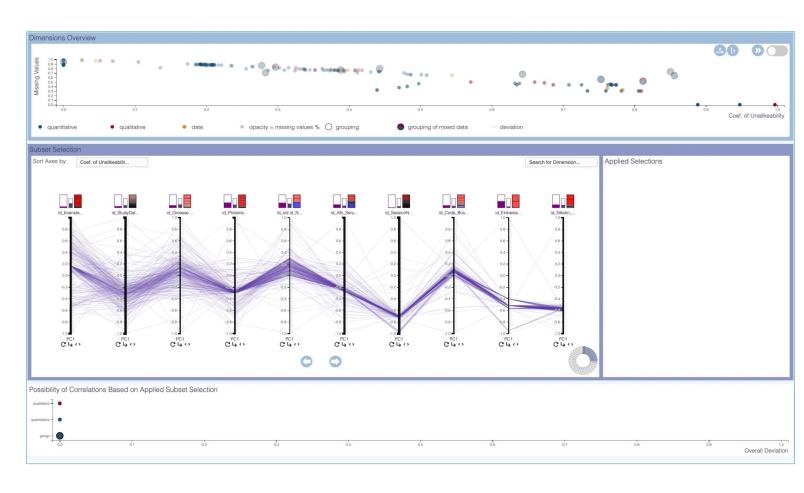






Why user in the loop?

- Visualization is not necessary if there is a trustworthy automatic solution
- Augment human capabilities, not replace
- Many analysis problems are ill-specified

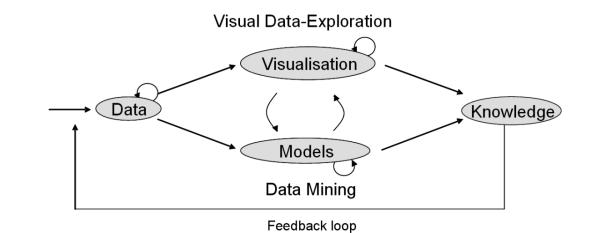


Munzner, T. (2014). Visualization analysis and design. AK Peters Visualization Series, CRC Press, Visualization Series. Chp 1 Garrison, L., Müller, J., Schreiber, S., Oeltze-Jafra, S., Hauser, H., & Bruckner, S. (2021). Dimlift: Interactive hierarchical data exploration through dimensional bundling. IEEE Transactions on Visualization and Computer Graphics, 27(6), 2908-2922.



Why user in the loop?

- Visualization integral throughout analysis pipeline to help steer a path of inquiry
- Help answer and form new questions:
 - What do my data look like?
 - What are the requirments for developing a more complex model?
 - What if I remove/adjust this parameter?
 - Do I trust/can I verify the data/model?

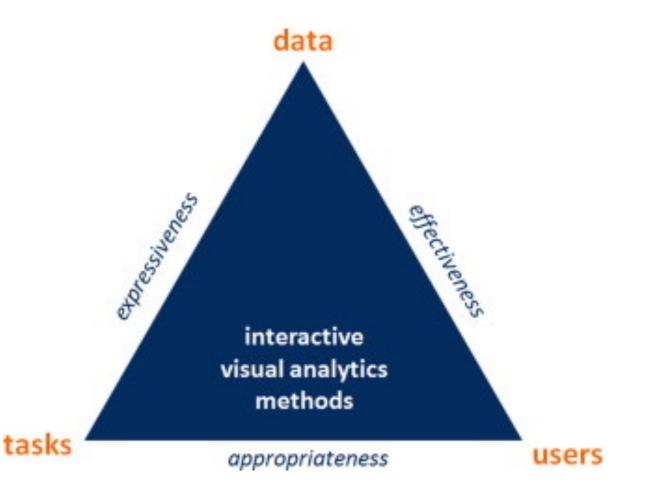




Keim, D., Andrienko, G., Fekete, J. D., Görg, C., Kohlhammer, J., & Melançon, G. (2008). Visual analytics: Definition, process, and challenges. In Information visualization (pp. 154-175). Springer, Berlin, Heidelberg.

Visualization in YOUR workflow

- Good visualization is possible for anyone
- Consider:
 - Data
 - Users
 - Tasks





Miksch, Silvia, and Wolfgang Aigner. "A matter of time: Applying a data–users–tasks design triangle to visual analytics of time-oriented data." Computers & Graphics 38 (2014): 286-290.

Visual Analysis Framework

- Domain
 - Who are you visualizing for? Yourself?
- Abstraction
 - What is shown (data abstraction)
 - Why showing (task abstraction)
- Idiom
 - How is it being shown
 - visual encoding idiom (how do you draw the picture)
 - interaction idiom (how do you manipulate the picture)
- Algorithm
 - efficient computation to show the picture

d	omain
ł	abstraction What?
	Why?
	idiom How?
	algorithm



What is shown [data abstraction]



Data

- 21.11.2022, 11, 58, 2665
 - What does/could this sequence of numbers mean?



Data

- 21.11.2022, 11, 58, 2665
 - What does/could this sequence of numbers mean?
 - Date, stair flight count, average heart rate, step count



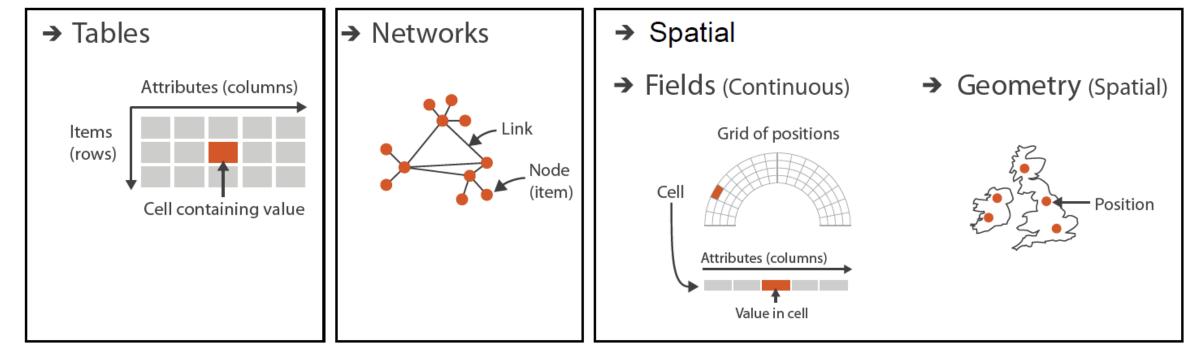
Data Interpretation

- Semantics
 - Real world meaning of the data
- Type
 - Structural or mathematical interpretation
 - item, link, attribute, position (if in a grid)
 - Different from data types in programming languages!
 - Think about operations that are meaningful for each type
 - e.g. 22.11/01.11 makes sense if this is a decimal, but not if this is a date



Major Data Types

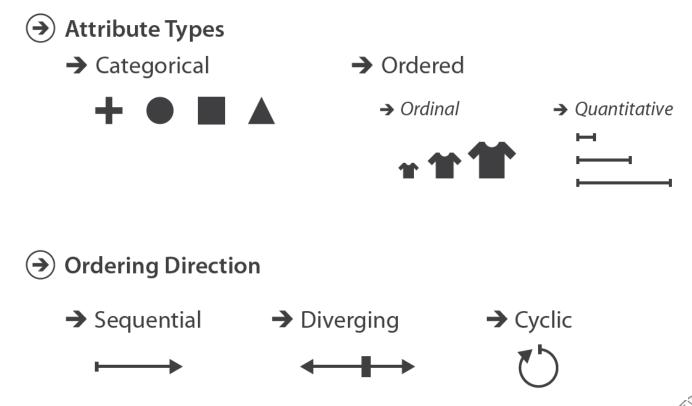
Dataset Types





Attribute Types and Ordering

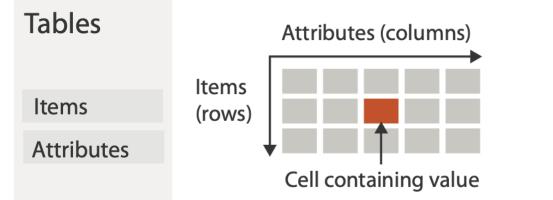
- Item
 - Discrete, individual entity
 - e.g., patient
 - "independent variable"
- Attribute
 - Property that is measured, observed
 - e.g., height, BMI
 - "dependent variable"





Table

- Flat table
 - Row = item
 - Column = attribute
 - Cell = item-attribute pair
 - Unique key (possibly implicit)

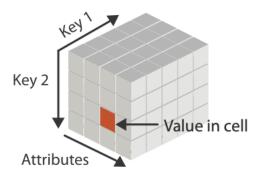


A	В	C	S	Т	U
Order ID	Order Date	Order Priority	Product Container	Product Base Margin	Ship Date
3	10/14/06	5-Low	Large Box	0.8	10/21/06
6	2/21/08	4-Not Specified	Small Pack	0.55	2/22/08
32	7/16/07		Small Pack	0.79	7/17/07
32	7/16/07	2-High	Jumbo Box	0.72	7/17/07
32	7/16/07	2-High	Medium Box	0.6	7/18/07
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		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		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		3-Medium	Wrap Bag	0.42	4/7/08

18

Multidimensional Table

 Index according to multiple keys

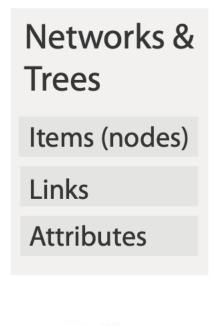


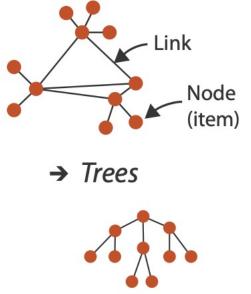
2		A		В	C		D		E		
1	1	1	A	B		С		D		E	
2	1	4	1	A I	8	C		D		E	
3	\$ 2	1	H .	I A	B		c		D		5
1	3	62	1	#1.2					12		2
i	1.4	13	6 2	1500	529						
5	15	F4	11 2	GeneName		TCCA-02-0001-	010-010-0177-01	TCGA-02-0003-0	14.010.0172.01	TOCA-02-00044	14.010.0208.0
7	6	15	PI 4	LTF	LTF	ICGH-02-00014	-1.265728057		2.377012066		4.1239795
8	17	16	TIS	POSTN	POSTN		2.66241180		3.932400324		5.0315853
9	8	F7	HG		TM5L8		-3.082217838		-2.243148513		
0	9	5.8	8 7	TMSL8							-0.023135
1	10	[9	-	HLA-DQA1	HLA-DQA1		-1.739664398		4.577962344		3.1277449
2	111	/10	0		RP11-35N6.1		-3.346352968		-2.895400157		-3.4730350
3	12	111	D 9 A 10	STMN2	STMN2		-2.578511100		-3.051605144		-1.7298928
4	13	\$12	~ 10	DOX	DCX		-2.25078976		-2.529795801		-2.8449662
5	14	113		AGXT2L1	AGXT2L1		-2.639493611		-3.113204863		-0.4039750
6	115	-	SI 12	IL13RA2	IL13RA2		-2.93596915		-1.873600916		2.9762569
7	16	\$14	N 13	SLN	SLN		-2.465718221	1	-2.208406749		1.0258279
8	-	115	C 14	MEOX2	MEOX2		-2.395054058	6	1.062676046		1.7832353
q	(17	F16	N 15	COL11A1	COL11A1		1.211934832	1	-0.359392588		4.7336089
0	18	¢17	E 16	NNMT	NNMT		0.703745154	1	0.664082419		3.0690307
1	119	18	C 17	F13A1	F13A1		-0.224094042	1	2.222197544		1.1713547
	20	119	N 18	CKCL14	CXCL14		-3.1309694	1	-1.395056071		2.5695406
2	(21	120	TI 19		MEP		-1.905390558		-2.037626447		-2.9357449
	22	(21	N 20	TF	TF		-4.334123292	1	-4.680680246		-2.9757888
		22	621		KCND2		-1.777692395	1	-2.100352021		-1.9963060

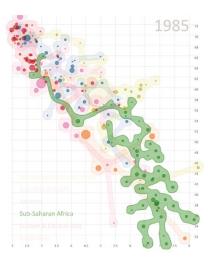


Network

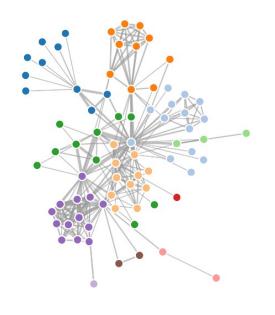
- Nodes (vertices)
 connected by links
- Trees are special case of acyclic network
 - usually directed
 - usually w/roots







vialab.science.uoit.ca/portfolio/bubblesets

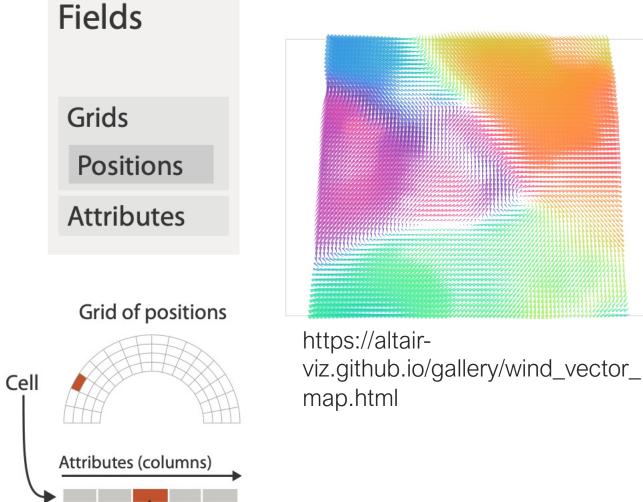


https://observablehq.com/@d3/force-directed-graph



Spatial Field

- Attributes associate with cells
- Grid subdivides continuous domain into cells
 - e.g. temperature
- Measured or simulated data
- Concerns
- Divisions





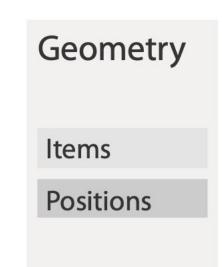
10 12

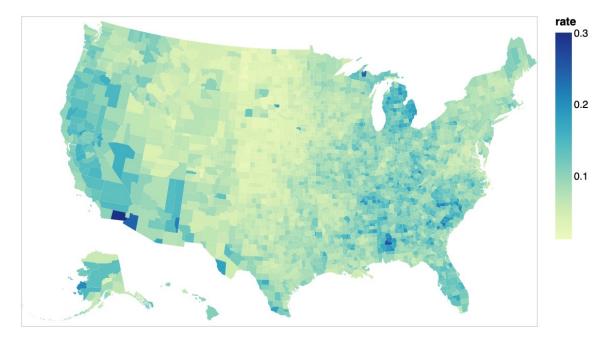
Munzner, T. (2014). Visualization analysis and design. AK Peters Visualization Series, CRC Press, Visualization Series. Chp 2.

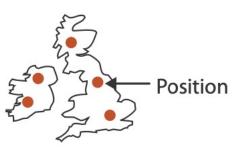
Value in cell

Geometry

- Explicit position
- Shape of items
 - Point
 - 1D line
 - 2D shape
 - 3D volume





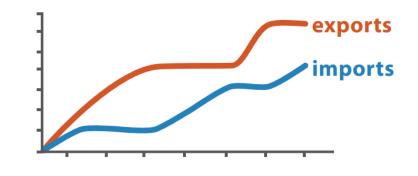


https://altair-viz.github.io/gallery/choropleth.html

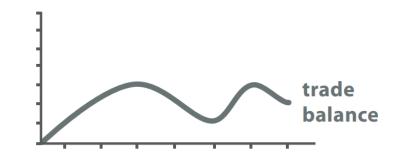


Data Abstraction Steps

- Identify dataset type(s) and attribute type(s)
- 2. Identify cardinality
 - 1. How many items in dataset?
 - 2. Cardinality of each attribute?
 - 1. Number of levels/segments for categorical data
 - 2. Range of continuous data
- 3. Consider whether to transform data
 - 1. Determine if necessary according to user task



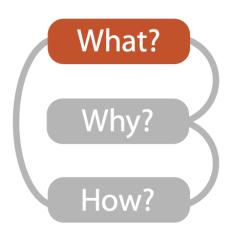




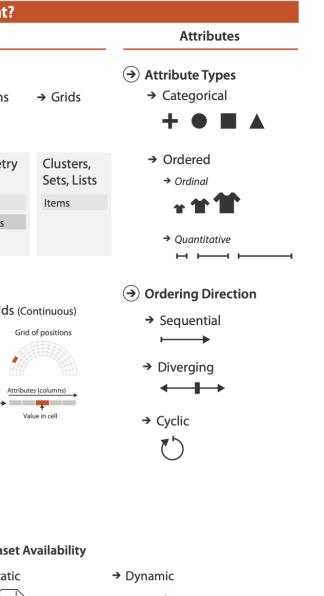
trade balance = exports – imports

Derived Data





			What?		
	D	atasets			Attributes
	Attributes	→ Links	→ Positions	→ Grids	 → Attribute Types → Categorical + ● ■ ▲
) Data and Data Tables Items Attributes	Networks & TreesItems (nodes)LinksAttributes	Fields Grids Positions Attributes	Geometry Items Positions	Clusters, Sets, Lists Items	 → Ordered → Ordinal ▲ ▲ ▲ → Quantitative
Items (rows)	→ N res (columns) training value		ode em)	ontinuous) of positions es (columns) es (columns)	 Ordering Direction Sequential Diverging Cyclic
→ Geometry	(Spatial)		→ Dataset A→ Static	vailability	→ Dynamic
E Contraction	- Position		•••		••••••••



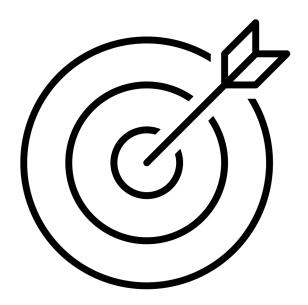


Why showing? [task abstraction]



{action, target}

- Action: Analyze, search, query
- Target: What is being acted on
- {action, target}"
 - Discover distribution
 - Compare trends
 - Browse outliers
 - Explore topology
 - •





Actions: High Level Choices

EXPLORATION

Browse data Generate hypotheses

Communication Present Enjoy Produce

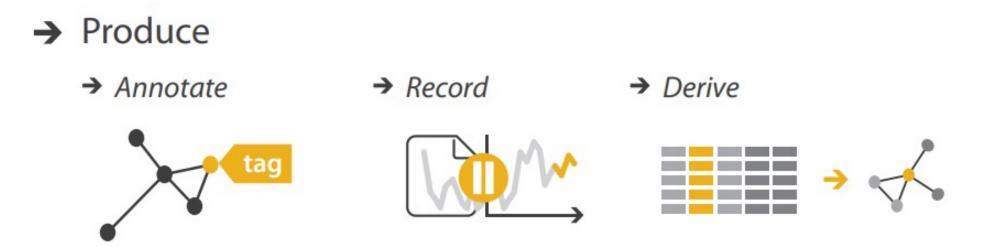
ANALYSIS

Verify Search: lookup, locate Query: identify, compare, summarize



Brehmer, M., & Munzner, T. (2013). A multi-level typology of abstract visualization tasks. *IEEE transactions on visualization and computer graphics*, *19*(12), 2376-2385.

Actions: High Level Choices





Actions: Search

- What does the user know?
 - Target
 - Location

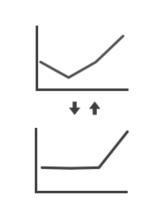
	Target known	Target unknown		
Location known	• • • Lookup	• • Browse		
Location unknown	< Ocate	C O Explore		



Actions: Query

- How much of the data matters?
 - One: identify
 - Some: compare
 - All: summarize





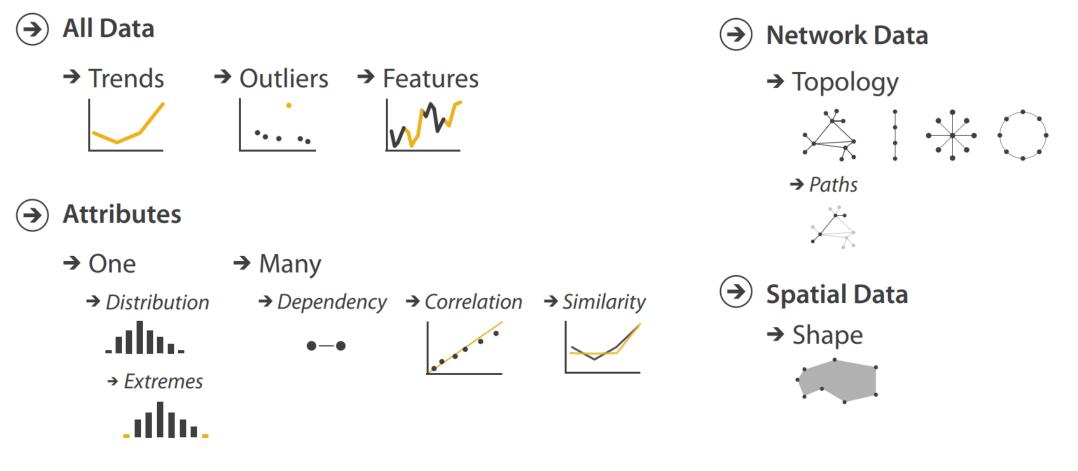
→ Compare







Targets

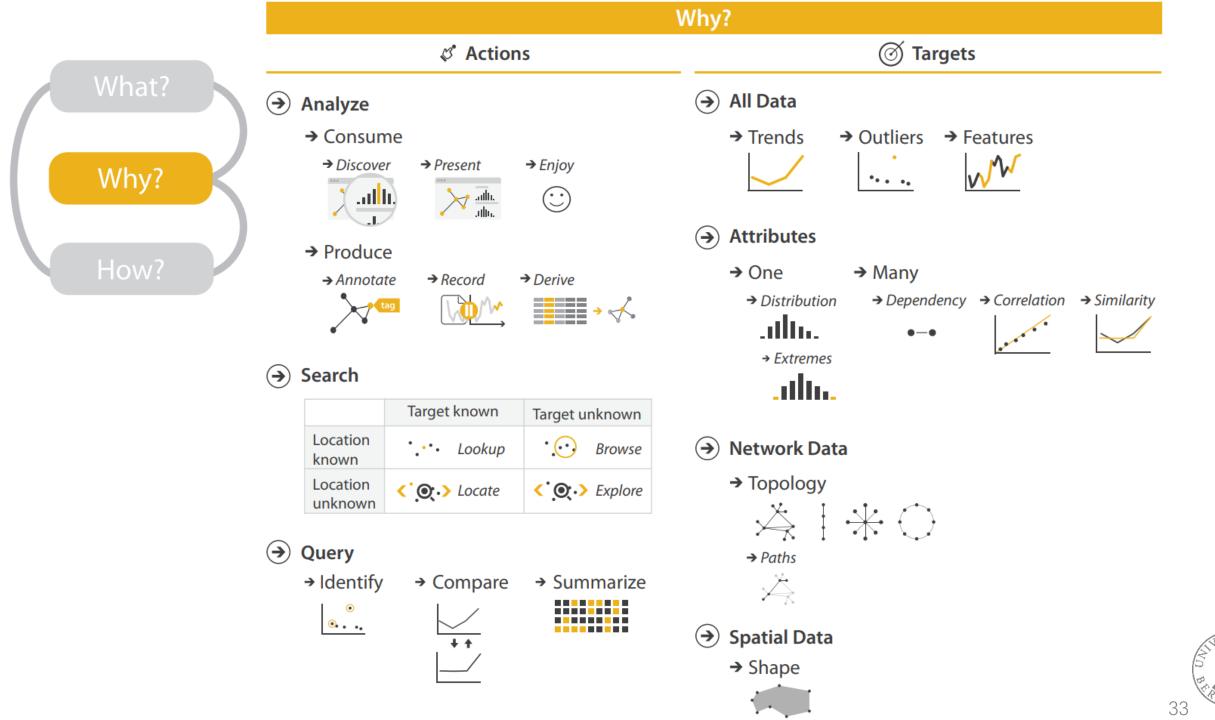




Data <-> Task Abstraction

- Data abstraction required within task abstraction
 - Specify targets
 - May lead to data transformation (if task requires)
 - e.g., species diversity index
- Iterate between data and task abstraction
 - Not just a one-way street





How showing? [visual abstraction]

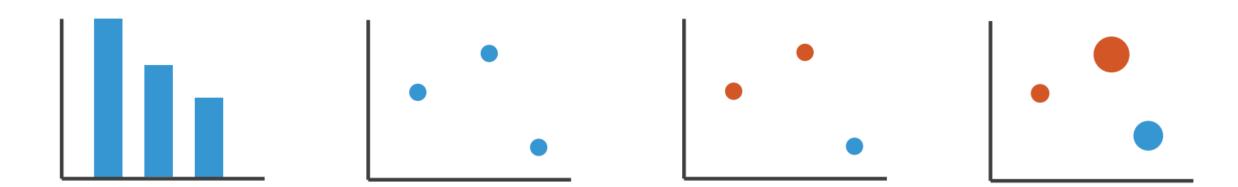


"A picture is worth a thousand words...if you know how to read it."

- Alberto Cairo



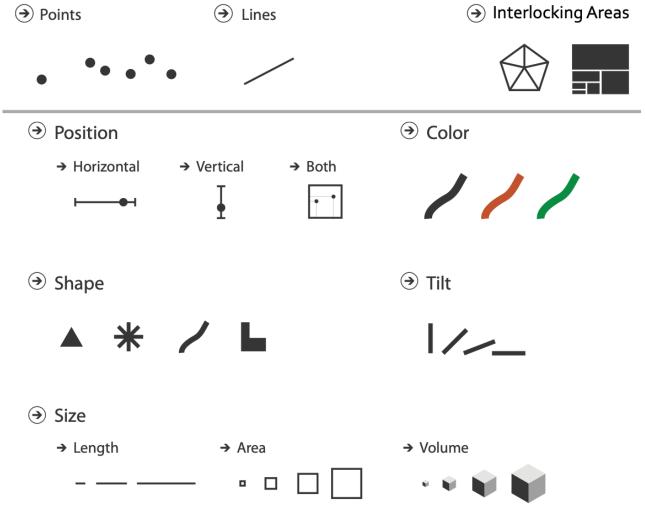
Visual Encoding





Marks and Channels

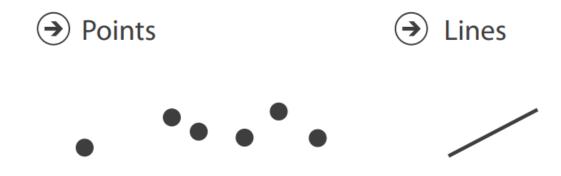
- Marks
 - Represent items or links
- Channels
 - Change the appearance of marks based on attributes





Marks for Items

• 0D, 1D, 2D (3D is rarely used)



• Can also indicate links

Containment
 Connection

• • • •



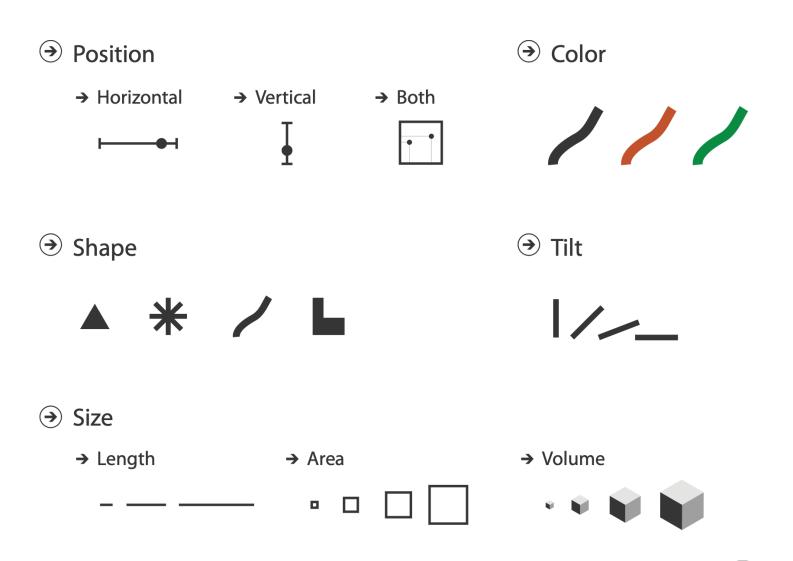
Interlocking Areas





Channels

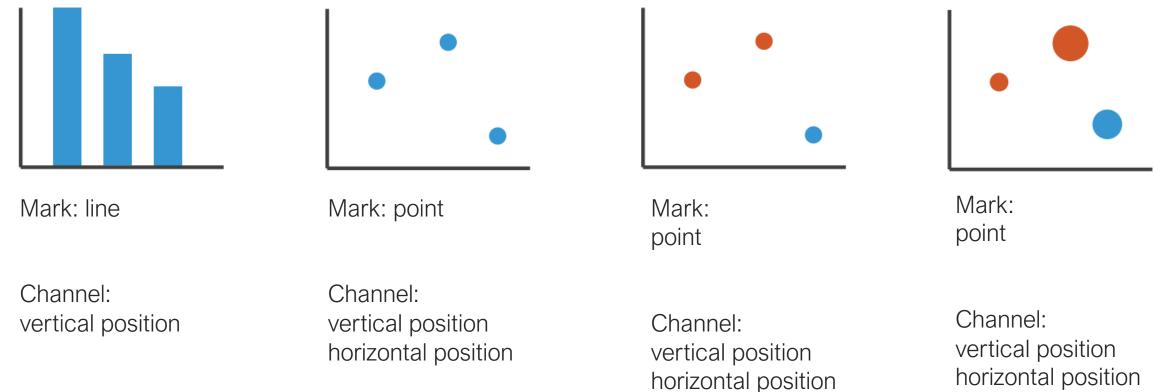
- Control appearance of marks
- Channel properties
 differ
 - Type and amount of information that can be conveyed to human perceptual system





Analyze idiom structures

Charts can be thought of combinations of marks and channels



color (hue)

vertical position horizontal position color (hue) size (area)

Match types

Expressiveness

 match channel type to data type

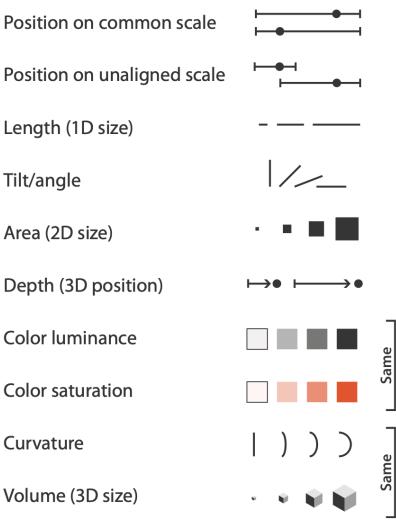
Position on common scale		Spatial region	
Position on unaligned scale		Color hue	
Length (1D size)		Motion	•••
Tilt/angle	//	Shape	+•
Area (2D size)	· • • •		
Depth (3D position)	$\longmapsto \bullet \longmapsto \bullet$		
Color luminance			
Color saturation	Same		
Curvature	اي ((ا		
Volume (3D size)	Same		

What is "best?"

Effectiveness

- some channels ARE better than others (perceptually)
- spatial position ranks high for both

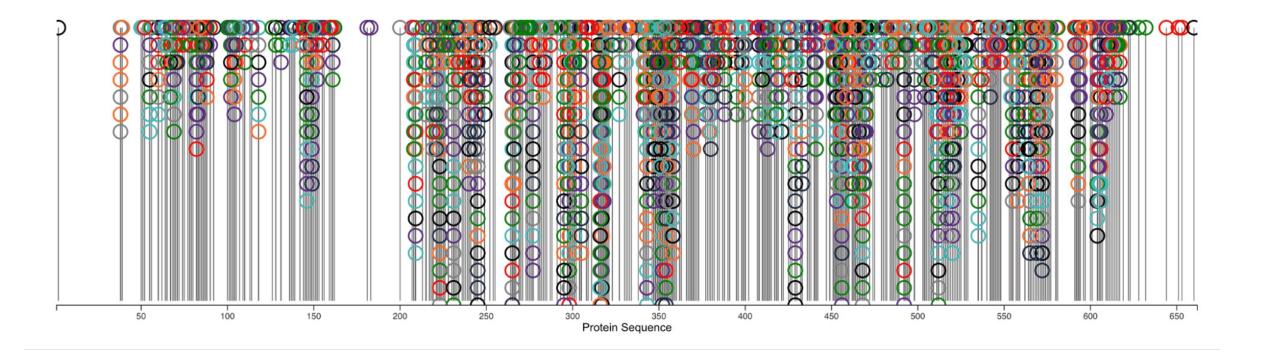
→ Magnitude Channels: Ordered Attributes



Identity Channels: Categorical Attributes

non scale		Spatial region	1 - E
gned scale		Color hue	
		Motion	••• G
	//	Shape	+ • • •
	· • •		
on)	$\longmapsto \bullet \longmapsto \bullet$		
	Same		
	Same		VERSY
	Same		



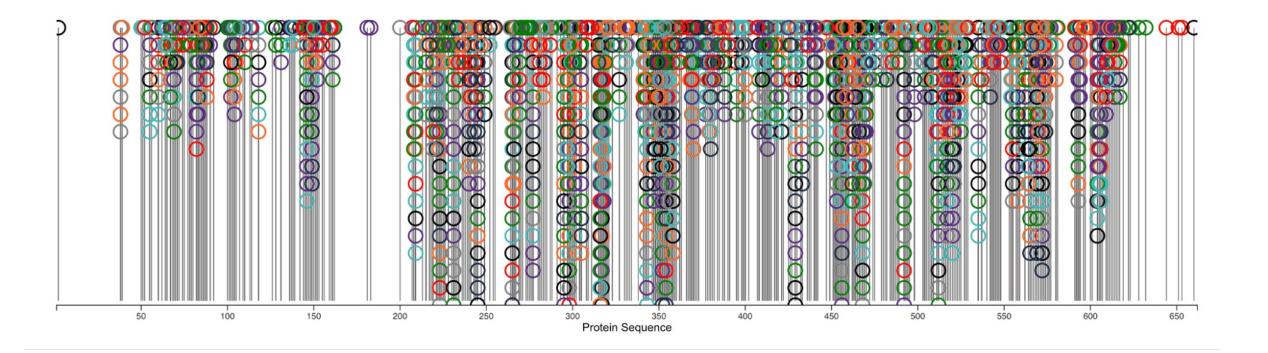








Other possibilities?





Other possibilities?

- → Magnitude Channels: Ordered Attributes
 - Position on common scale
 - Position on unaligned scale
 - Length (1D size)
 - Tilt/angle
 - Area (2D size)
 - Depth (3D position)
 - Color luminance
 - Color saturation
 - Curvature
 - Volume (3D size)

- - 1_

Same

Same

Shape

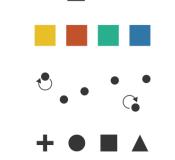
Motion

Spatial region

Color hue

 (\mathbf{i})

Points



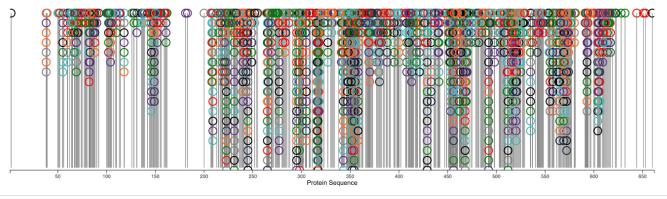




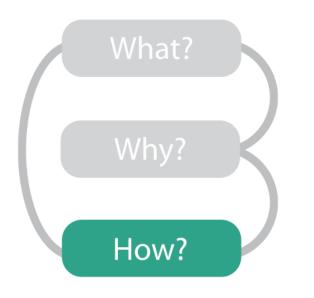
→ Identity Channels: Categorical Attributes

 (\mathbf{i})

Lines







Encode

→ Arrange

→ Express

→ Order

➔ Separate



→ Align



→ Use



 Map from categorical and ordered attributes

> → Color → Hue → Saturation → Luminance

→ Size, Angle, Curvature, ...



→ Shape
+ ● ■ ▲

→ Motion Direction, Rate, Frequency, ...



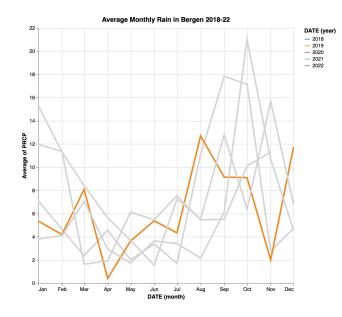


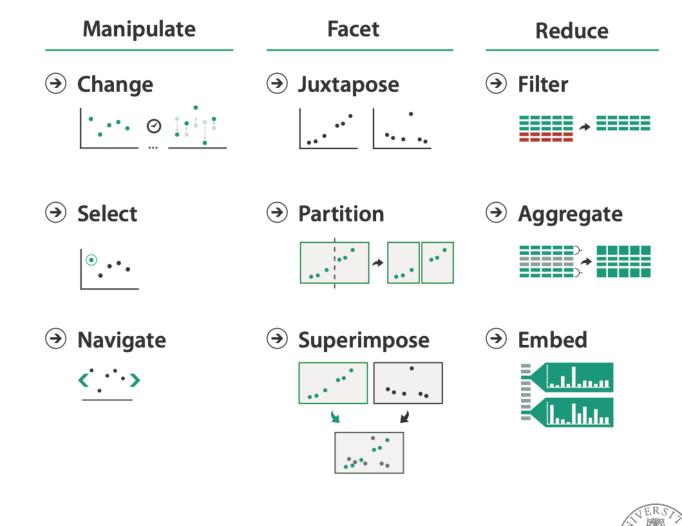
How showing? [interaction abstraction]



Interaction encodings

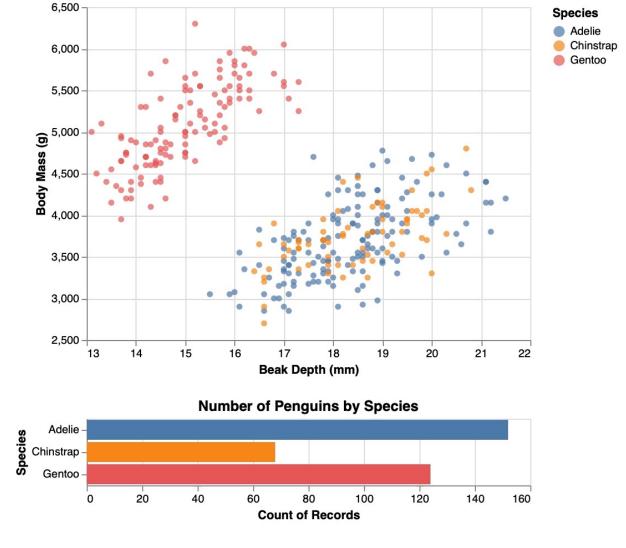
- Efficiency
- Deal with complexity through interactions in addition to directly visually encoding data





Manipulate

- Change over time
 - e.g., animated transitions
- Select
 - e.g., highlighting
- Navigate
 - e.g., pan within a view, scrollytelling, reduce attributes



50

day_05/basic_charts/linked_scatter_bar.py

Facet

- Juxtapose
- Partition
 - e.g., split into regions by attributes
- Superimpose
 - e.g., map layers (roads, terrain)

!! Tooltips/rollover are a type of faceting – but if you employ these, assume the user may not discover this information

- → Coordinate Multiple Side By Side Views
 - → Share Encoding: Same/Different
 - → Linked Highlighting



→ Share Data: All/Subset/None



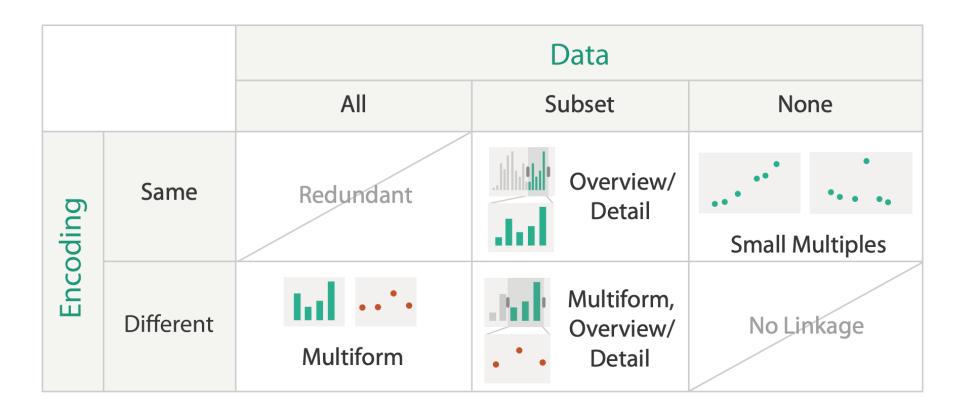
→ Share Navigation





Design Choices

Coordinated views





Coordinated Multiple Views

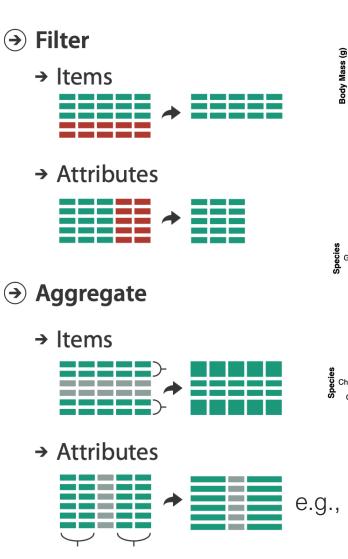


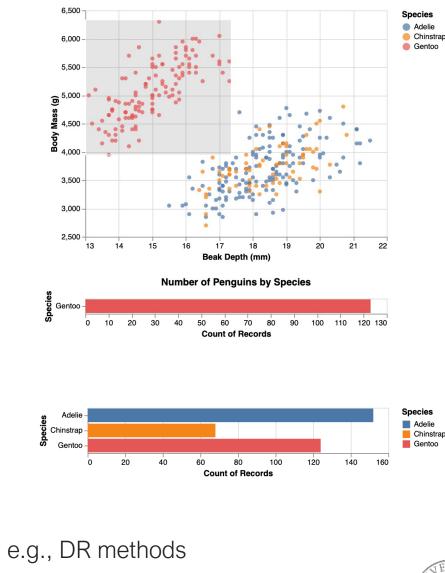
Garrison, L., Müller, J., Schreiber, S., Oeltze-Jafra, S., Hauser, H., & Bruckner, S. (2021). Dimlift: Interactive hierarchical data exploration through dimensional bundling. *IEEE Transactions on Visualization and Computer Graphics*, *27*(6), 2908-2922.

Reduce

- Filter
- Aggregate
- Embed

Not mutually exclusive! Can combine these and other interaction classes

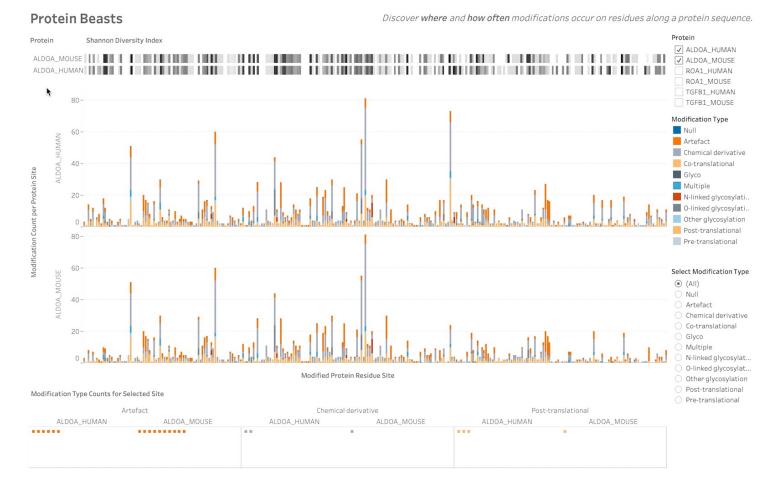




Derive

Derived H-index (top view)

Can you identify other interaction techniques in this visualization?

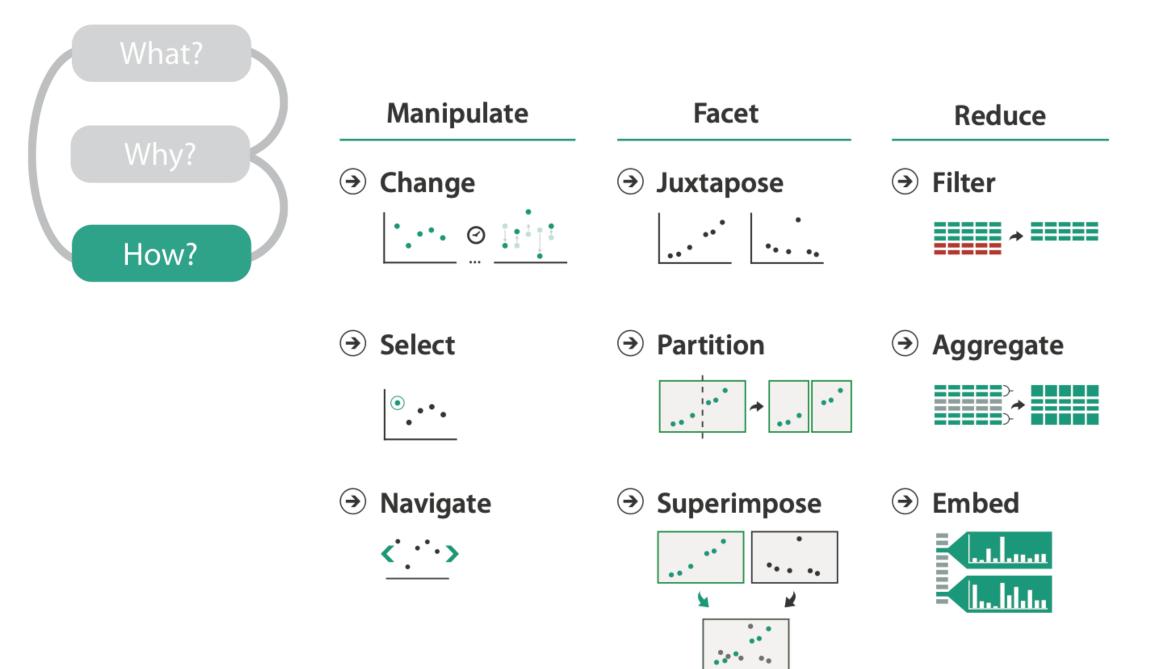






- Interaction has a time cost
- Users might not interact as you expect
 - NYTimes found that 90% of users don't interact beyond scrollytelling (Aisch 2016)
- Cognitive load for remembering prior state
- Controls vs invisible functionality
 - controls may take up valuable real estate
 - invisible functionality may be too invisible (no one figures it out)





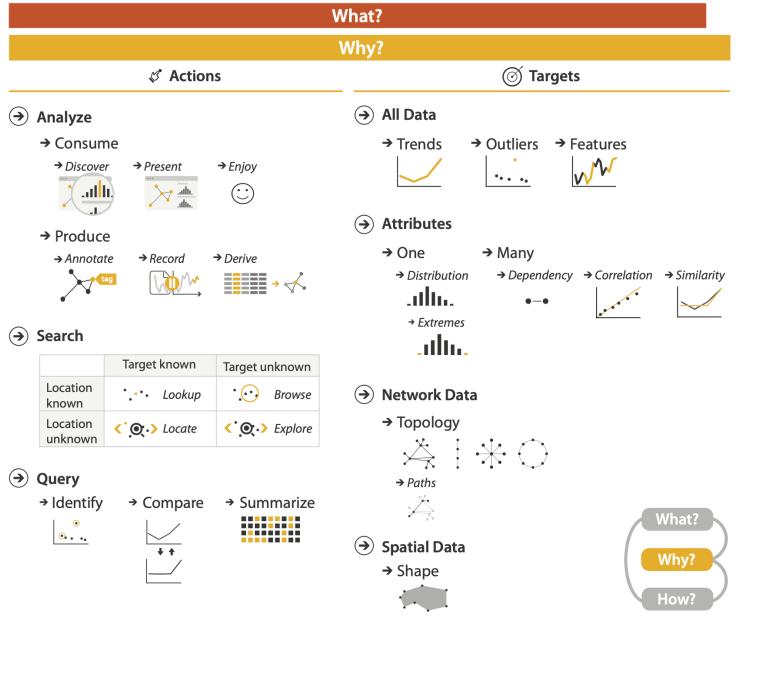


Summary

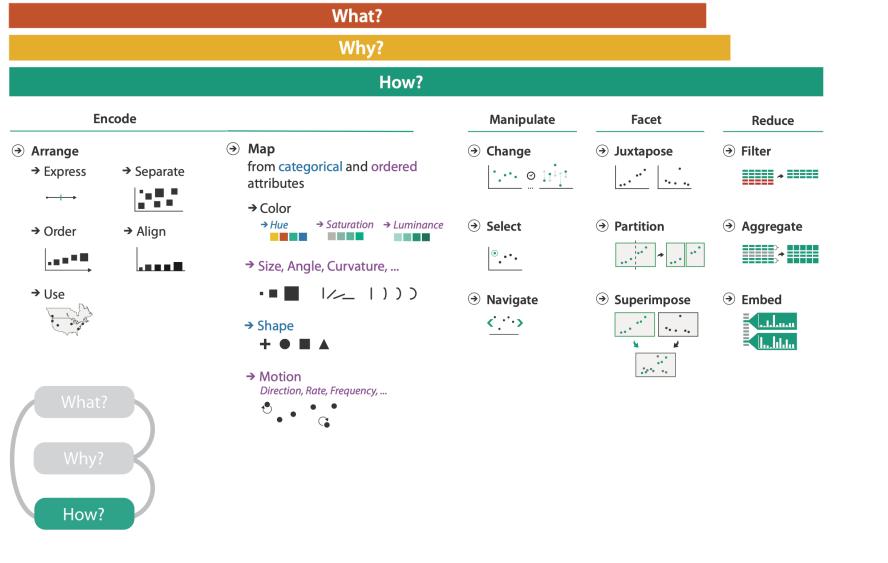


What?						
Datasets				Attributes		
	Data Types → Items → Data and Data	Attributes aset Types	→ Links -	→ Positions	→ Grids	 → Attribute Types → Categorical + ● ■ ▲
	Tables Items Attributes	Networks & Trees Items (nodes) Links Attributes	Fields Grids Positions Attributes	Geometry Items Positions	Clusters, Sets, Lists Items	 → Ordered → Ordinal → ▲ ▲ → Quantitative → ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓ ↓
 ◆ Dataset Types ◆ Tables ◆ Networks ◆ Fiel ✓ Curriculation (columns) ✓ Curriculation (columns)<					ontinuous) of positions	 Ordering Direction Sequential Diverging Cyclic
	→ Geometry (S	Spatial) Position		→ Dataset A→ Static	vailability	→ Dynamic ••••••••••••••••••••••••••••••••••







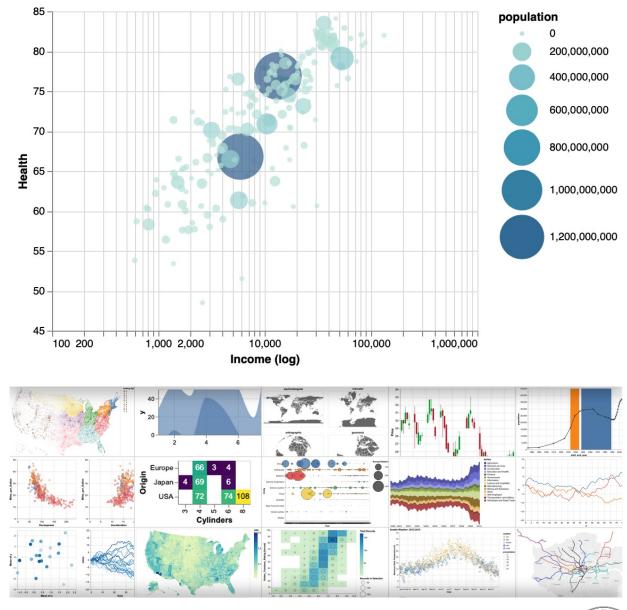


Data abstraction -> Task abstraction -> Visual + Interaction Encoding



Next Up

- Visualization in practice
- After lunch
 - Introduction to Vega-Altair
 - Mini-project in visualization



https://altair-viz.github.io/gallery/index.html



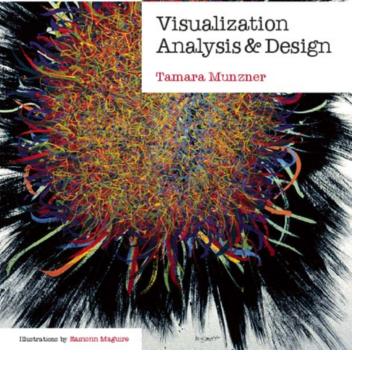
Questions?



Further Reading & Acknowledgement

 Web material for Visual Analysis & Design: <u>https://www.cs.ubc.ca/~tmm/talks/vadbook</u> (source material for many slides in this lecture)







Munzner, T. (2014). Visualization analysis and design. AK Peters Visualization Series, CRC Press, Visualization Series. Chp 2.