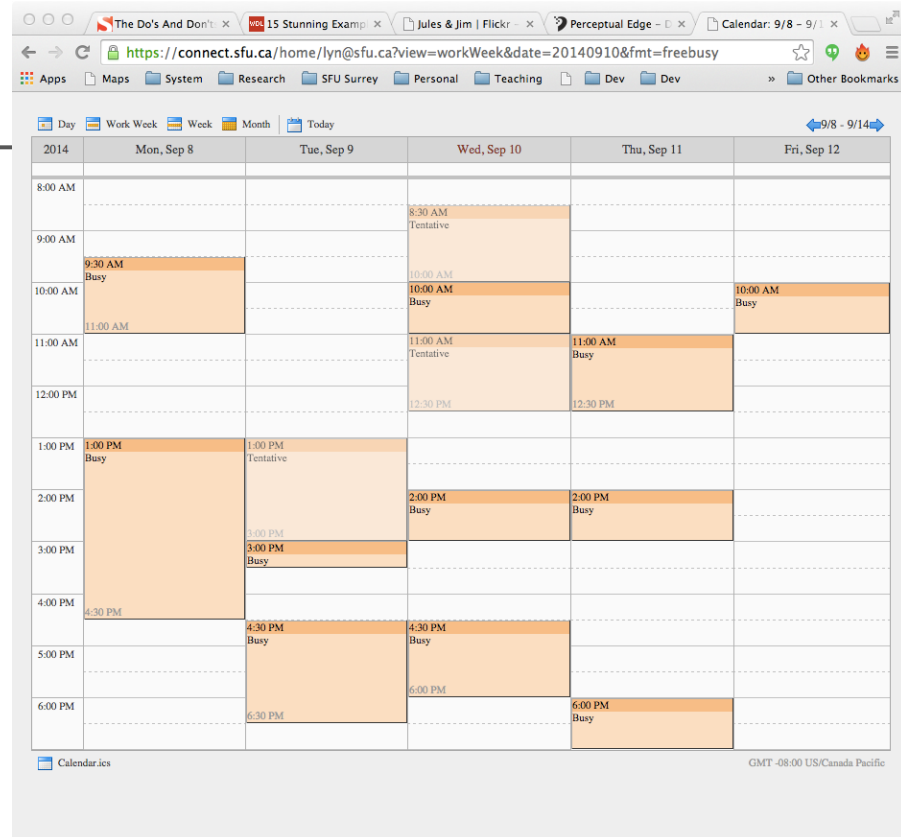


Administrivia: Assignment 1

- Visualization in practice
- Find one good and one bad example
- Put on Canvas discussion
- Present BRIEFLY in class
- We'll need to run over
- We will revisit these choices later!

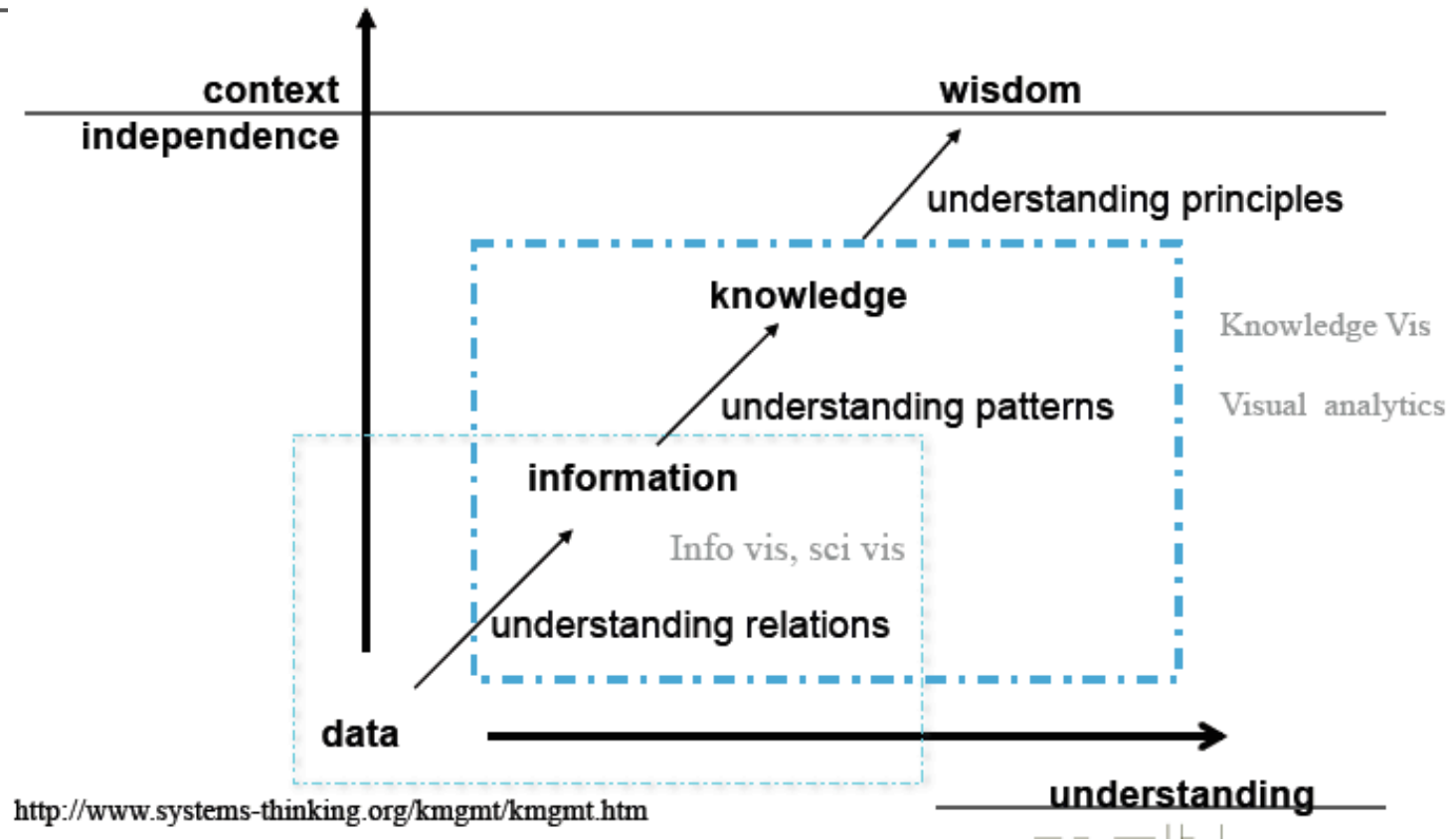
Administrivia 2

WHEN YOU WANT
TO MEET WITH ME



<https://connect.sfu.ca/home/lyn@sfu.ca?fmt=freebusy>

The desired progression



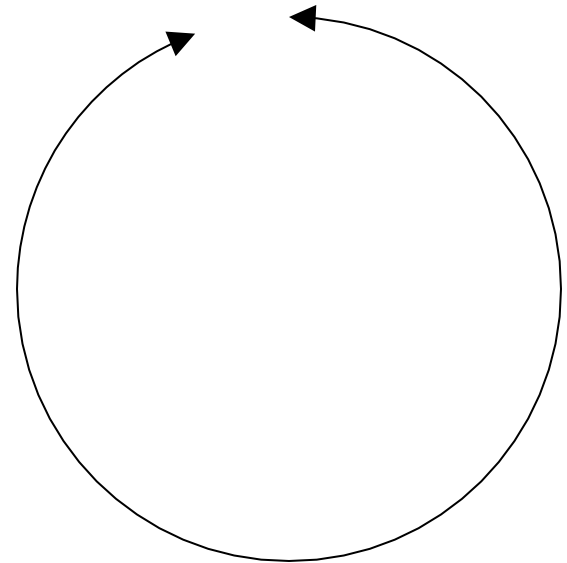
Another way of looking at it

Data

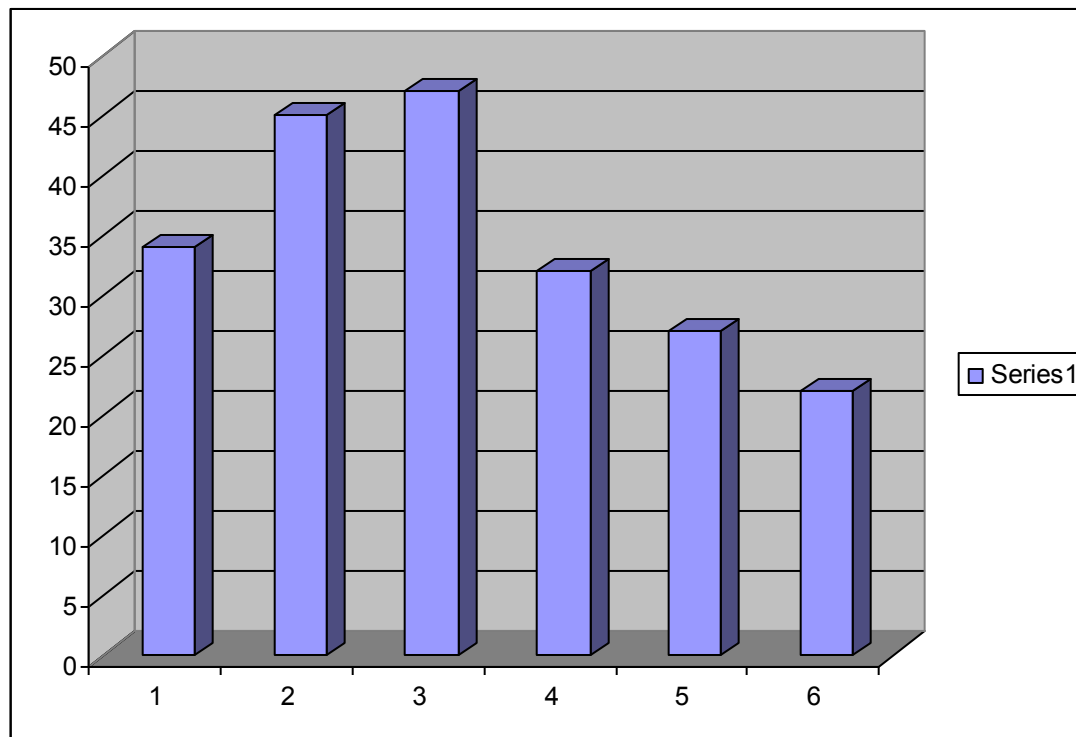


The value of visualization

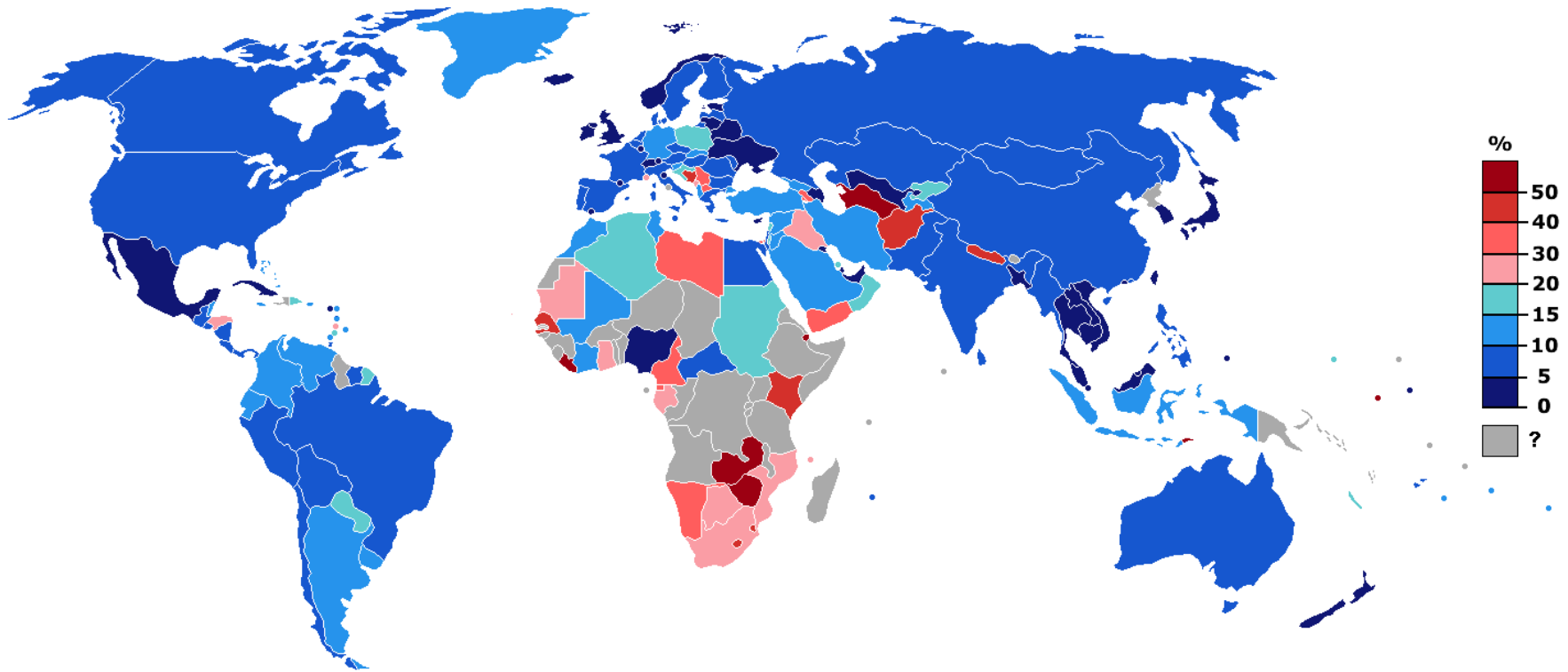
- Capture information
 - Blueprints, photographs, sensors, seismographs, maps... metadata!
- Analyse data to support reasoning
 - Develop and test hypotheses
 - Discover errors
 - Find patterns
 - Expand memory
- Communicate
 - Share, persuade, educate



Excel



Unemployment rates



SCHOOL OF INTERACTIVE ARTS & TECHNOLOGY PRE-REQUISITE MAP

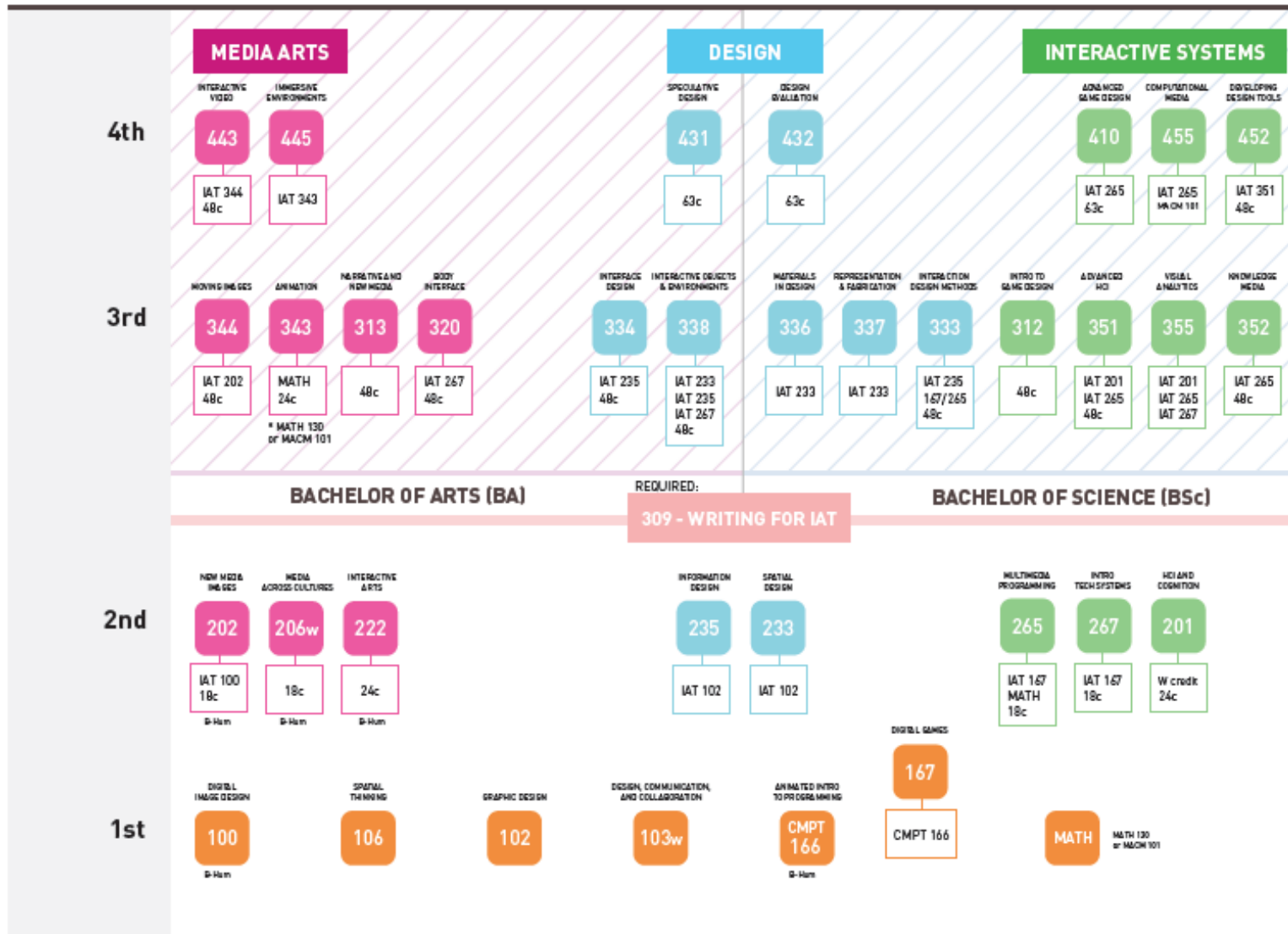
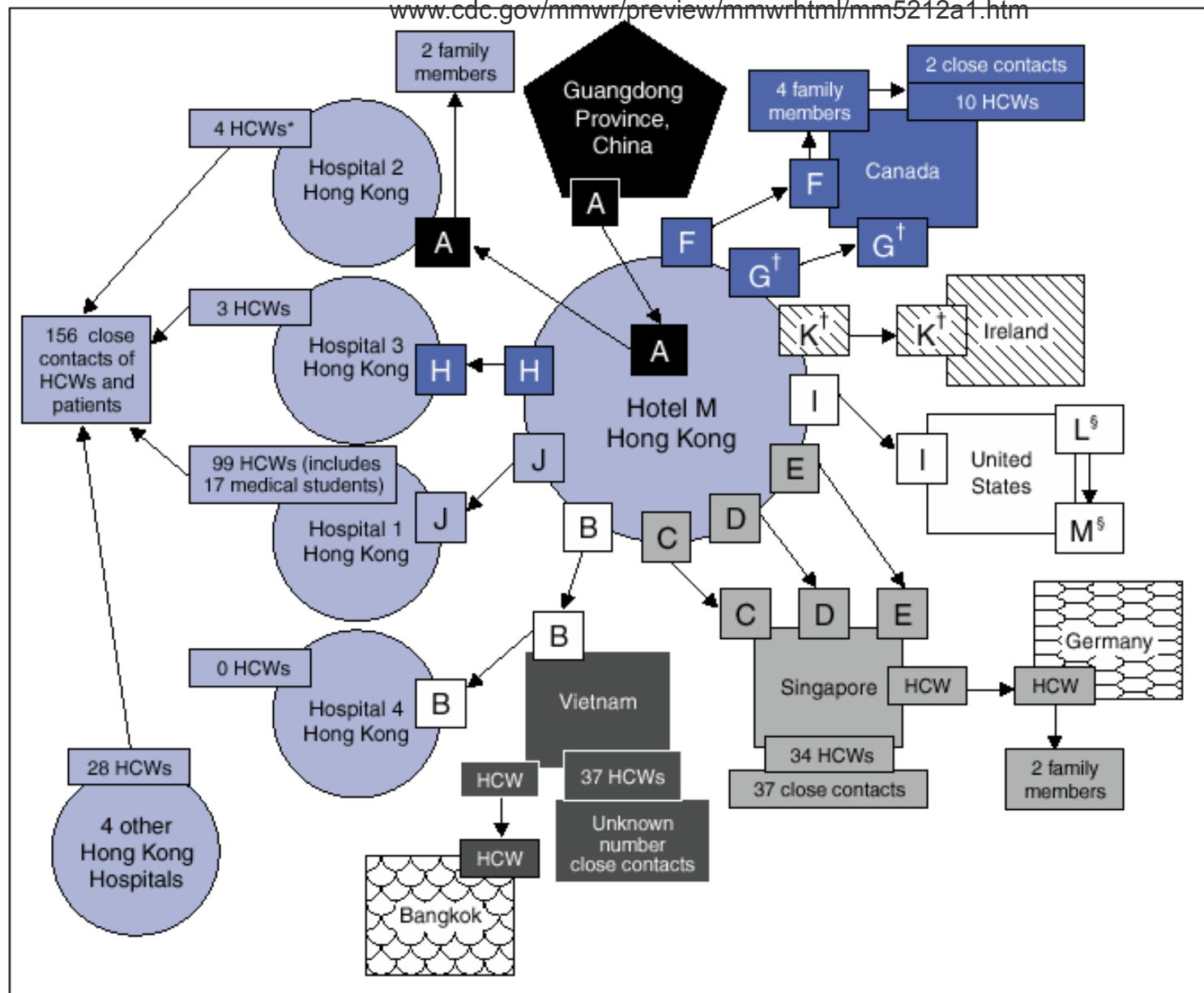


FIGURE 1. Chain of transmission among guests at Hotel M — Hong Kong, 2003

www.cdc.gov/mmwr/preview/mmwrhtml/mm5212a1.htm

SARS



Information Visualization: Examples

[†] All guests except G and K stayed on the 9th floor of the hotel. Guest G stayed on the 14th floor, and Guest K stayed on the 11th floor.

[§] Guests L and M (spouses) were not at Hotel M during the same time as index Guest A but were at the hotel during the same times as Guests G, H, and I, who were ill during this period.

Train Schedule

Piedmont	Palmetto	Caro-linian	Silver Star	Silver Meteor	Crescent	◀ Train Name ▶	Silver Meteor	Crescent	Silver Star	Caro-linian	Palmetto	Pied-mont		
73	89	79	91	97	19	◀ Train Number ▶	98	20	92	80	90	74		
Daily	Daily	Daily	Daily	Daily	Daily	◀ Days of Operation ▶	Daily	Daily	Daily	Daily	Daily	Daily		
						◀ On Board Service ▶								
Read Down						Mile	▼	Symbol	▲	Read Up				
67	67	95/9195	93/83/161	93/83/161		Connecting Train Number				172/164	88/176	194/94	66	66
9 45P	9 45P	6 05A	9 35A	9 35A	0	Dp Boston, MA—South Sta. (ET)	● & [9]	Ar	3 15P	7 05P	9 05P	7 52A	7 52A	
R 9 50P	R 9 50P	R 6 10A	R 9 40A	R 9 40A	1	Boston, MA—Back Bay Sta.	● [9]	▲	D 3 10P	7 00P	D 9 00P	D 7 47A	D 7 47A	
R10 00P	R10 00P	R 6 20A	R 9 51A	R 9 51A	11	Route 128, MA	● [9]	▲	D 3 00P	6 49P	D 8 49P	D 7 30A	D 7 30A	
10 25P	10 25P	6 55A	10 16A	10 16A	43	Providence, RI	● & [9]	▲	2 27P	6 18P	8 19P	7 04A	7 04A	
12 26A	12 26A	8 45A	12 11P	12 11P	156	New Haven, CT	○ [9]	▲	12 36P	4 36P	6 33P	4 55A	4 55A	
1 12A	1 12A	9 28A	12 56P	12 56P	195	Stamford, CT	○ [9]	Ar	11 48A	3 48P	5 48P	4 04A	4 04A	
2 01A	2 01A	10 20A	1 50P	1 50P	231	New York, NY—Penn Sta.	● & [9]	Dp	11 00A		5 00P	3 15A	3 15A	
7 00A	7 00A	2 00P	5 20P	5 20P	456	Washington, DC	● & [9]	Dp	7 25A	11 25A	1 25P	10 00P	10 00P	
6 15A	7 15A	11 08A	3 15P	2 15P	0	Dp New York, NY—Penn Sta.	● & [9]	Ar	10 30A	2 02P	3 43P	8 58P	10 56P	
R 6 31A	R 7 36A	R11 30A	R 3 38P	R 2 37P	10	Newark, NJ—Penn Sta.	● & [9]	▲	D10 10A	D 1 43P	D 3 20P	D 8 37P	D10 35P	
R 7 05A	8 13A	R12 10P	R 4 18P	R 3 15P	58	Trenton, NJ	● & [9]	▲	D 9 30A	D 1 03P	D 2 42P	D 7 55P	D 9 58P	
R 7 36A	8 48A	R12 45P	R 4 55P	R 3 55P	91	Philadelphia, PA—30th St. Sta.	● & [9]	▲	D 8 55A	D12 25P	D 2 07P	D 7 22P	D 9 23P	
R 8 01A	9 11A	R 1 10P	R 5 20P	R 4 19P	116	Wilmington, DE	● & [9]	▲	D 8 30A	D12 01P	D 1 42P	D 6 57P	D 8 57P	
R 8 51A	9 58A	R 2 00P	R 6 15P	R 5 17P	185	Baltimore, MD—Penn Sta.	● & [9]	Ar	D 7 40A	D11 12A	D12 51P	D 6 05P	D 8 06P	
R 9 54A	10 55A	R 3 05P	R 7 30P	R 6 30P	225	Washington, DC	● & [9]	Dp	D 6 32A	D10 10A	D11 53A	D 5 08P	D 7 05P	
10 11A	11 12A	3 25P	R 7 50P	6 49P	234	Charlottesville—see right	● & [9]	Ar	D 5 41A	D 9 43A	D11 09A	4 32P	6 20P	
				7 22P	258	Alexandria, VA	○	▲		8 46A				
				7 55P	293	Manassas, VA	○	▲		8 12A				
				8 52P	338	Culpeper, VA	○	▲		7 20A				
				10 06P	398	Charlottesville, VA	● & [9]	▲		6 07A				
				11 14P	461	Lynchburg, VA	● & [9]	▲		4 57A				
					260	Danville, VA	○	▲				4 01P		
		11 39A			280	Quantico, VA	○	▲						
		11 57A			334	Fredericksburg, VA	○	▲				3 41P		
11 49A	1 02P	5 09P	9 40P		362	Richmond, VA—Staples Mill Rd.	● & [9]	Dp	4 00A		D 9 18A	2 46P	4 40P	
11 59A	1 10P	5 19P	9 50P		476	Petersburg, VA	○ [9]	Dp	3 50A		8 29A	2 39P	4 30P	
12 29P	1 43P	5 57P	10 24P		460	Rocky Mount, NC	○ [9]	Dp	3 07A		7 01A	1 54P	3 51P	
1 56P	3 13P	7 29P	11 56P		476	Wilson, NC	● & [9]	▲	1 43A			12 30P	2 29P	
2 14P	3 32P				502	Selma-Smithfield, NC	○	▲				12 10P	2 11P	
2 41P	4 03P				531	Raleigh, NC	● & [9]	Dp				11 41A	1 39P	
7 05A	4 42P	9 13P			540	Cary, NC	○ [9]	▲		5 40A		11 06A	8 40P	
7 16A	4 50P	9 27P			557	Durham, NC	○ [9]	▲		5 11A		10 58A	8 21P	
7 35A	5 03P				591	Burlington, NC	○	▲				10 43A	8 02P	
8 11A	6 01P				615	Greensboro, NC	● & [9]	Dp				10 23A	7 26P	
8 35A	6 32P			12 15A	615	Winston-Salem—see right	● & [9]	Dp				9 43A	7 05P	
8 52A	6 49P			12 22A	628	High Point, NC	○ [9]	▲		3 58A		9 21A		
9 26A	7 26P			12 39A	628	Salisbury, NC	○ [9]	▲		3 30A		8 58A	6 45P	
9 42A	7 42P			1 17A	662	Kannapolis, NC	○ [9]	▲		2 49A		8 25A	6 13P	
10 14A	8 14P				677	Charlotte, NC	○	▲				8 07A	5 57P	
	3 27P			2 20A	704	Fayetteville, NC	● & [9]	Dp			2 03A	7 40A	5 30P	
	4 14P			1 29A	550	Dillon, SC	● & [9]	Ar		11 59P			12 54P	
	5 06P			3 20A	603	Florence, SC (Myrtle Beach)	○	Ar					12 05P	
	5 11P			3 28A	633	Kingstree, SC	○	Ar		10 41P			11 31A	
	5 47P			4 05A	672	Charleston, SC	○	Ar		10 31P			11 26A	
	6 46P			5 06A	728	Yemassee, SC	● & [9]	▲		9 38P			10 40A	
	7 35P			5 56A	782	Southern Pines, NC (Pinehurst)	○	▲		8 44P			9 44A	
					599	Hamlet, NC	○	▲		7 48P			8 48A	
		10 39P			628	Camden, SC	○	▲			4 02A			
		11 21P			701	Columbia, SC	○	▲			3 25A			
		12 50A			734	Denmark, SC	○	▲				1 12A		
		2 41A			784	Savannah, GA	● & [9]	Dp				11 57P		
	8 34P	4 29A	6 44A		829	Jesup, GA (Brunswick)	● & [9]	Dp		6 59P		10 34P	8 00A	
		4 34A	6 50A		*871		○	Ar		6 53P		10 28P		
			7 44A		881		○	Ar		5 59P				

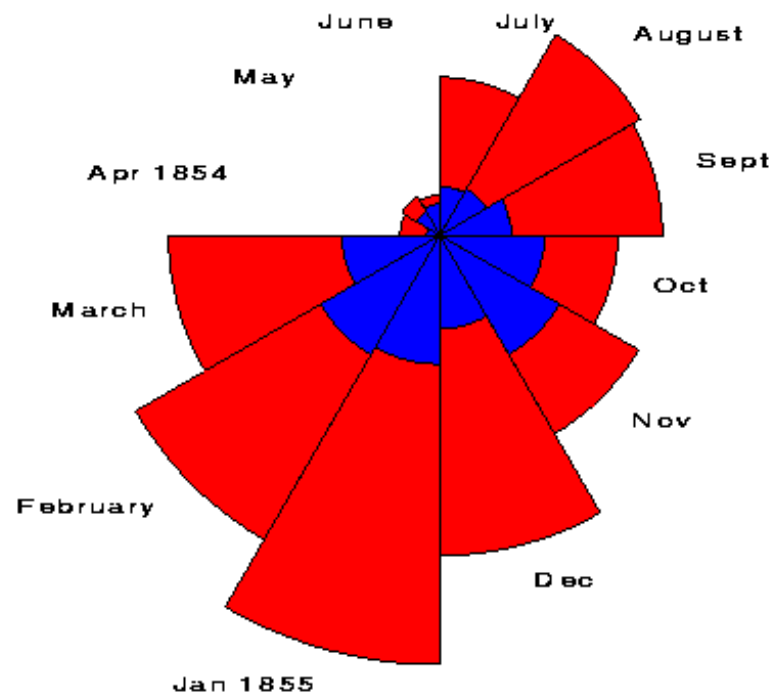
Continuation of service to/from Florida continues at right



Nightingale's Coxcomb

Causes of Mortality in the Army in the East
April, 1854 to March 1855

■ Non-Battle
■ Battle



From: F. Nightingale, "Notes on Matters Affecting the Health, Efficiency and Hospital Administration of the British Army", 1858

Napoleon's Invasion of Russia

Carte Figurative des pertes successives en hommes de l'Armée Française dans la campagne de Russie 1812-1813.
 Dessiné par M. Minard, Inspecteur Général des Ponts et Chaussées en retraite. Paris, le 20 Novembre 1869.

Les nombres d'hommes présents sont représentés par les largueurs des zones colorées à raison d'une millimètre pour dix mille hommes; ils sont de plus écrits en lettres des zones. Le rouge désigne les hommes qui restent en Russie, le noir ceux qui en sortent. Les renseignements qui ont servi à dresser la carte ont été puisés dans les ouvrages de M. M. Chiers, de Chéjar, de Fozzardac, de Chambray et le journal inédit de Jacob, pharmacien de l'Armée depuis le 23 Octobre. Pour mieux faire juger à l'œil la diminution de l'armée, j'ai supposé que les corps de Prince Sviatou et du Maréchal Davoust qui avaient été détachés sur Minsk et Mohilew et qui rejoignent l'armée, ont toujours marché avec l'armée.

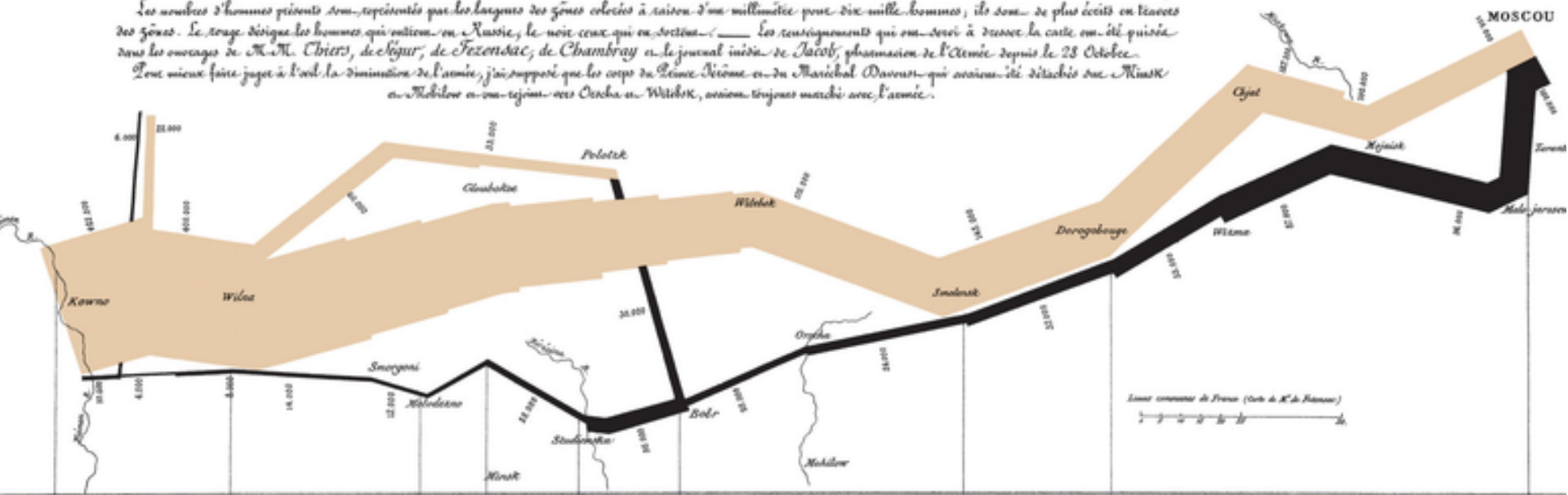
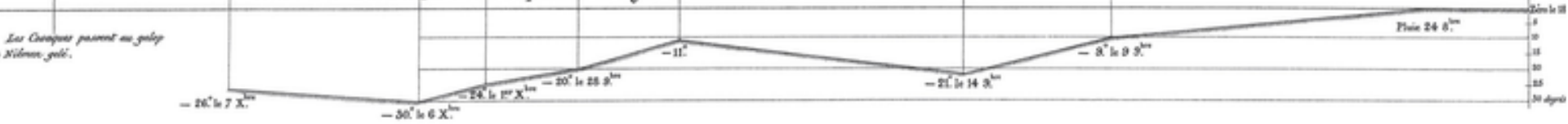


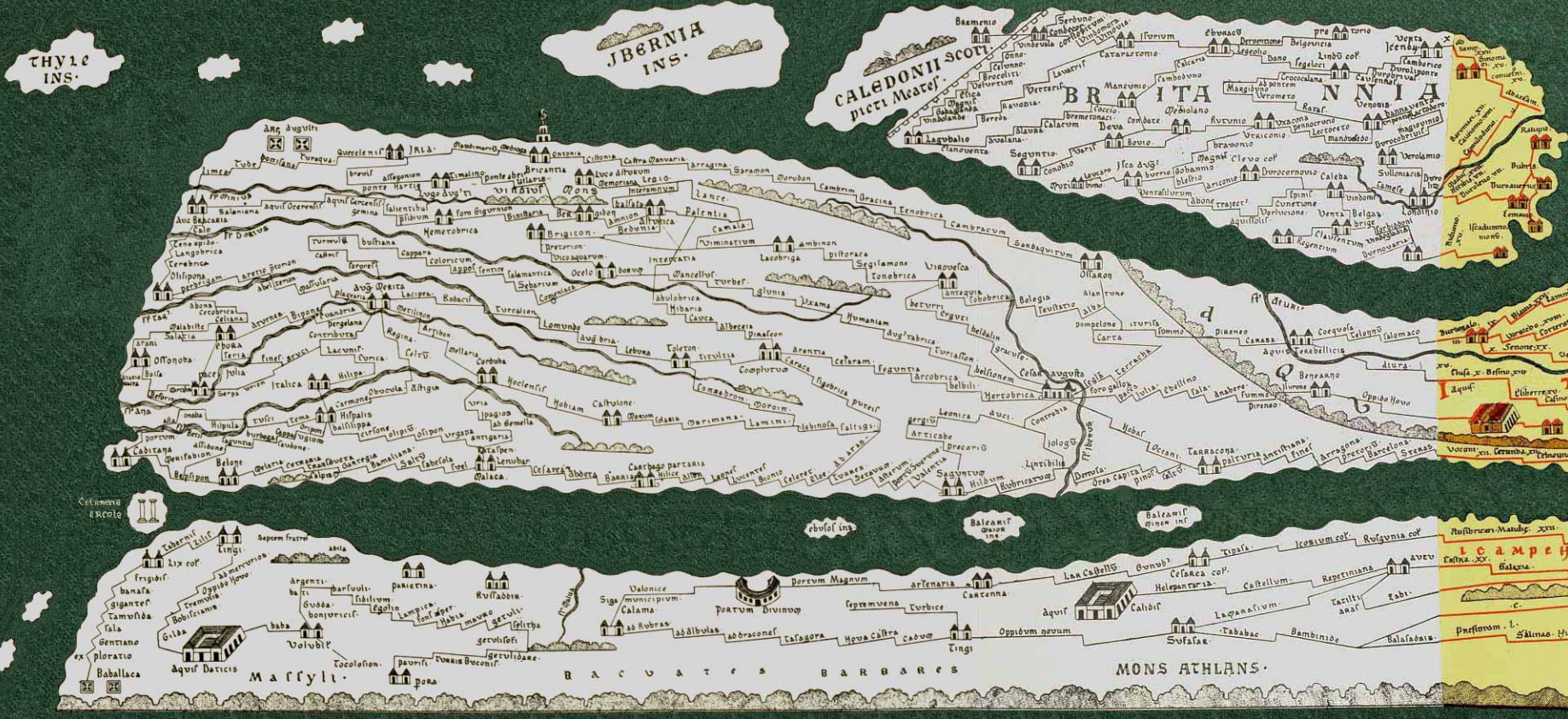
TABLEAU GRAPHIQUE de la température en degrés du thermomètre de Réaumur au dessous de zéro.



Mediaeval Europe

Tabula Peutingeriana, road map of ancient Rome

Segmentum I

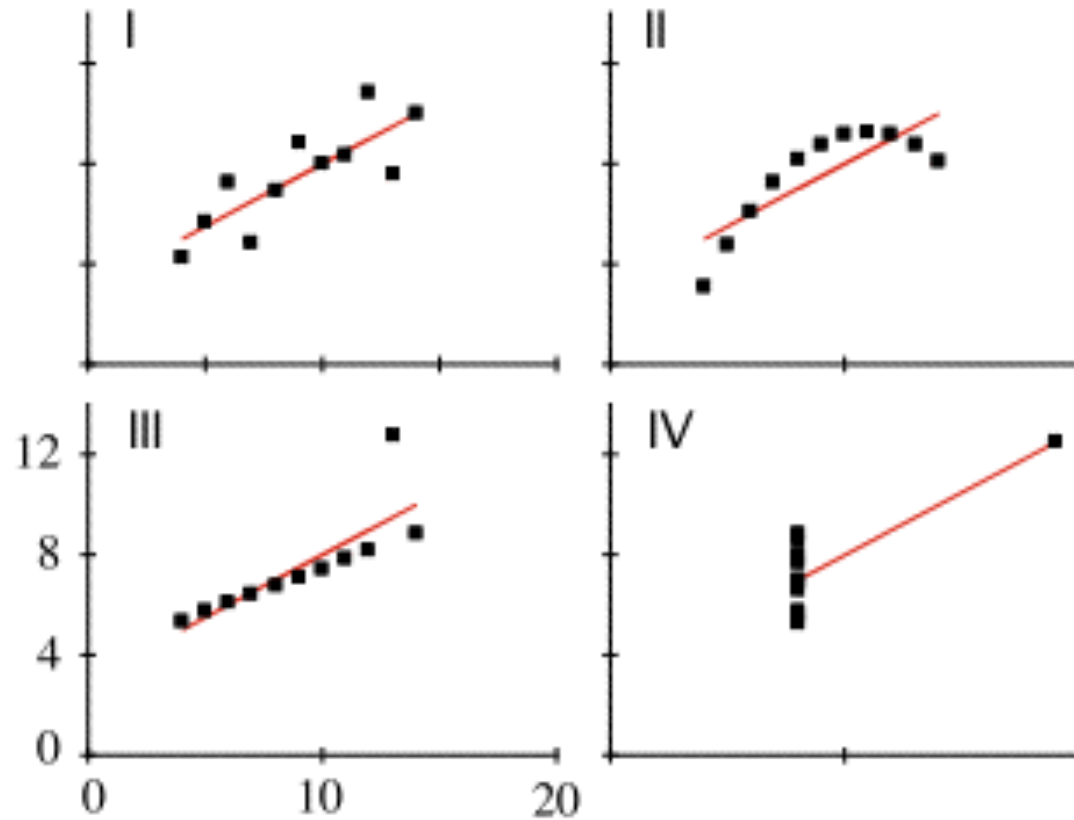


-
- Portray data, usually abstract data
 - Use visual features to represent properties, quantities, attributes
 - Explicitly
 - derived
 - Give rise to *emergent features*

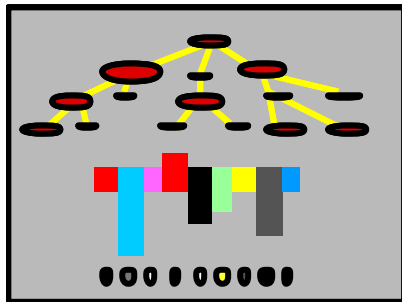
Review: why it helps

- Mean of the x values = 9.0
- Mean of the y values = 7.5
- Equation of the least-squared regression line: $y = 3 + 0.5x$
- Sums of squared errors (about the mean) = 110.0
- Regression sums of squared errors (variance accounted for by x) = 27.5
- Residual sums of squared errors (about the regression line) = 13.75
- Correlation coefficient = 0.82
- Coefficient of determination = 0.67

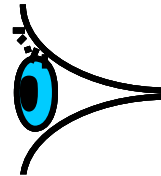
What the data look like ...



Data



We look at
that picture



Ah HA !!

and gain
insight

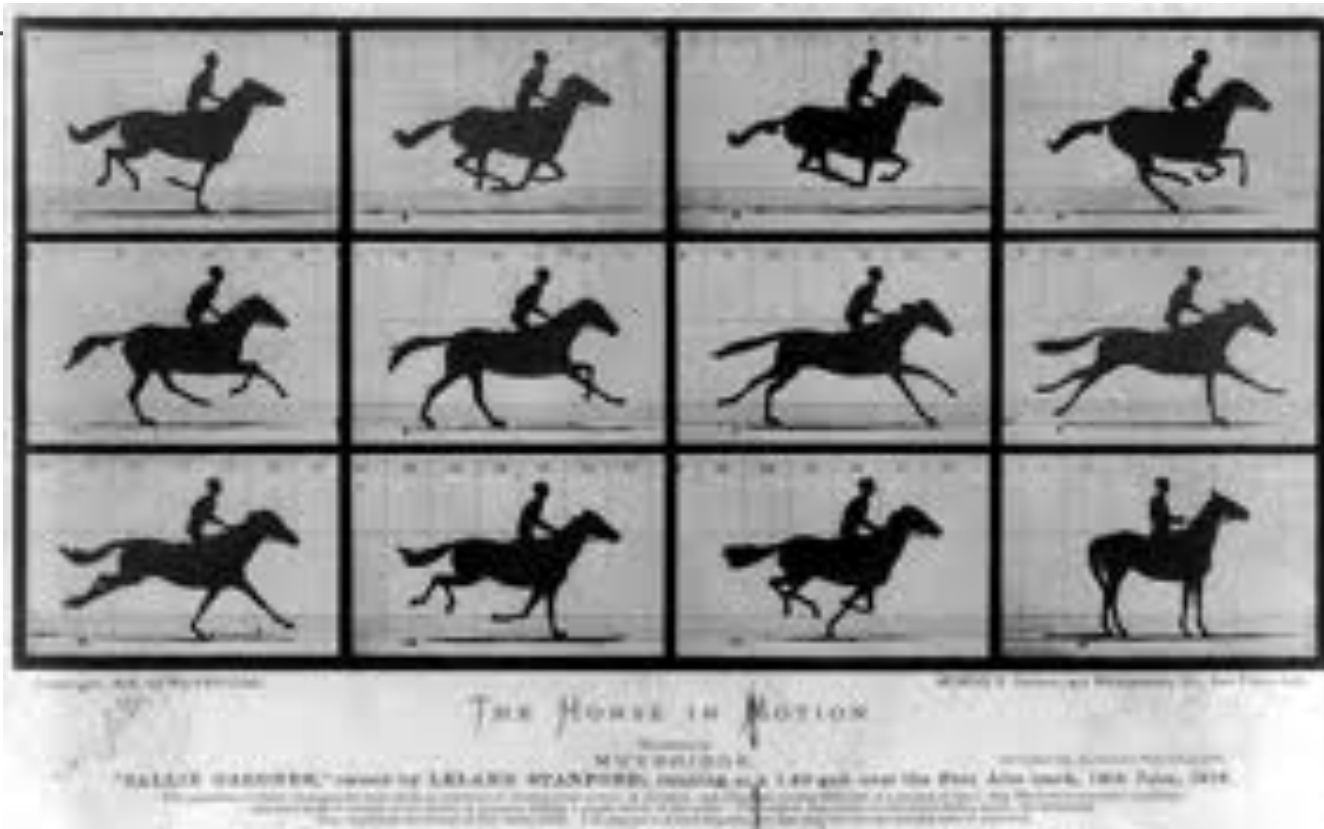
Information visualization

The process of information visualization. Graphically encoded data is viewed in order to form a **mental model** of that data

Recall: Why do we create visualizations

- Answer questions (or discover them)
- Make decisions
- See data in context
- Expand memory
- Support graphical calculation
- Find patterns
- Present argument or tell a story
- Inspire

Organize and capture information



Combine different data sources

evand.name/a-to-cedar/

A 1856 E 61st Ave
 B Clark Drive & E 21st A
 Get Directions

Possible Routes

Argyle & Dumfries 4.9km, 21 mins
 Districts: Frasersview, Kensington, Kingsway

Inverness & Windsor 5.7km, 27 mins
 Districts: Frasersview, Kensington

Directions
 Via Argyle & Dumfries

- Head west on E 61st Ave toward Thornhill Dr
- Turn right onto Thornhill Dr

Bike Routes

- off-street
- neighbourhood bikeway
- bike lane
- shared roadway
- regular street (not an official bike route)

Hills

- gentle
- steep

Map data © OpenStreetMap Contributors. More credits.

Neighbourhoods

- residential
- commercial
- industrial

Map data © OpenStreetMap Contributors. More credits.



Information Visualizat

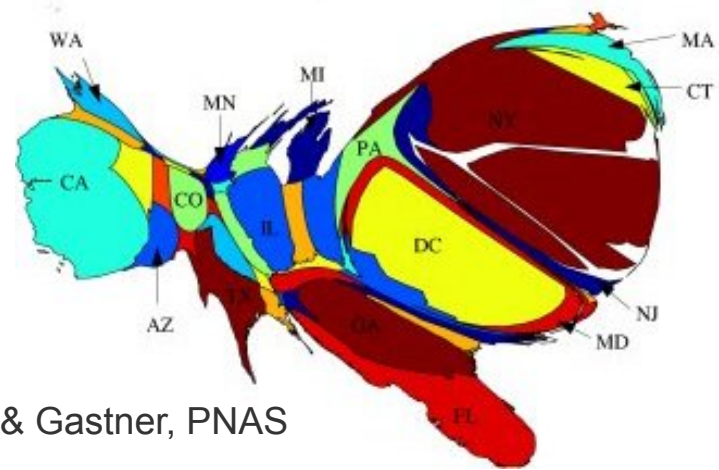
- Turn right onto Spadina Dr
- Turn left onto Braeburn St
- Turn left onto E 56th Ave

07.19.2012_BikeBrow....pdf | naxis-presentation.pdf Cancelled | TableauDesktop.exe

Show All

Answer (and discover) questions

- Where do the most news stories originate?
- Inspire!



Newman & Gastner, PNAS

Make decisions/expand memory

- Train schedule

Piedmont	Palmetto	Carolinian	Silver Star	Silver Meteor	Crescent	◀ Train Name ▶	Silver Meteor	Crescent	Silver Star	Carolinian	Palmetto	Piedmont	
73	89	79	91	97	19	◀ Train Number ▶	98	20	92	80	90	74	
Daily	Daily	Daily	Daily	Daily	Daily	◀ Days of Operation ▶	Daily	Daily	Daily	Daily	Daily	Daily	
☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	◀ On Board Service ▶	☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	☐☐☐☐☐	
Read Down						Mile	Symbol	Read Up					
67	67	95/195	93/161	93/161	93/161	Connecting Train Number	172/164	88/176	194/94	66	66		
9 45P	9 45P	6 05A	9 35A	9 35A	0	Boston, MA-South Sta. (ET)	● & [9] Ar	3 15P	7 05P	9 05P	7 52A	7 52A	
R 9 50P	R 9 50P	R 6 10A	R 9 40A	R 9 40A	1	Boston, MA-Back Bay Sta.	● [9] D	D 3 10P	7 00P	D 9 00P	D 7 47A	D 7 47A	
R10 00P	R10 00P	R 6 20A	R 9 51A	R 9 51A	11	Route 128, MA	● [9] D	D 3 00P	6 49P	D 8 49P	D 7 30A	D 7 30A	
10 25P	10 25P	6 55A	10 16A	10 16A	43	Providence, RI	● & [9] Ar	2 27P	6 18P	8 19P	7 04A	7 04A	
12 26A	12 26A	8 45A	12 11P	12 11P	158	New Haven, CT	○ [9] Ar	12 36P	4 36P	6 33P	4 55A	4 55A	
1 12A	1 12A	9 28A	12 56P	12 56P	195	Stamford, CT	○ [9] Ar	11 48A	3 48P	5 48P	4 04A	4 04A	
2 01A	2 01A	10 20A	1 50P	1 50P	231	New York, NY-Penn Sta.	● & [9] Dp	11 00A	5 00P	3 15A	3 15A	3 15A	
7 00A	7 00A	2 00P	5 20P	5 20P	456	Washington, DC	● & [9] Dp	7 25A	11 25A	1 25P	10 00P	10 00P	
6 15A	7 15A	11 08A	3 15P	2 15P	0	New York, NY-Penn Sta.	● & [9] Ar	10 30A	2 02P	3 43P	8 58P	10 56P	
R 6 31A	R 7 36A	R11 30A	R 3 38P	R 2 37P	10	Newark, NJ-Penn Sta.	● & [9] Ar	D10 10A	D 3 20P	D 3 42P	D 8 37P	D10 35P	
R 7 05A	8 13A	R12 10P	R 4 18P	R 3 15P	58	Trenton, NJ	○ [9] Ar	D 9 30A	D 1 03P	D 2 42P	D 7 56P	D 9 58P	
R 7 36A	8 48A	R12 45P	R 4 56P	R 3 55P	91	Philadelphia, PA-30th St. Sta.	● & [9] Ar	D 8 55A	D12 05P	D 2 07P	D 7 22P	D 9 23P	
R 8 01A	9 11A	R 1 10P	R 5 20P	R 4 19P	116	Wilmington, DE	● & [9] Ar	D 8 30A	D12 01P	D 2 42P	D 6 57P	D 8 57P	
R 8 51A	9 58A	R 2 00P	R 6 15P	R 5 17P	185	Baltimore, MD-Penn Sta.	● & [9] Ar	D 7 40A	D11 12A	D12 51P	D 6 05P	D 8 06P	
R 9 54A	10 55A	R 3 05P	R 7 30P	R 6 30P	225	Washington, DC	● & [9] Dp	D 6 32A	D10 10A	D11 53A	5 08P	D 7 05P	
10 11A	11 12A	3 25P	R 7 50P	6 49P	234	Charlottesville—see right	● & [9] Ar	D 5 41A	D 9 43A	D11 05A	4 32P	6 20P	
				7 22P	258	Alexandria, VA	● & [9] Ar						
				7 59P	293	Manassas, VA	○			8 48A			
				8 59P	338	Culpeper, VA	○			8 12A			
				10 06P	398	Charlottesville, VA	● & [9] Ar			7 20A			
				11 14P	461	Lynchburg, VA	● & [9] Ar			6 07A			
					260	Danville, VA	○			4 57A			
					280	Quantico, VA	○				4 01P		
					280	Fredericksburg, VA	○				3 41P		
					334	Richmond, VA-Staples Mill Rd.	● & [9] Dp	4 00A		D 9 18A	2 46P	4 40P	
					362	Petersburg, VA	○ [9] Dp	3 07A		2 39P	4 30P		
					460	Rocky Mount, NC	● & [9] Ar	1 43A		7 01A	1 54P	3 51P	
					476	Wilson, NC	● & [9] Ar				12 10P	2 11P	
					502	Selma-Smithfield, NC	○				11 41A	1 39P	
					531	Raleigh, NC	● & [9] Dp			5 40A	11 05A	8 40P	
					540	Cary, NC	○ [9] Ar			5 11A	10 43A	8 21P	
					557	Durham, NC	● & [9] Ar				10 23A	8 02P	
					591	Burlington, NC	○				9 43A	7 26P	
					615	Greensboro, NC	● & [9] Dp			3 58A	9 21A	7 05P	
					628	Winston-Salem—see right	○ [9] Ar			3 51A			
					662	High Point, NC	○ [9] Ar			3 30A	8 58A	6 45P	
					677	Salisbury, NC	○ [9] Ar			2 49A	8 25A	6 13P	
					704	Kannapolis, NC	○			8 07A	5 57P	5 57P	
					704	Charlotte, NC	● & [9] Dp			2 03A	7 40A	5 30P	
					550	Fayetteville, NC	● & [9] Ar			11 59P		12 54P	
					603	Dillon, SC	○					12 05P	
					633	Florence, SC (Myrtle Beach)	● & [9] Dp			10 41P		11 31A	
					672	Kingstree, SC	○			10 31P		11 26A	
					728	Charleston, SC	● & [9] Ar			9 38P		10 40A	
					782	Yemassee, SC	○ [9] Ar			8 44P		9 44A	
					599	Southern Pines, NC (Pinehurst)	○				4 02A	8 48A	
					628	Hamlet, NC	○				3 25A		
					701	Camden, SC	○				1 53A		
					734	Columbia, SC	● & [9] Ar				1 12A		
					784	Denmark, SC	○				11 57P		
					829	Savannah, GA	● & [9] Dp			6 59P	10 34P	8 00A	
					871	(Yemassee via Columbia)	○ [9] Dp			6 53P	10 26P		
					881	Jesup, GA (Brunswick)	○ Ar			5 59P			

Continuation of service to/from Florida continues at right

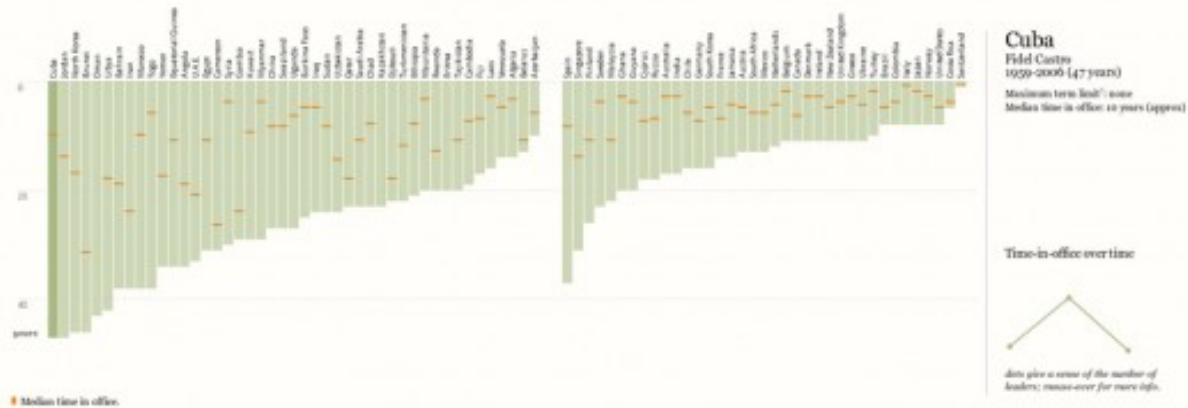


Information Visualization

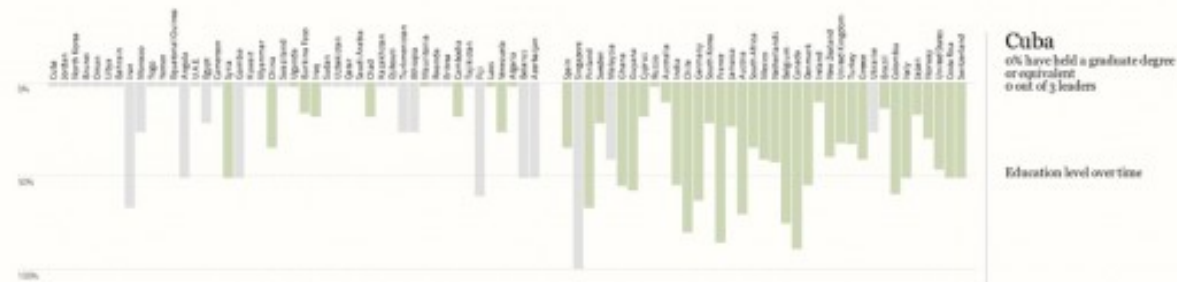
See data in context/expand graphical calculation

How educated are world leaders?

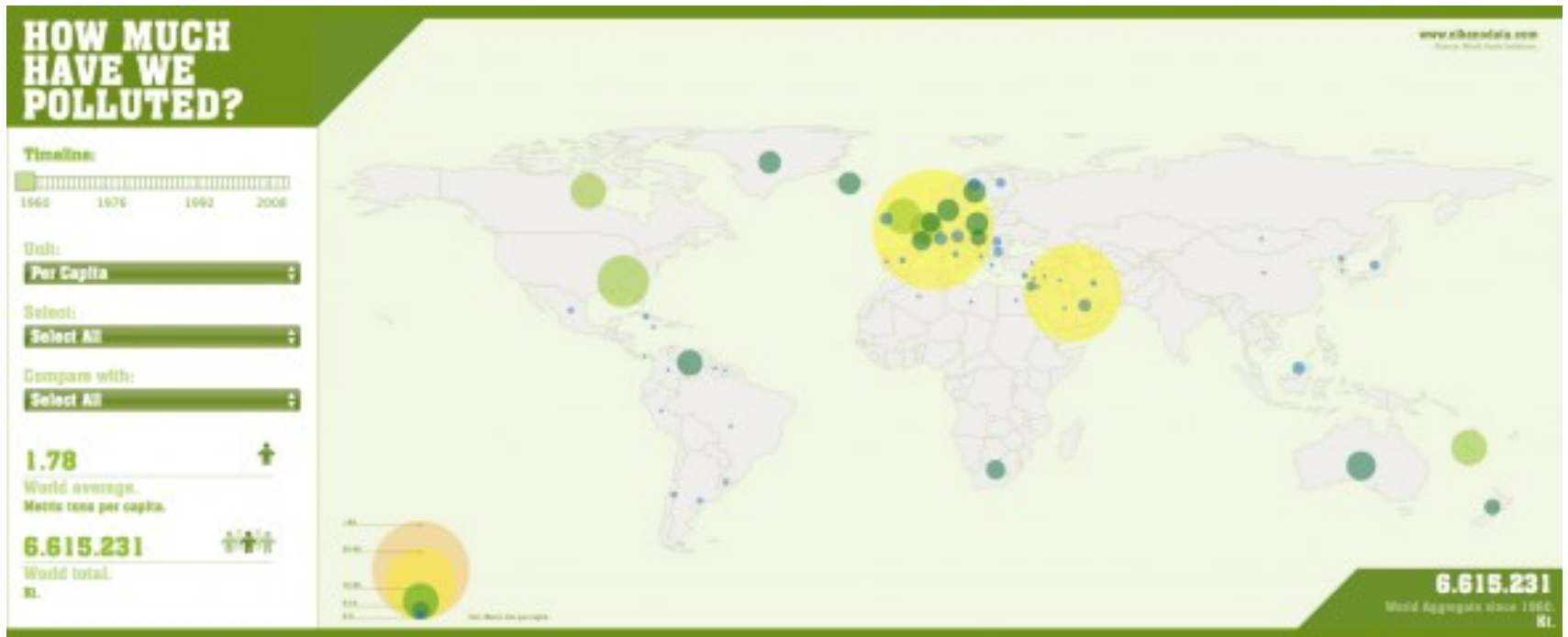
In democracies¹, the driver of change is typically elections; in autocracies, it is typically biology. It is often said that autocratic leaders are therefore less likely to be educated. Leaders' education level has been found to be a positive predictor of, among other things, economic growth². Here's a look at the tenure of around 700 leaders³ sorted by political longevity in the world's autocratic and anocratic countries⁴, along with a selection from present-day democracies⁵, since 1950 or thereabouts. For each country, the leader who has ruled the longest, contiguously, is shown. Hover over a bar for more details.



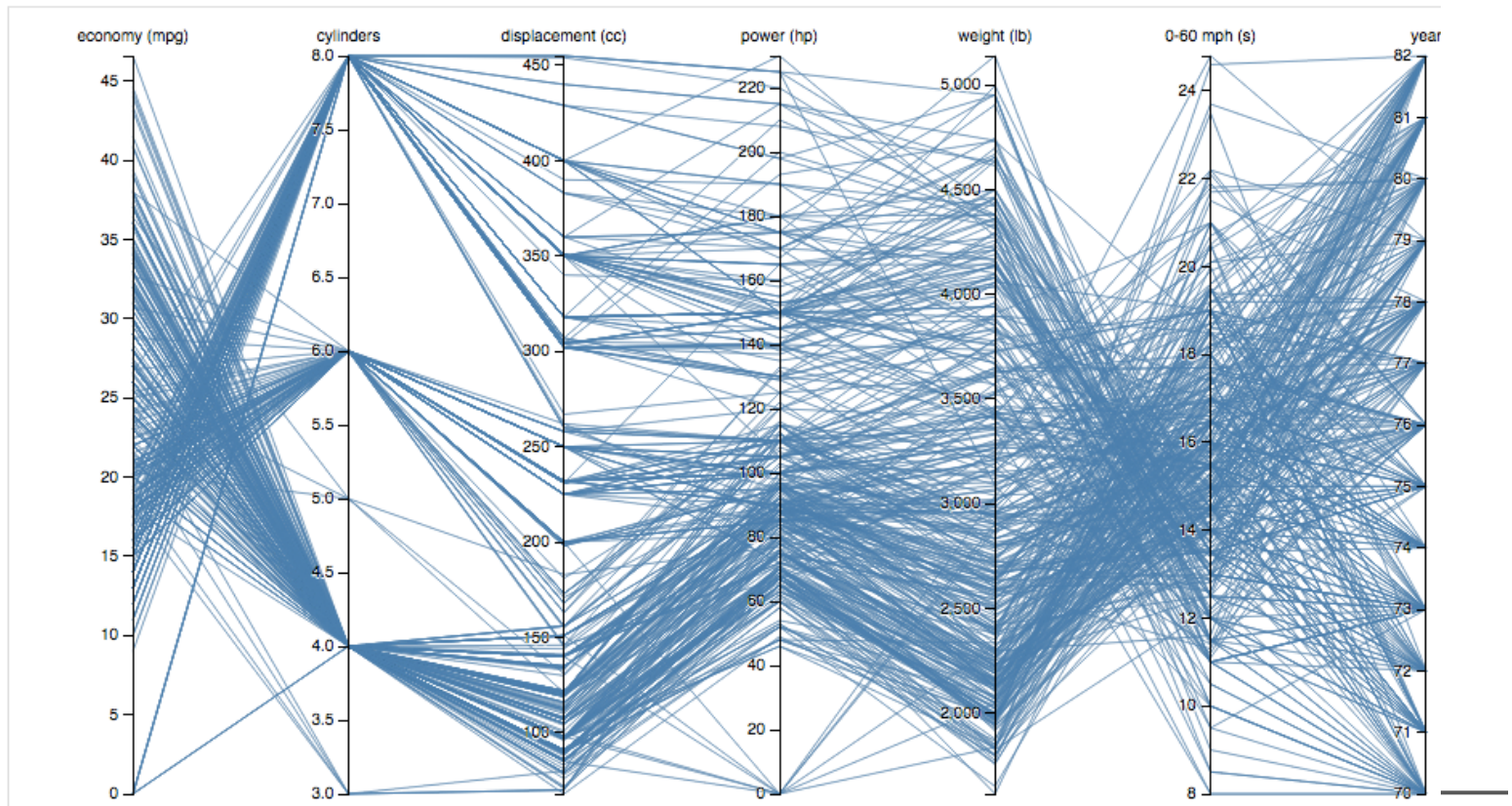
Here's a look at the number of leaders with an advanced degree, e.g. Master's or Doctorate, or equivalent - such as certain law degrees in some countries - for each of the above countries during the same time period. Lack of education is not necessarily indicative of ineptness, though it may be. It may also be the result of extenuating life circumstances or lack of means, as in the case of former British Prime Minister James Callaghan, for example.



find patterns, compare



Find patterns, expand calculations, find questions



Inspire and compel/tell a story

Gun ownership in Westchester County, USA

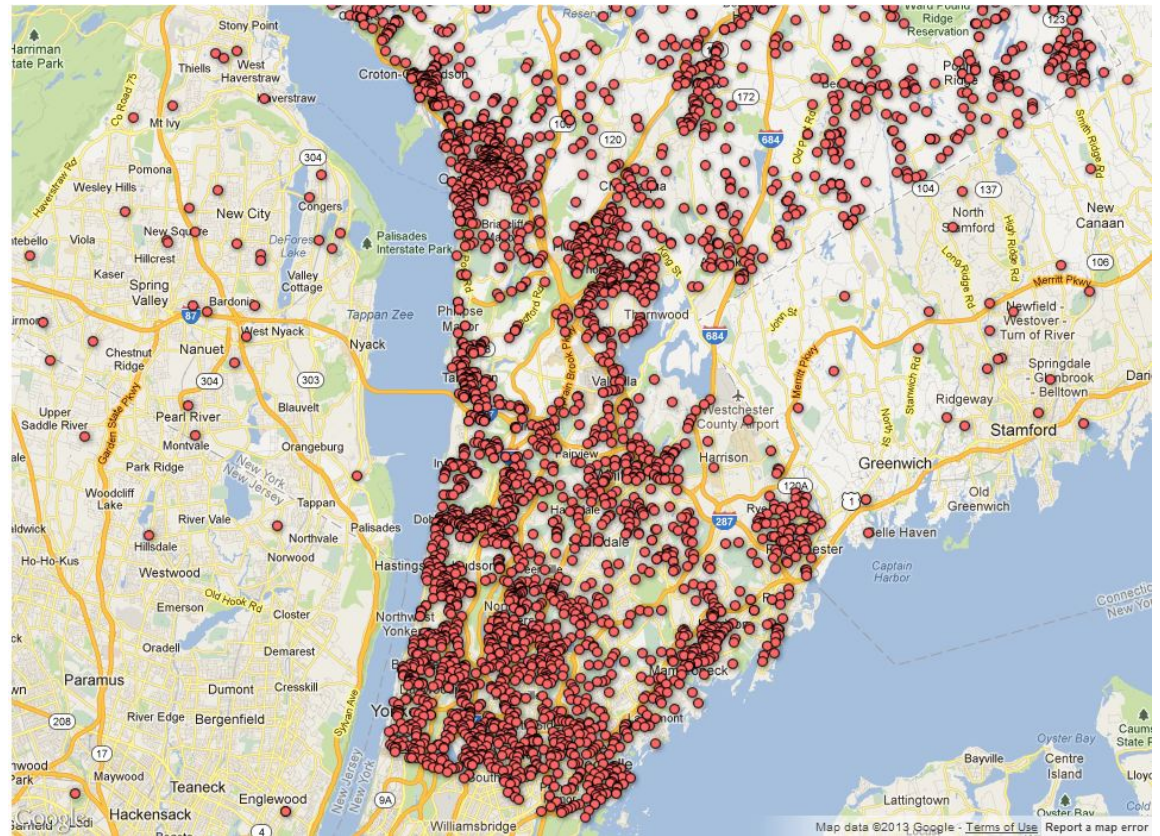


Image courtesy of S. Few, www.perceptualedge.com

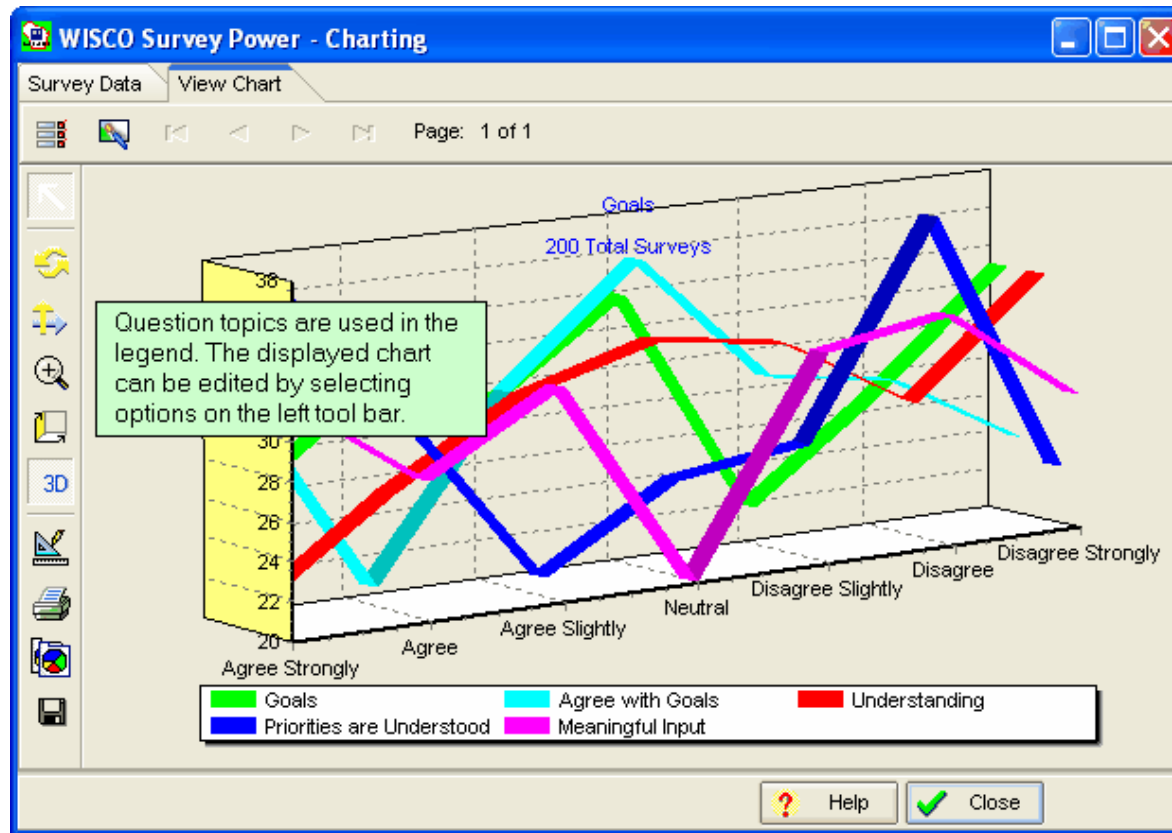


Image courtesy of S. Few, www.perceptualedge.com

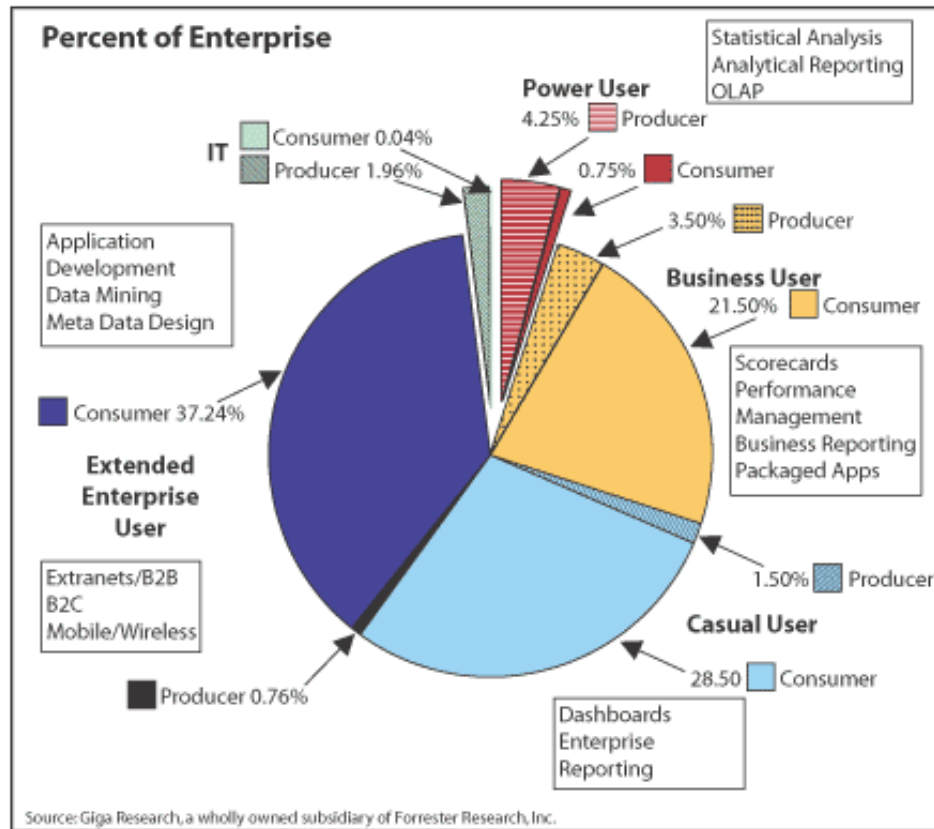
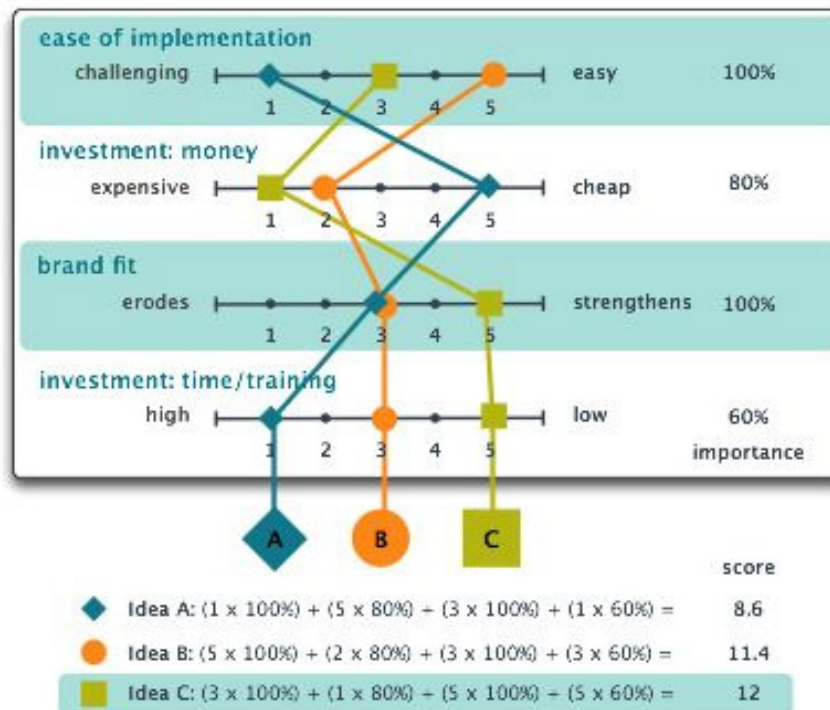


Image courtesy of S. Few, www.perceptualedge.com



Using this advanced method I was able to confirm my initial assessment that Idea C would be our best bet.

What's really going on here?

- The purpose of visualization is INSIGHT not IMAGES (Stasko)
- Cognitive process of building a mental image and model and internalising understanding

“The use of computer-supported, interactive visual representations of data to amplify cognition.” [Card, Mackinlay Shneiderman '98]

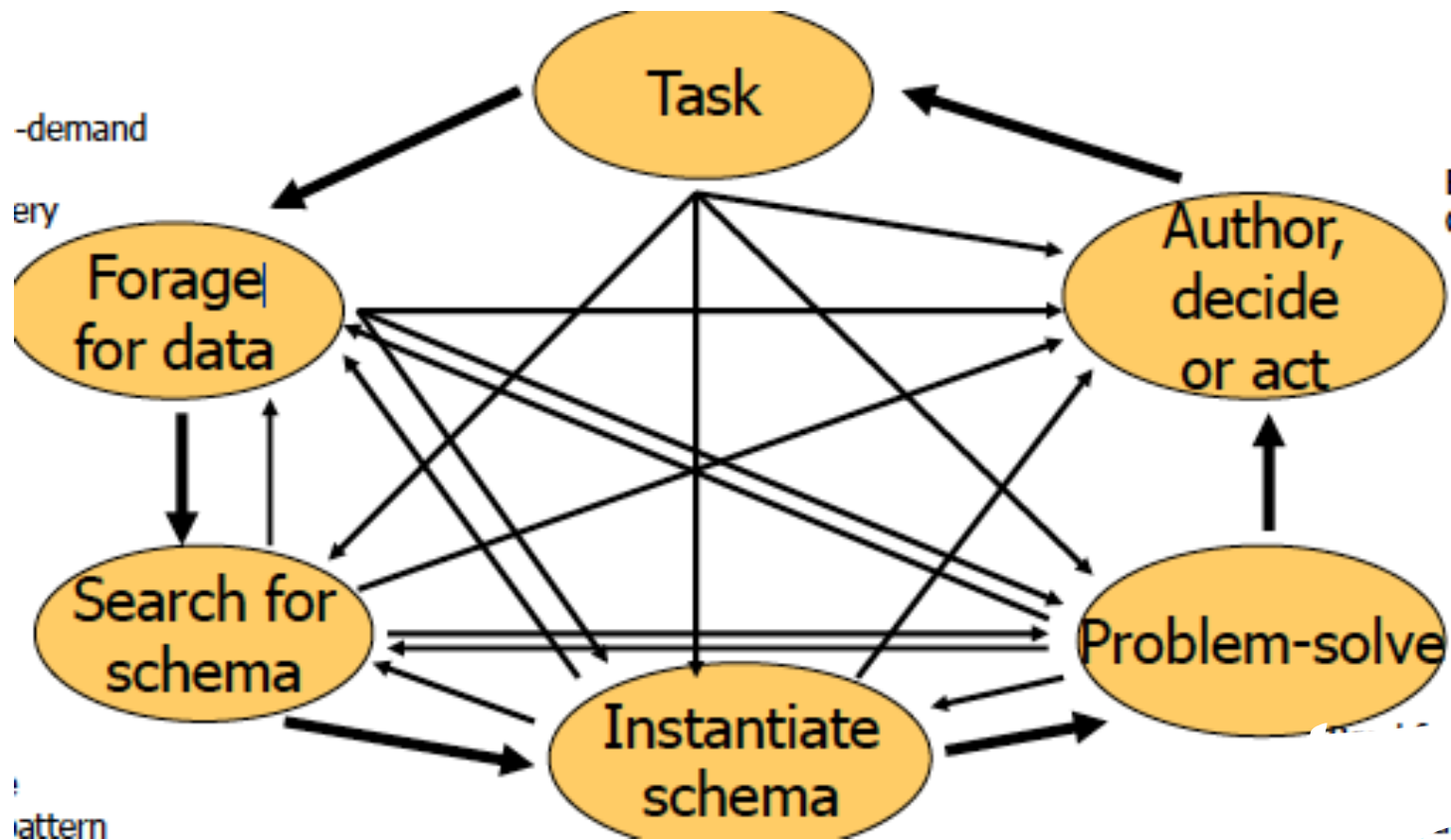
Distributed Cognition

- Cognitive system is composed of people and the artifacts they use
- Cognition isn't only internal
- Changes in **external** representation spur changes in **internal** representation and understanding
- It is **interaction** with the external representations that drives this process

Recap: how vis amplifies cognition

- Increasing memory and processing resources available
- Reducing search for information
- Enhancing the recognition of patterns
- Enabling perceptual inference operations
- Using perceptual attention mechanisms for monitoring
- Encoding info in a manipulable medium

Knowledge Crystallization

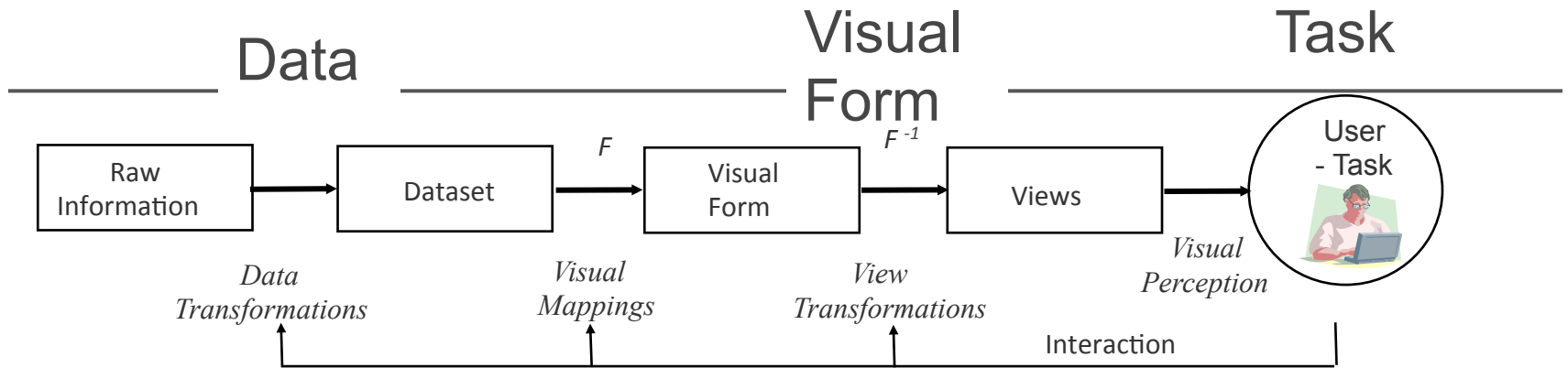


More simply

Acquire → Parse → Filter → Mine/Prune → Represent → Refine → Interact



Visualization Stages

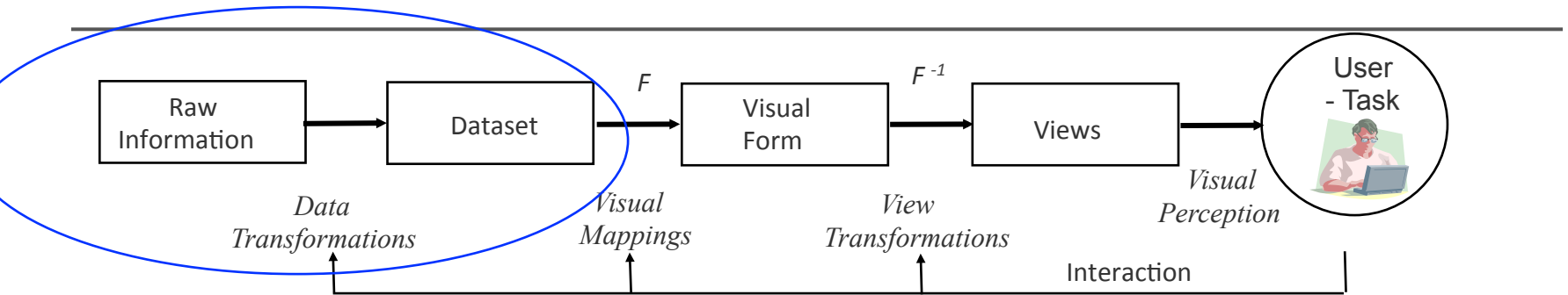


Example: house real estate listing data

- Price
- Bedrooms
- Lot size
- Type
- etc

Visualization Stages

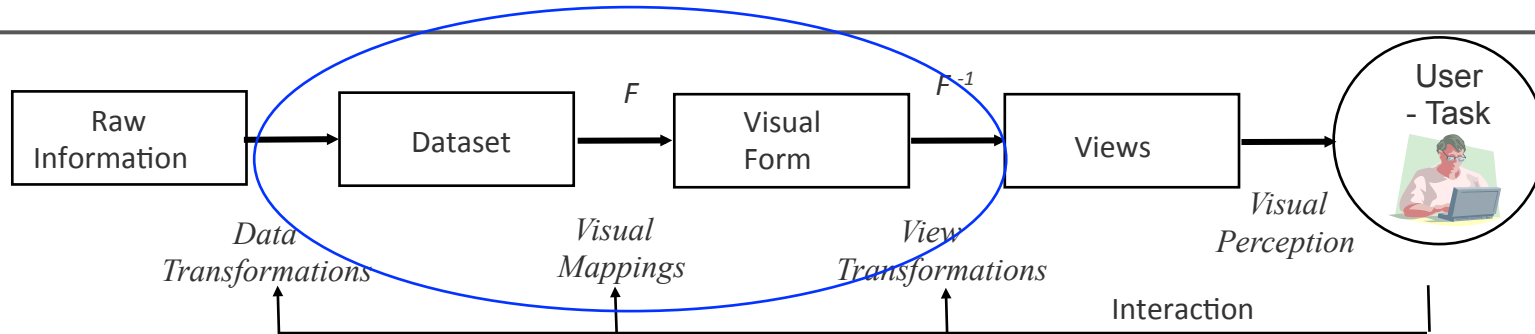
Data transformation – create a visual spatial model



- Data transformation
 - Map raw data into data tables – e.g. text to similarity matrix

Visualization Stages

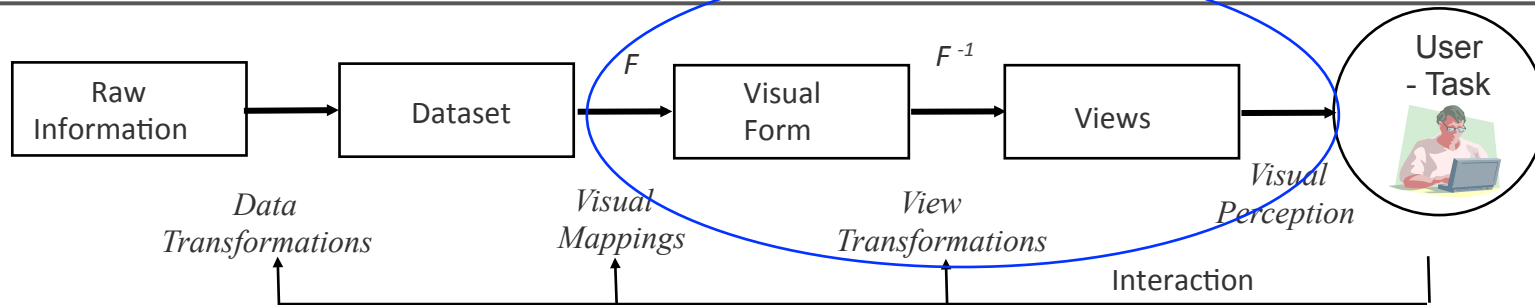
Visual mapping– create a visual spatial model



- Data transformation
 - Map raw data into data tables – e.g. text to similarity matrix
- Visual Mappings:
 - Transform data tables into visual structures – e.g., house price, #bedrooms to 2 dims – x, y

Visualization Stages

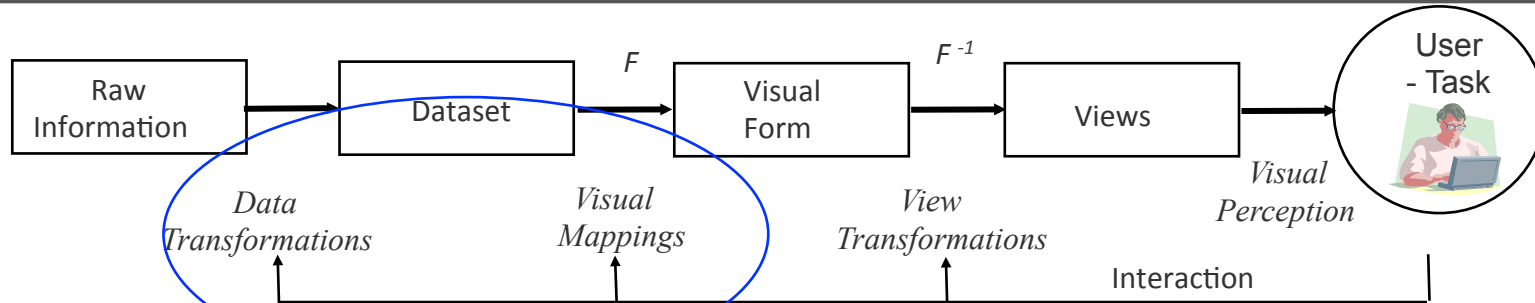
Display the data that now have visual form



- Data transformation
 - Map raw data into data tables – e.g. text to similarity matrix
- Visual Mappings:
 - Transform data tables into visual structures – e.g. 2 dims – x, y
- View Transformations:
 - Create views of the Visual Structures by specifying graphical parameters such as position, scaling, and clipping

Visualization Stages

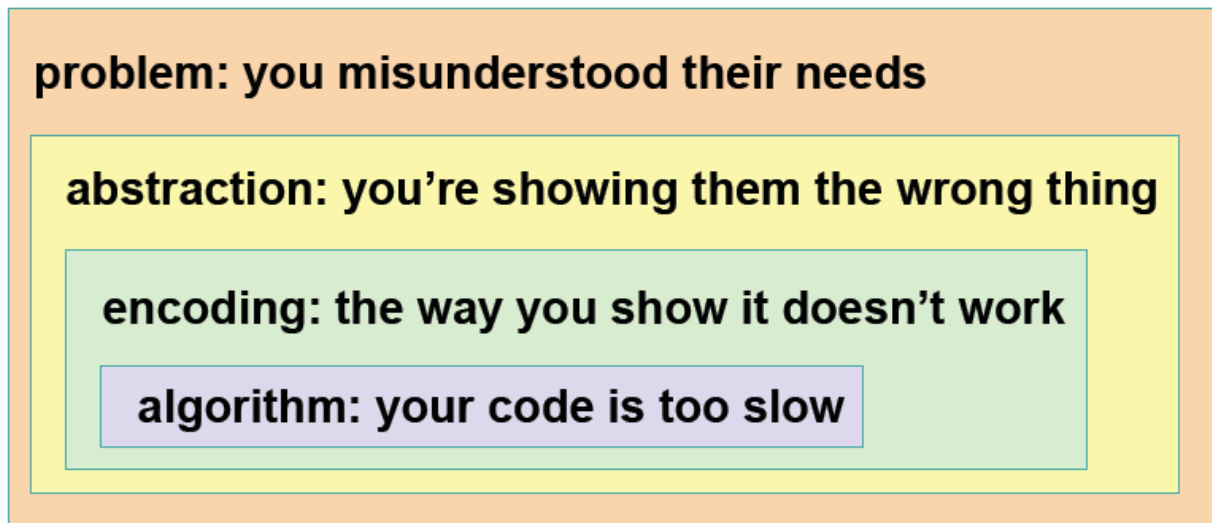
The user may change transformations and mappings



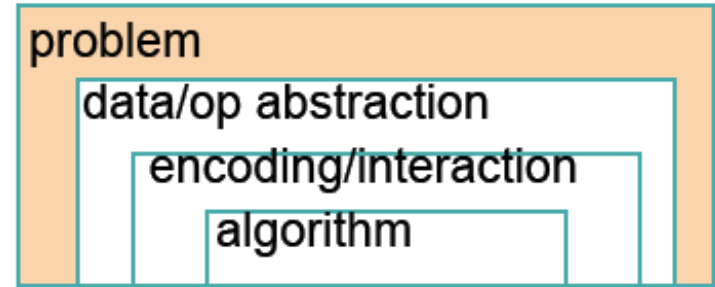
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How do we find the right design?

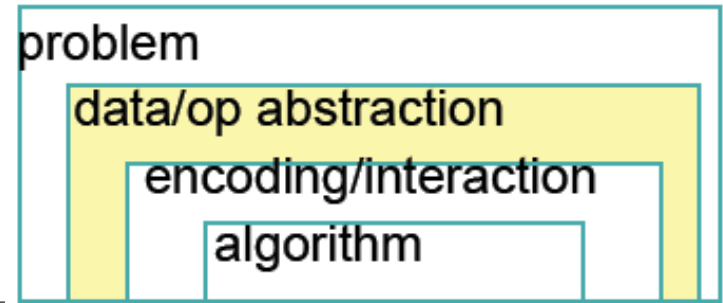
- 4 levels of design [Munzner 2014]
- Validate against the right “threat”



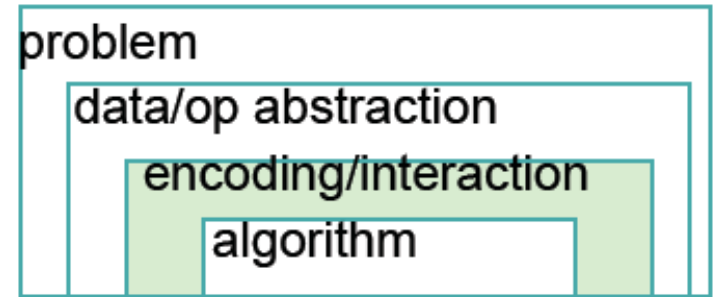
Solve the right problem



- Characterise the domain problem
 - identify a problem amenable to vis
 - provide novel capabilities
 - speed up existing workflow
- validation
 - immediate: interview and observe target users
 - downstream: notice adoption rates



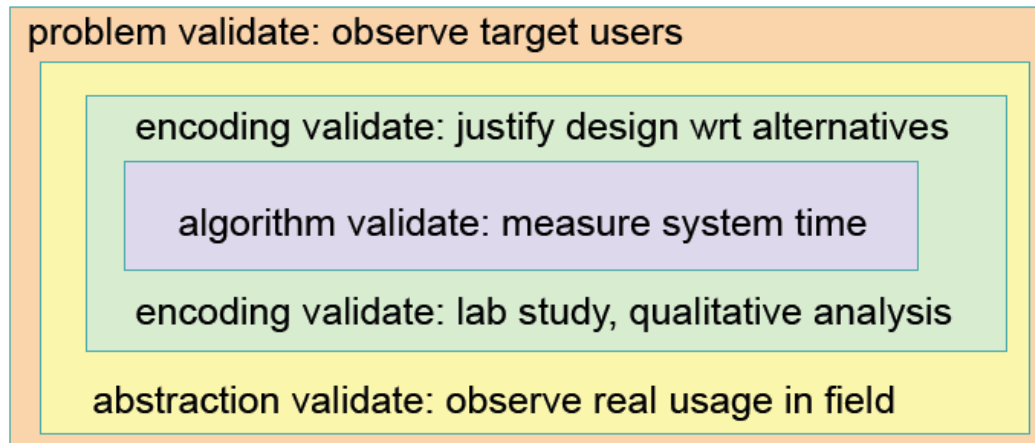
- abstract from domain-specific to generic operations/tasks
 - sorting, filtering, browsing, comparing, finding trend/outlier, characterizing distributions, finding correlation
- data types
 - tables of numbers, relational networks, spatial data
 - transform into useful configuration: derived data model
 - more next time
- validation
 - deploy in the field and observe usage



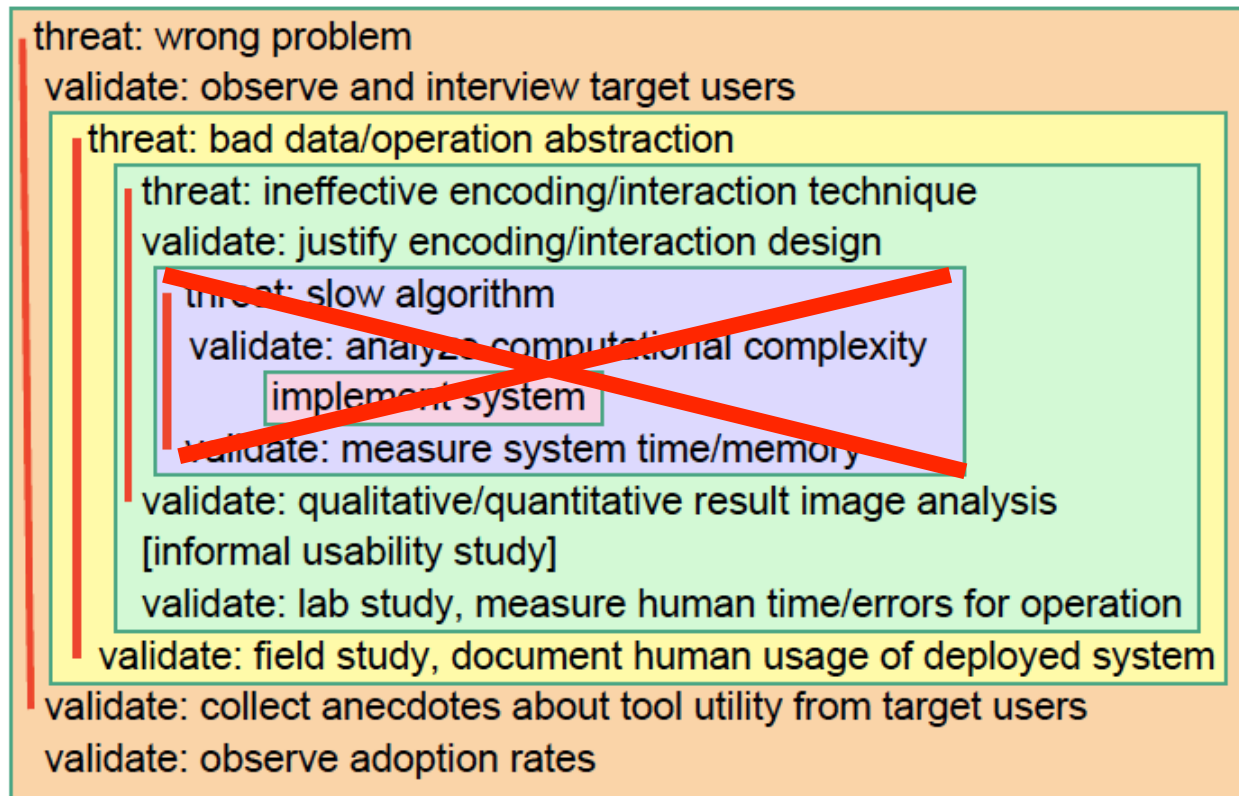
-
- visual encoding: drawings they are shown
 - interaction: how they manipulate drawings
 - validation
 - immediate: careful justification wrt known principles
 - downstream: qualitative or quantitative analysis of results
 - downstream: lab study measuring time/error on given task

Validation

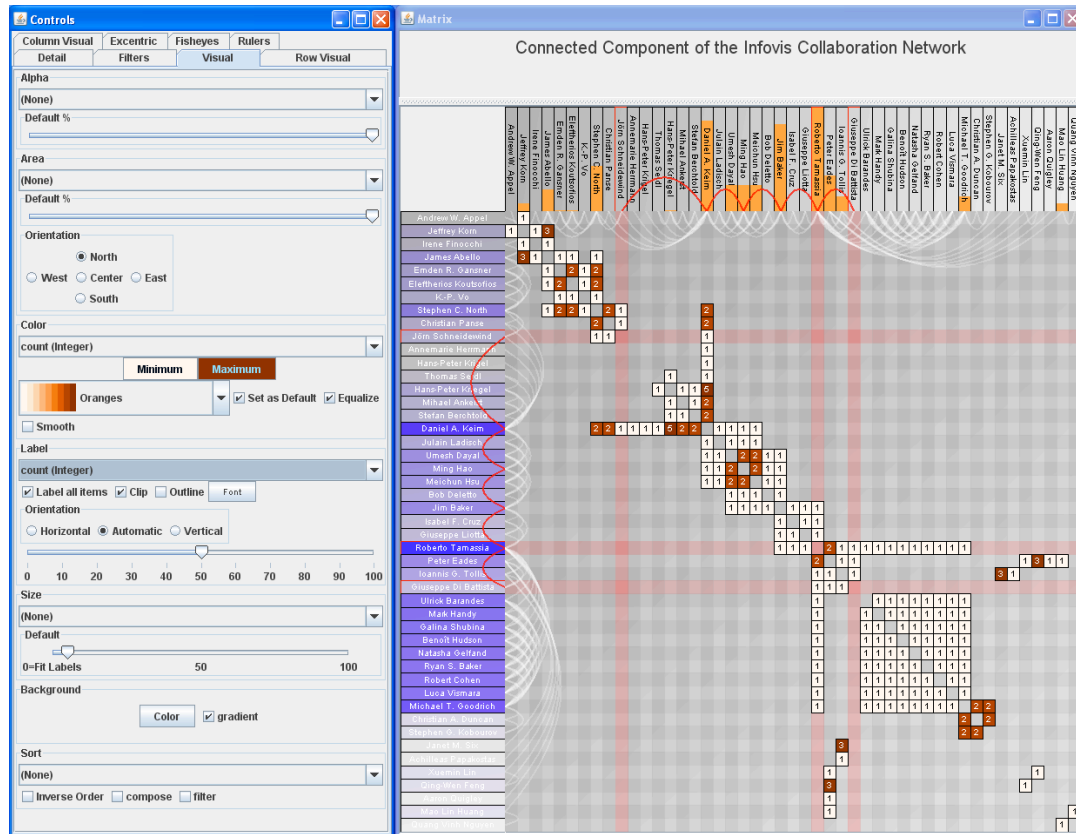
- Can't necessarily answer "Is it worth it?"
- Can try to address "where is it failing"?
- Evaluation is a hard problem, more later



Human in the loop stages



Matrix Explorer case study

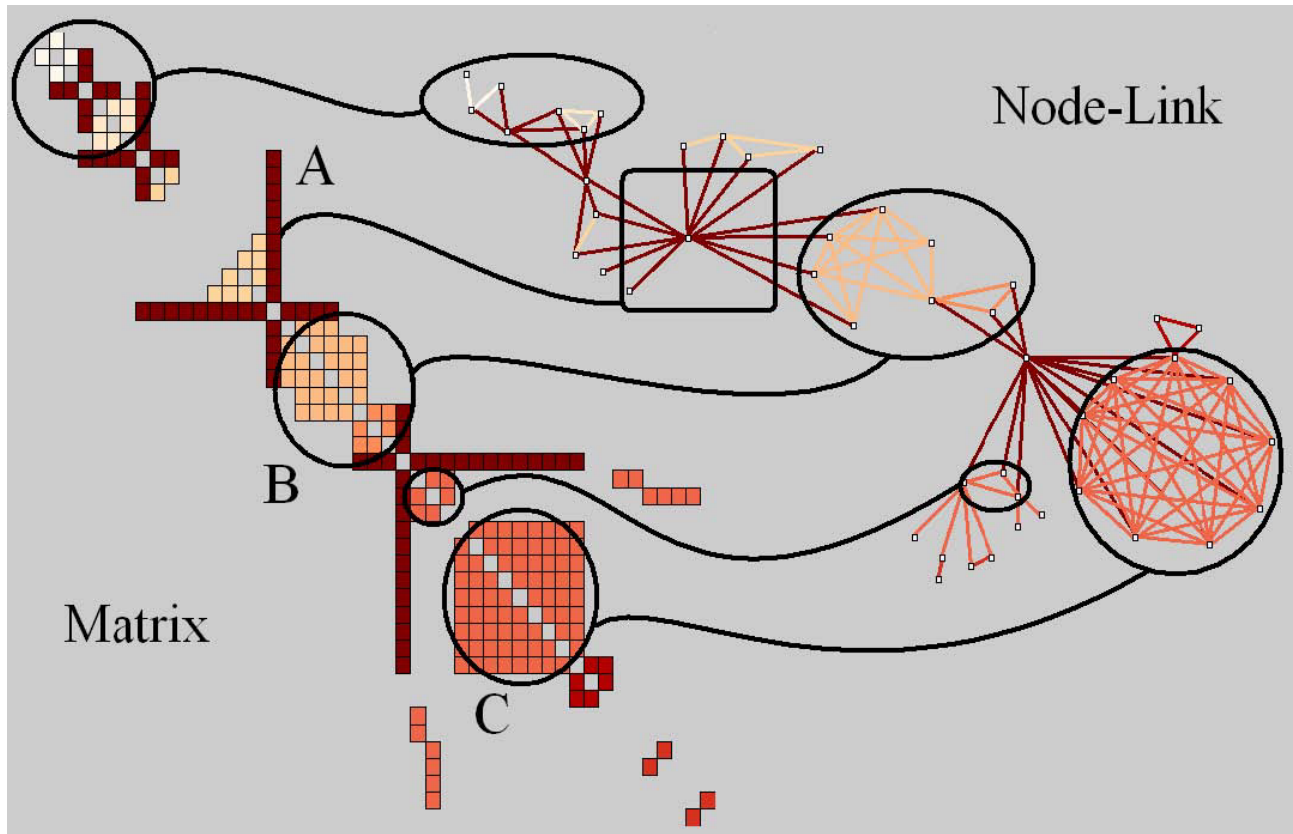


-
- domain: social network analysis
 - early: participatory design to generate requirements
 - later: qualitative observations of tool use by target users
 - techniques
 - interactively map attributes to visual variables
 - user can change visual encoding on the y axis
 - filtering
 - selection
 - sorting by attribute

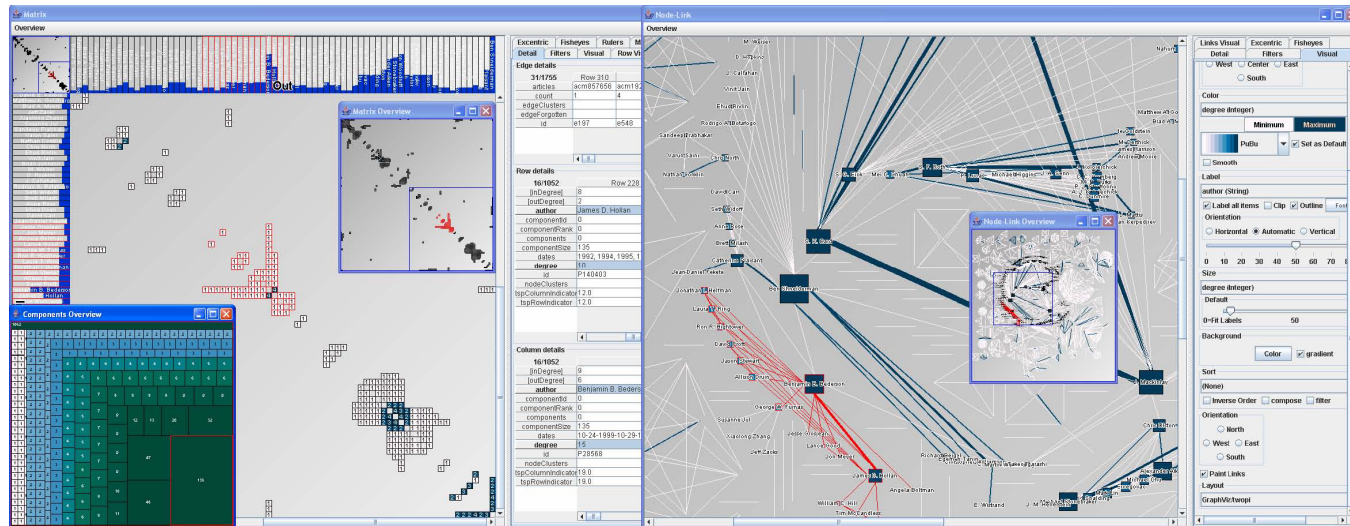
requirements

- use multiple representations
- handle multiple connected components
- provide overviews
- display general dataset info
- use attributes to create multiple views
- display basic and derived attributes
- minimize parameter tuning
- allow manual finetuning of automatic layout
- provide visible reminders of filtered-out data
- support multiple clusterings, including manual
- support outlier discovery
- find where consensus between different clusterings
- aggregate, but provide full detail on demand

Techniques: dual views



- overviews: matrix, node-link, connected components
- details: matrix, node-link
- controls



[Fig 1. Henry and Fekete. MatrixExplorer: a Dual-Representation System to Explore Social Networks. IEEE TVCG 12(5):677-684 (Proc InfoVis 2006)]

Data Sketching exercise

- Data set 1 : Titanic casualties
- Cabin class, Age, Gender, Survived, Survival

- What kinds of insights might you seek from these data?
- How would you represent it?

Data sketching exercise

Life Expectancy data

- #physicians/per capita , country, #Tvs/capita

Add in –

- Change in survival rates over time (20 years)
- Change in physicians

More speculation

- Highway deaths on the Labour Day weekend were much higher than anticipated in BC. You want to find an explanation.
- Decide what kinds of data you need
- Sketch how you would visualize it