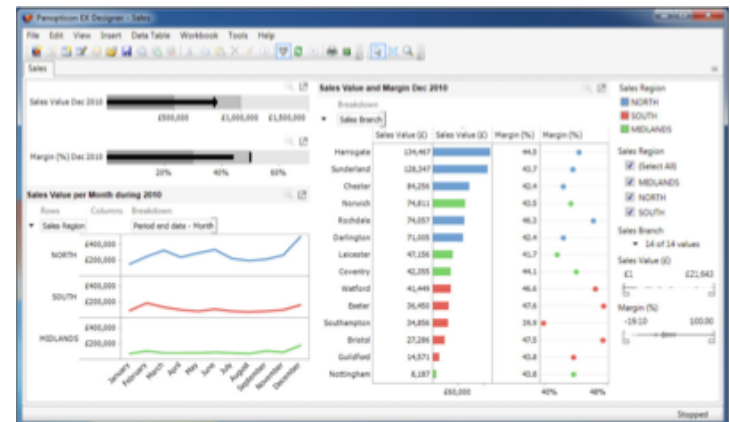
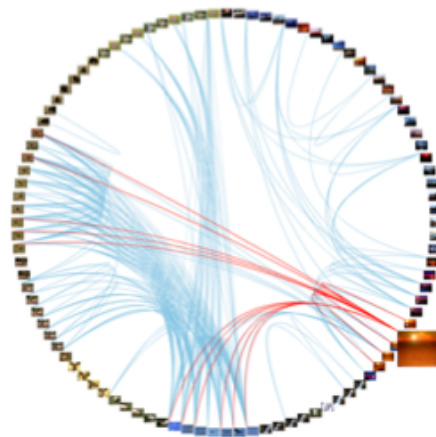


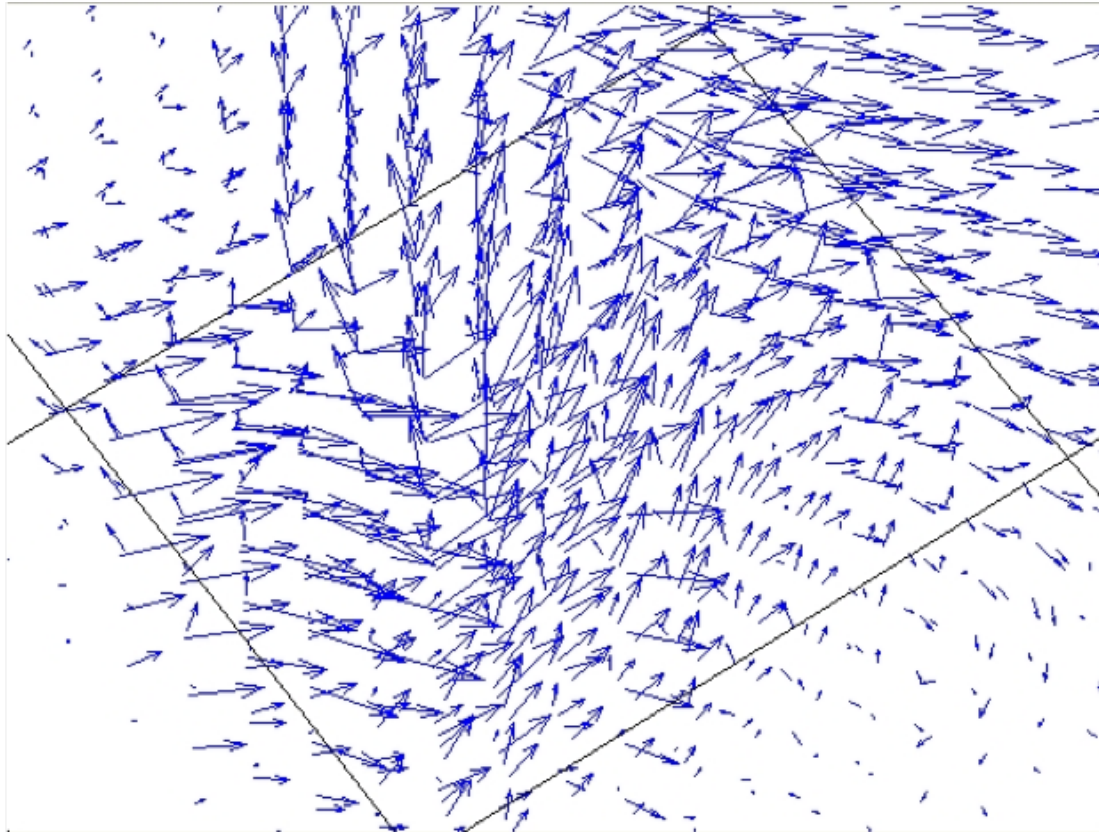
IAT 355

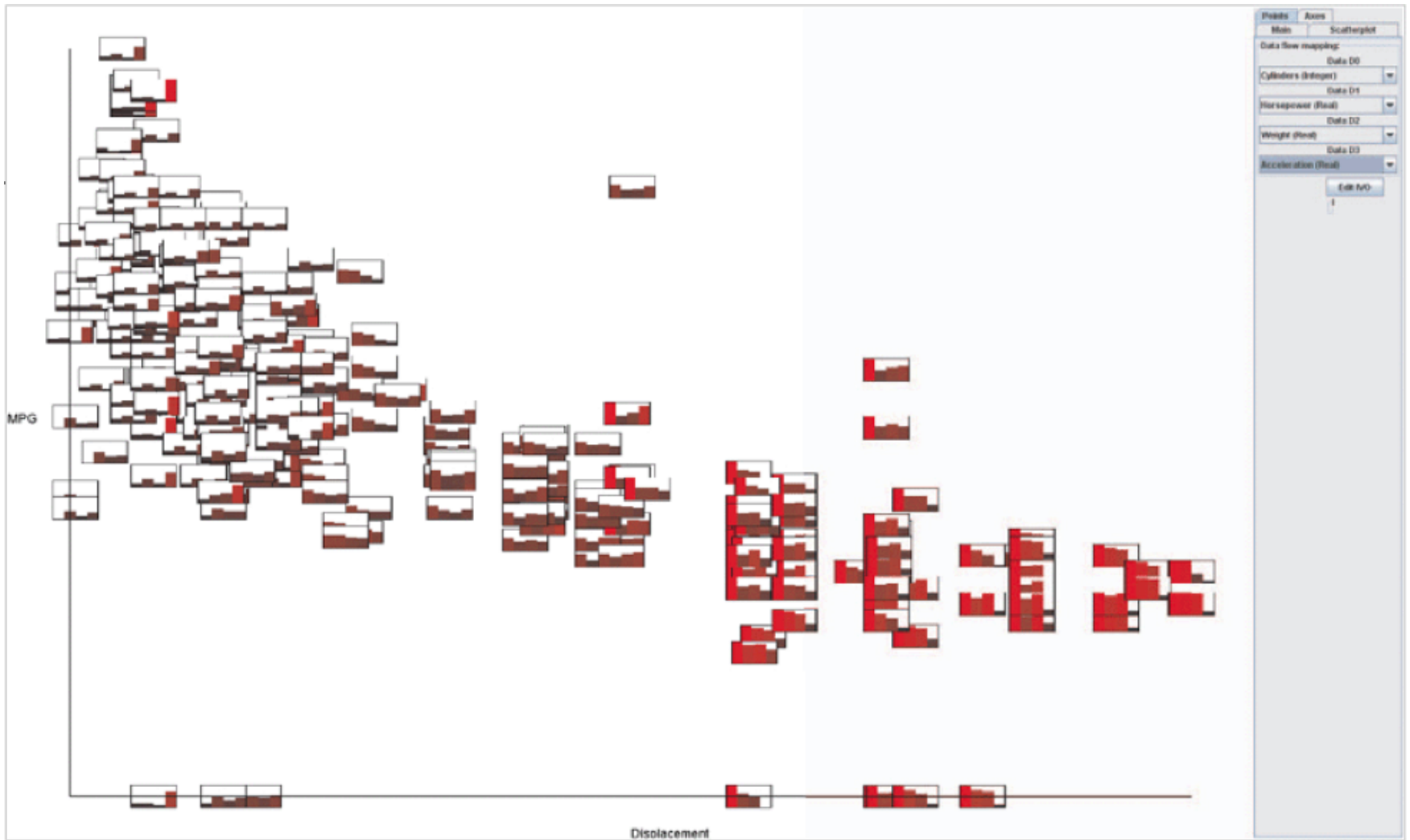
Multiple dimensions, multiple views

Lyn Bartram



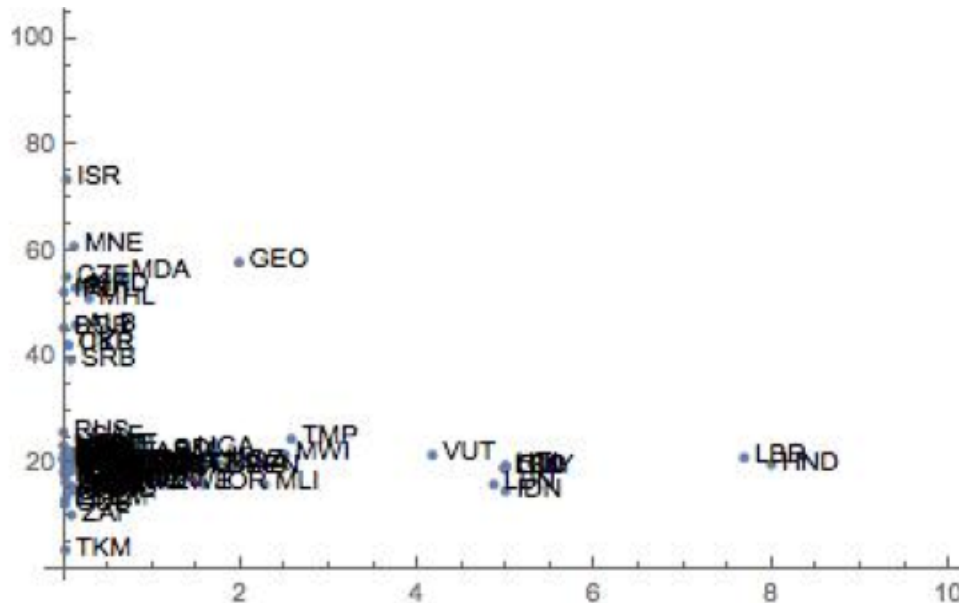
Multidimensional representation: glyphs





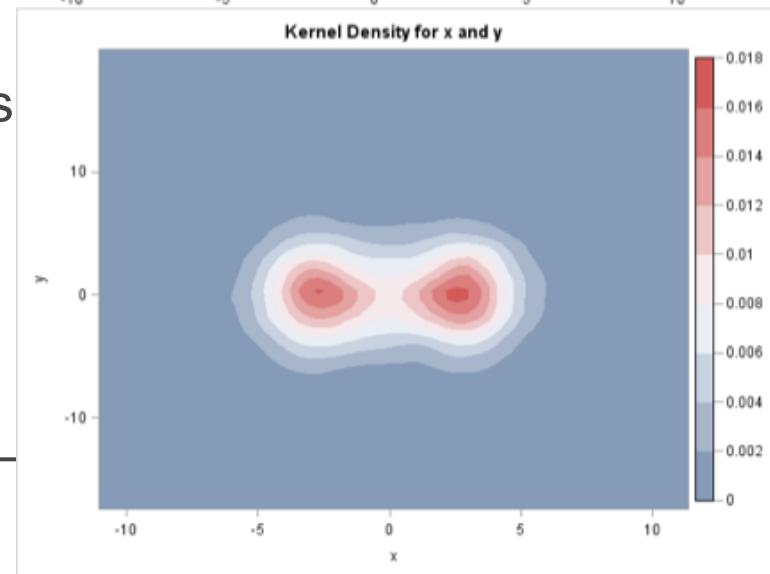
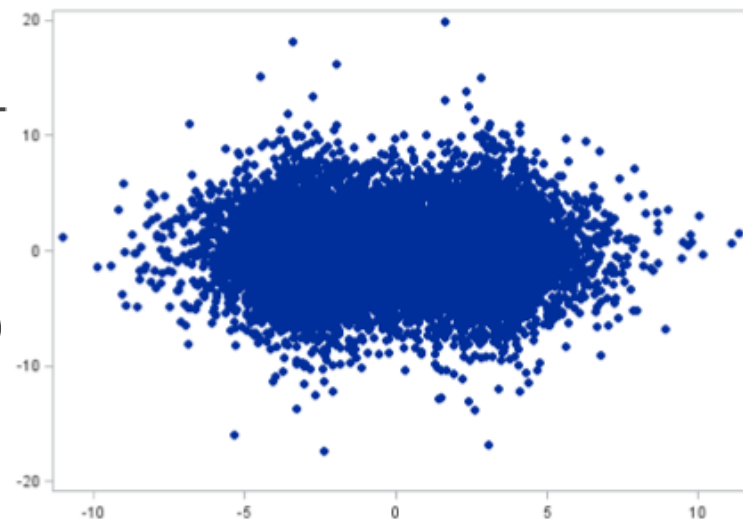
Overplotting

- At some point there are just too many points



Reducing overplotting

- Reduce size of objects
 - Remove fill colour
 - Change shape from container (eg circle) to non-container (X)
 - *Jitter* the data
 - Make data objects transparent
 - Transform the data into density measures (KDE, heatmaps)
-
- White space challenges vis
 - Graphical tricks only go so far.



Approaches to reducing complexity

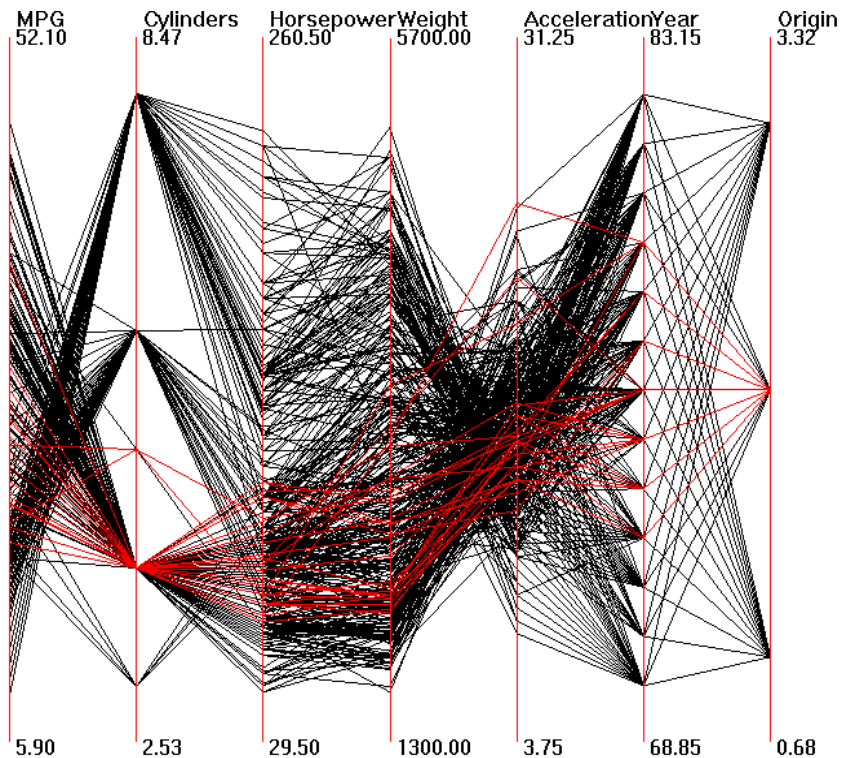
1. Deriving new data (statistics, choosing a design idiom)
2. Navigating through the information space
 - (change over time)
3. Reorganizing the dimension layout
4. Faceting the data/ dimensions into coordinated views
 - Change by using space
5. Reducing the data in the view
6. Embedding focus and context in single view

Reduce complexity: redistribute data

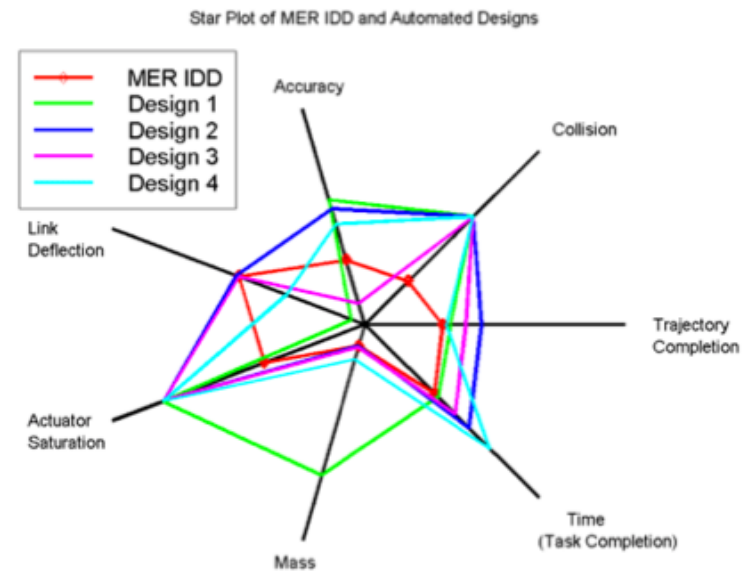
- Dimensional reorganization
- Dimensional embedding
- Dimensional subsetting : *faceting*

Dimensional reorganization

- Parallel coordinates

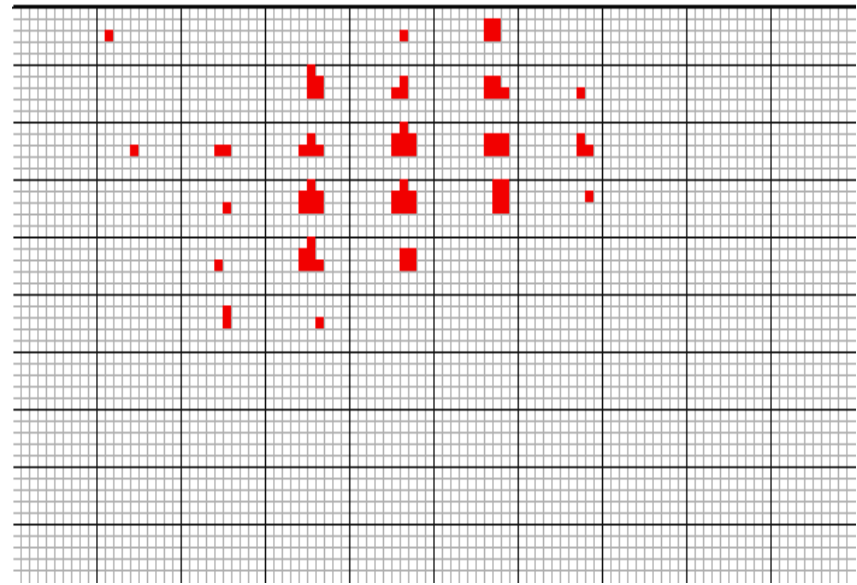


- Radar/star plots

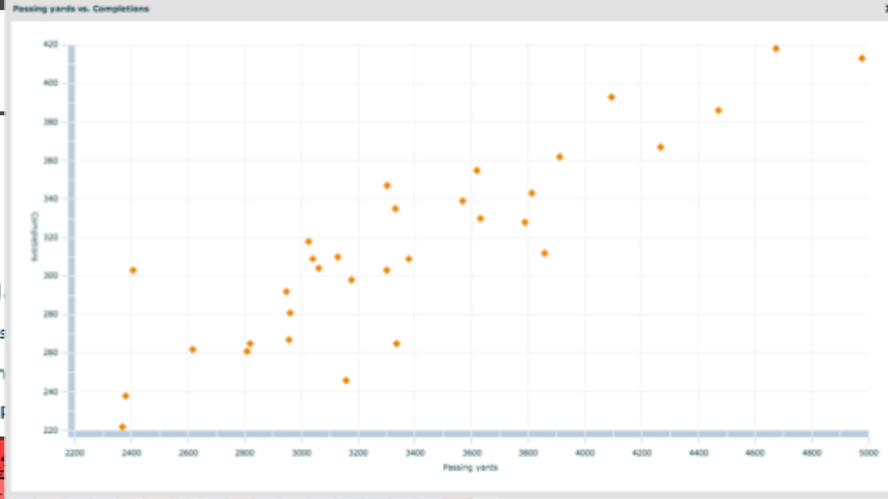
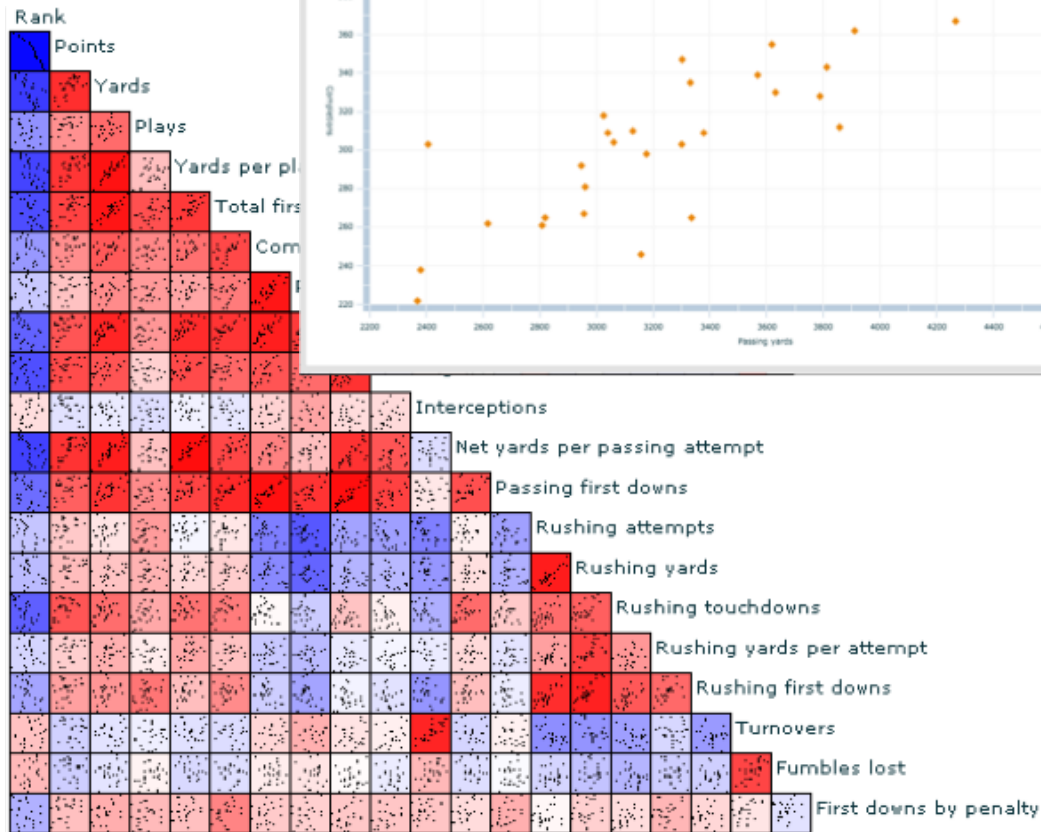


Dimensional Embedding

- Dimensional stacking divides data space into bins
- Each N-D bin has a unique 2-D screen bin
- Screen space recursively divided based on bin count for each dimension
- Clusters and trends manifested as repeated patterns



Dimensional embedding



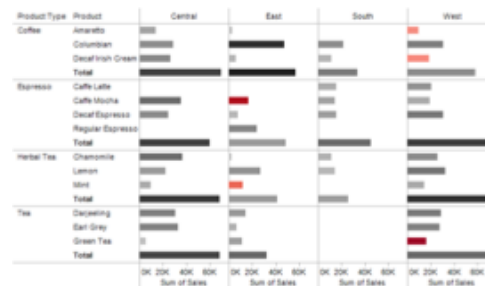
[Visualizing NFL 2008](#)

Bertin's Three Levels of Reading

- **Elementary:** single value

Product Type	Product	Central		East		South		West	
		Sum of Profit	Sum of Sales	Sum of Profit	Sum of Sales	Sum of Profit	Sum of Sales	Sum of Profit	Sum of Sales
Coffee	Amaretto	\$5,100	\$14,611	\$1,000	\$2,893			(\$1,230)	\$9,800
	Columbian	\$8,528	\$28,913	\$27,253	\$47,386	\$8,767	\$21,664	\$11,253	\$36,357
	Decaf Irish Cream	\$9,632	\$26,156	\$2,727	\$6,261	\$2,833	\$11,592	(\$1,305)	\$16,235
	Total	\$23,260	\$69,880	\$30,880	\$56,640	\$11,700	\$33,256	\$8,724	\$57,856
Espresso	Caffa Latte					\$3,872	\$15,442	\$7,002	\$26,458
	Caffa Mocha	\$14,640	\$35,218	(\$6,230)	\$16,646	\$5,201	\$14,563	\$4,064	\$18,876
	Decaf Espresso	\$8,860	\$24,485	\$2,410	\$7,722	\$5,838	\$15,384	\$12,362	\$36,578
	Regular Espresso			\$10,862	\$24,036				
	Total	\$23,500	\$59,703	\$6,242	\$48,405	\$15,063	\$44,989	\$23,868	\$69,911
Herbal Tea	Chamomile	\$14,434	\$36,578	\$765	\$2,194	\$3,180	\$11,586	\$8,852	\$25,632
	Lemon	\$6,201	\$21,978	\$7,061	\$27,176	\$2,583	\$14,497	\$13,120	\$32,274
	Mint	\$4,869	\$9,337	(\$2,242)	\$11,992			\$4,330	\$14,386
	Total	\$24,754	\$67,893	\$6,424	\$41,362	\$5,774	\$26,083	\$26,301	\$72,291
Tea	Decafening	\$10,772	\$30,289	\$6,497	\$14,096			\$11,700	\$26,760
	Earl Grey	\$10,331	\$32,891	\$3,405	\$6,565			\$10,420	\$27,387
	Green Tea	\$1,227	\$1,211	\$5,854	\$11,571			(\$7,490)	\$16,063
	Total	\$22,330	\$64,391	\$15,557	\$32,132			\$15,007	\$70,210

- **Intermediate:** relationships between values



- **Global:** relationships of the whole

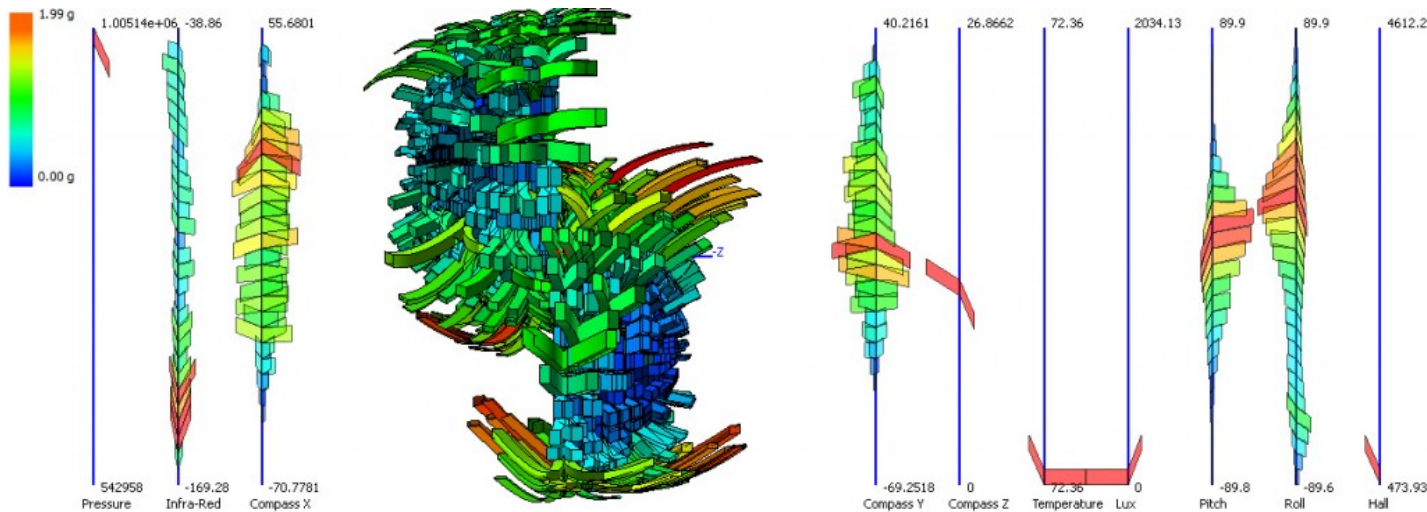
But what's the core problem?

So much data

So many dimensions

Way too much information

- multidimensional view
 - Compound idioms



Solution: Multiple views



Incident Clustering Explorer, University of Maryland

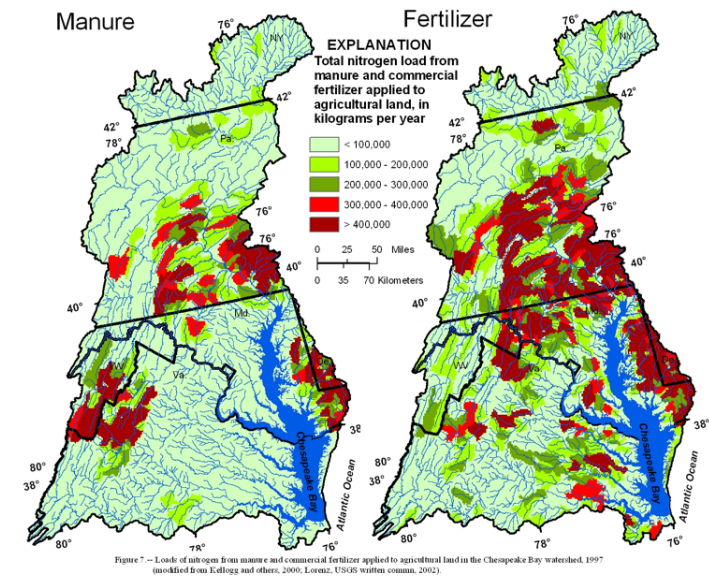


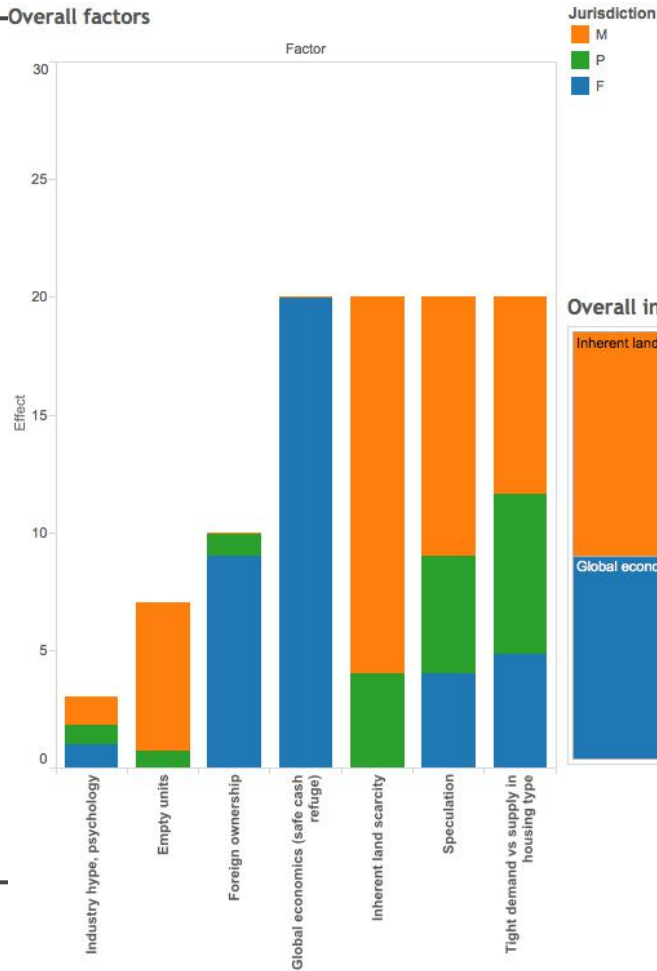
Figure 7 -- Loads of nitrogen from manure and commercial fertilizer applied to agricultural land in the Chesapeake Bay watershed, 1997 (modified from Edrington and others, 2000; Lorenz, USGS written communication, 2002)

Nutrients inputs to Chesapeake Bay Water Quality



Same data, different encoding

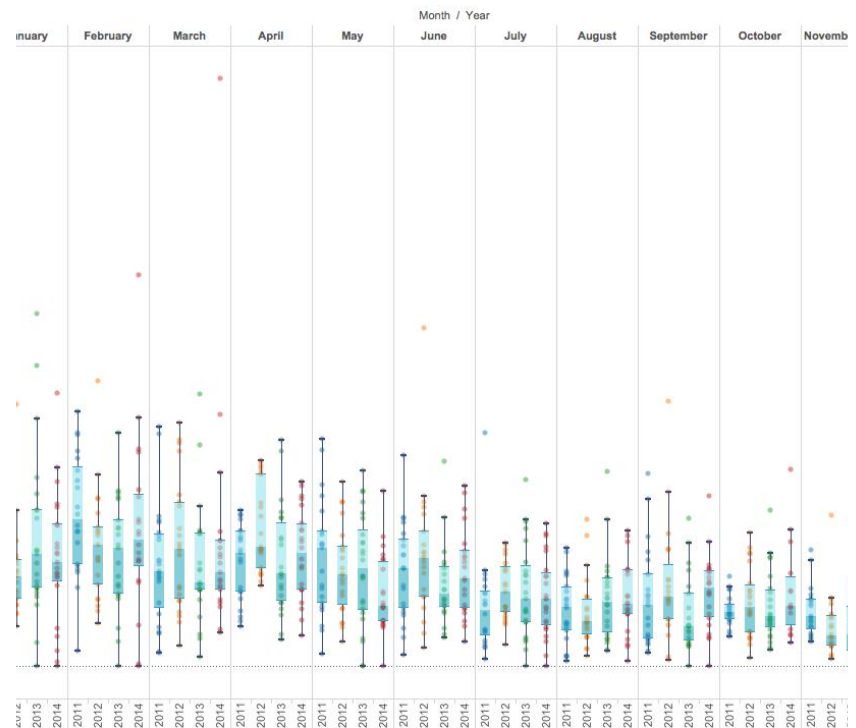
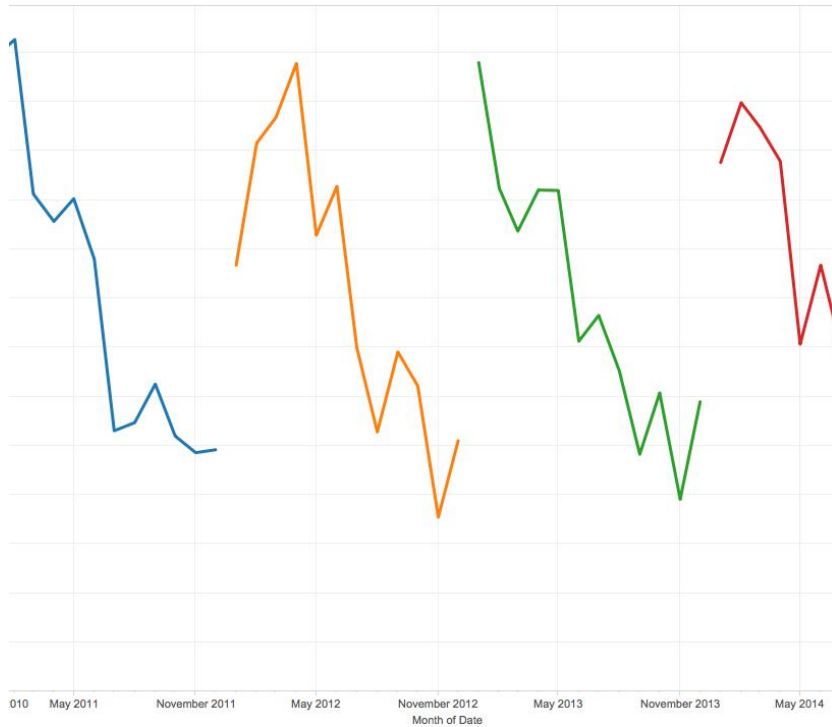
Overall factors



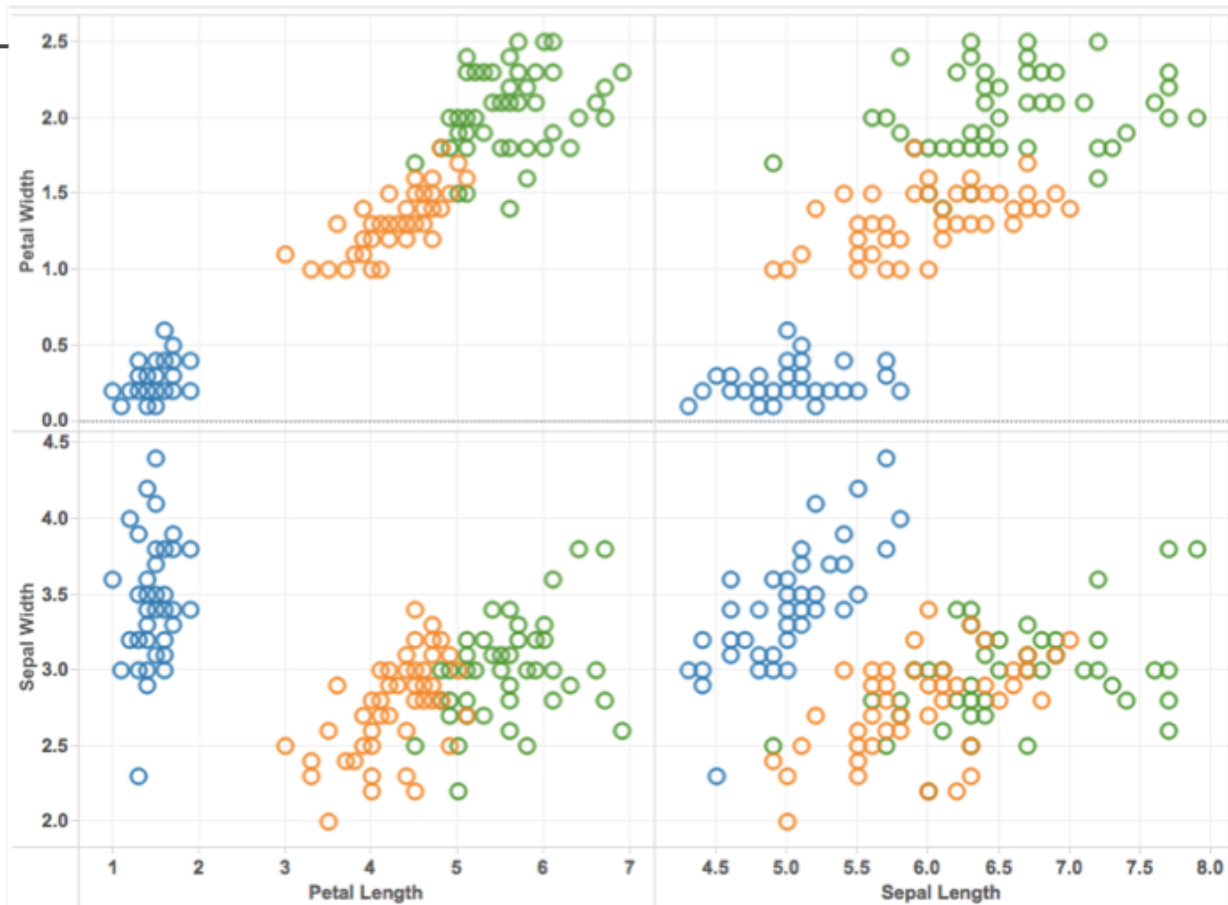
Overall influence



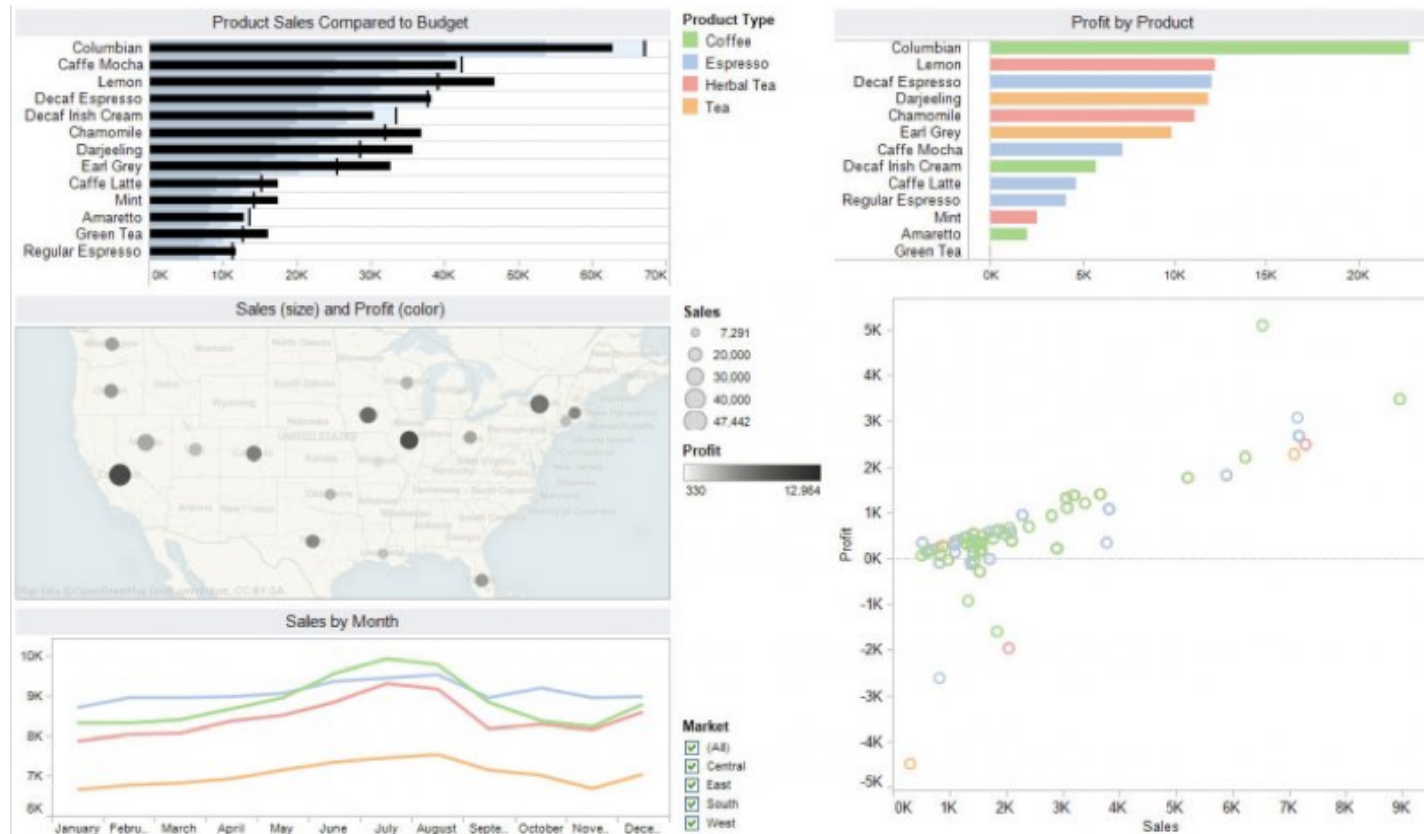
Same data (different levels of detail), different encoding



same encoding, different subsets (by dimension)







Different subsets, different encoding



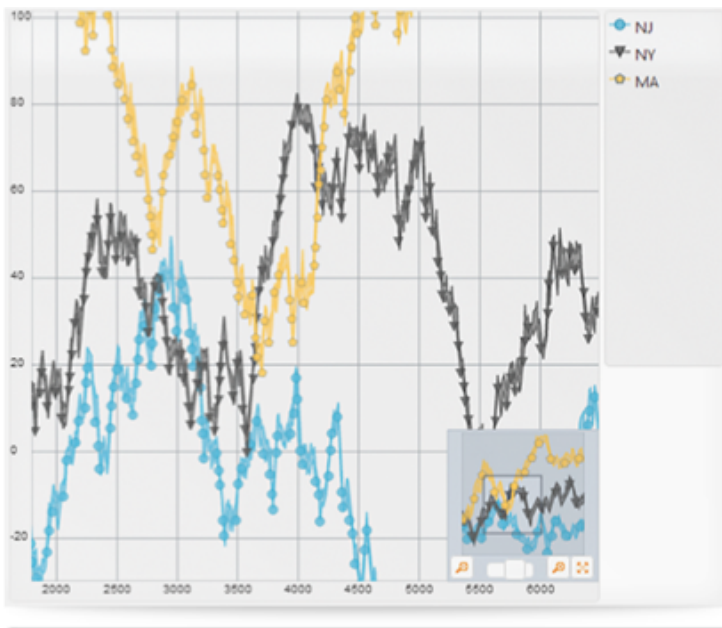
Data and View Facets

➞ Partition into Side-by-Side Views



		Data		
		All	Subset	None
Encoding	Same	Redundant	 Overview/ Detail	 Small Multiples
	Different	 Multiform	 Multiform, Overview/ Detail	No Linkage

Overview and detail: subset data, same encoding



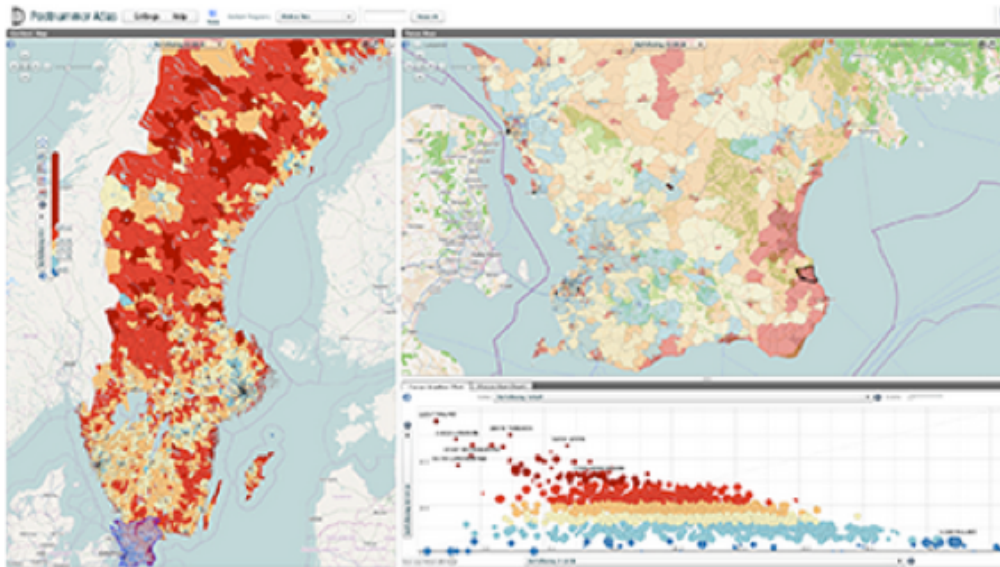
Infragistics,
http://www.infragistics.com/community/blogs/taz_abdeal_i/archive/2012/03/13/netadvantage-for-jquery-2012-volume-1-sneak-peek.aspx



http://www.wikiviz.org/wiki/Overview_plus_detail

Faceting data: multiform overview and detail

Postnummer- och SAMS-atlasen

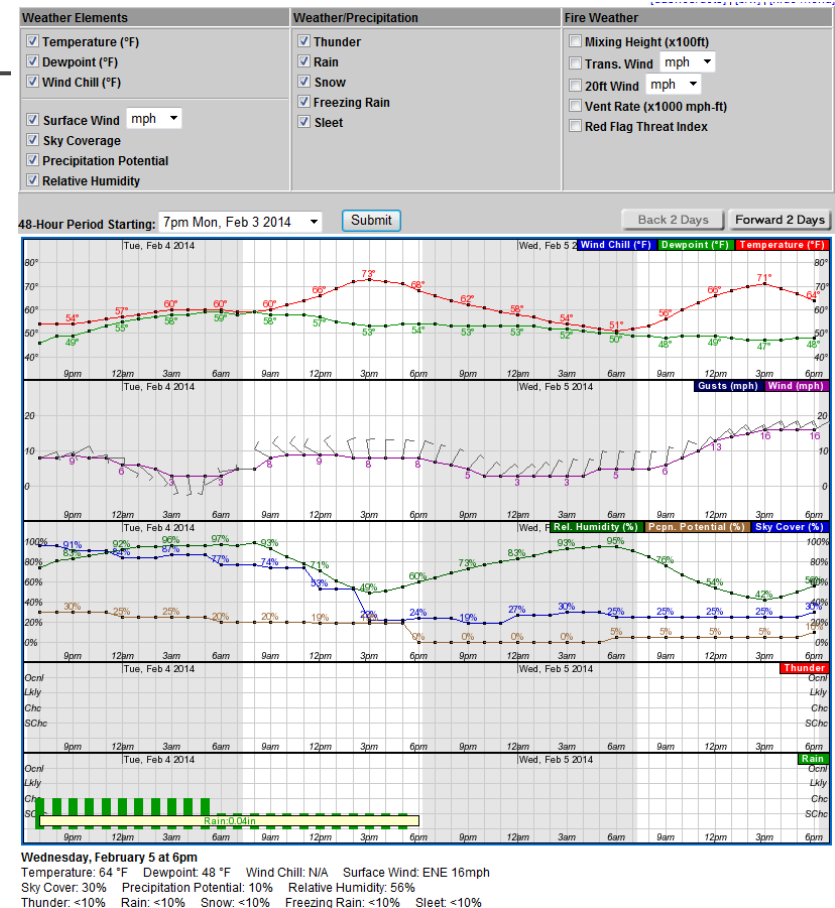


[GAV HTML5 TOOLKIT, Linköping University](#)



Subsetting dimensions

- “splitting” dimensions across linked views
- Small multiples
- Trellis displays
- Scatterplot matrices

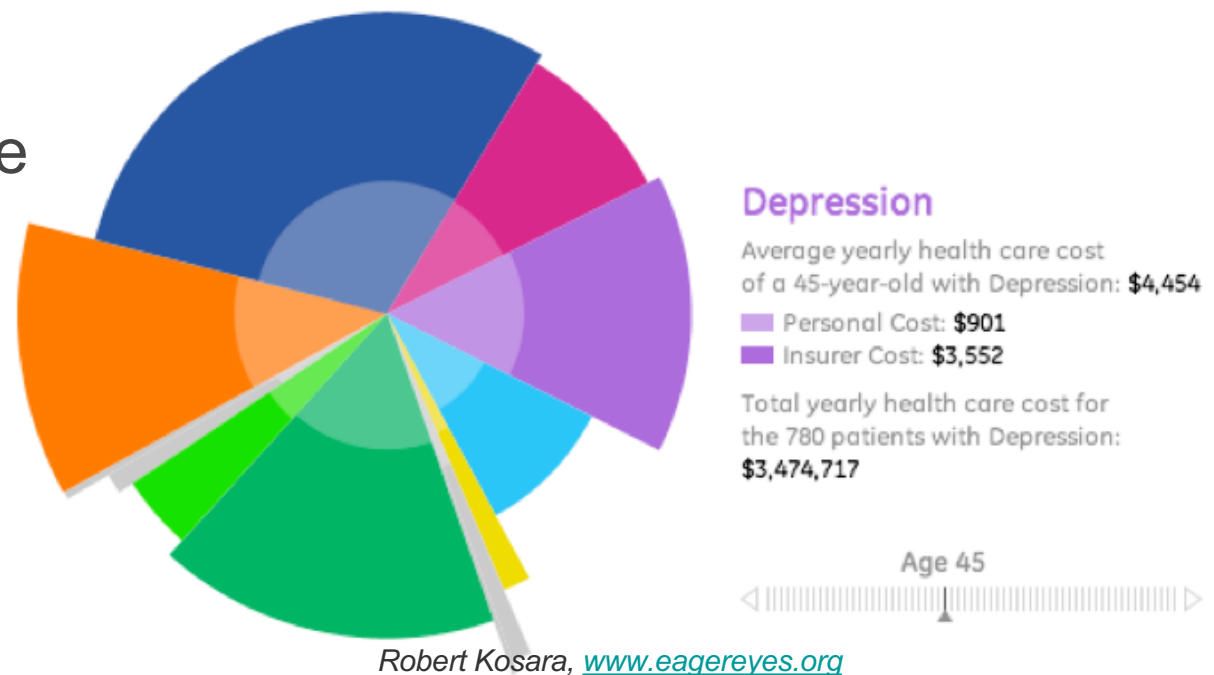


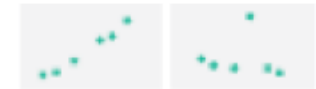


Small Multiples

Multiform: Small multiples

- Splits dimensions across multiple views

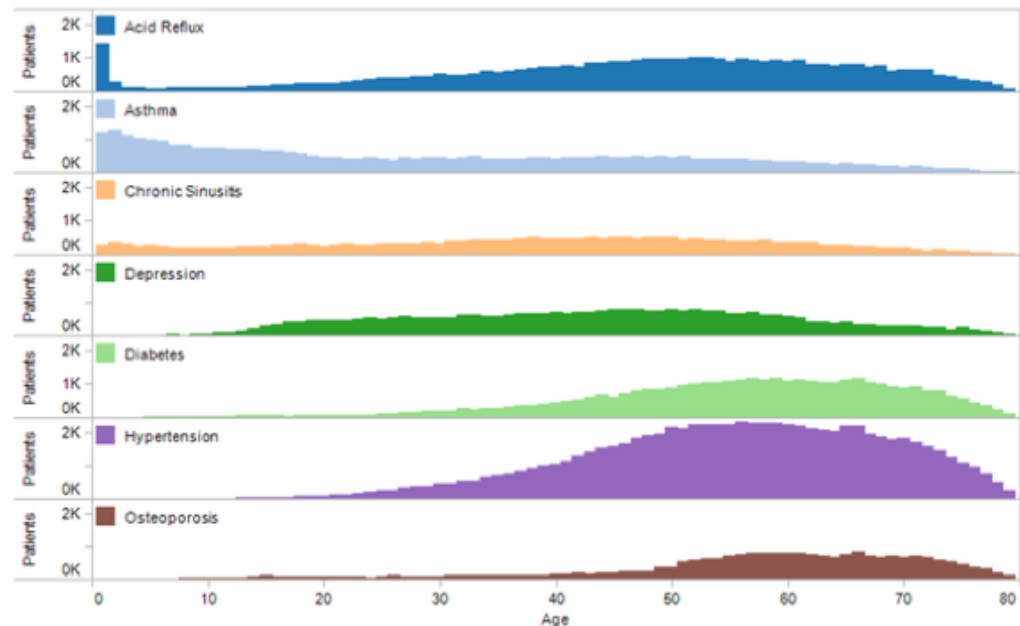




Small Multiples

Multiform: Small multiples

- Splits dimensions across multiple views



Robert Kosara, www.eagereyes.org

Monthly Unemployment Rates by State, Jan 1976 - Apr 2009



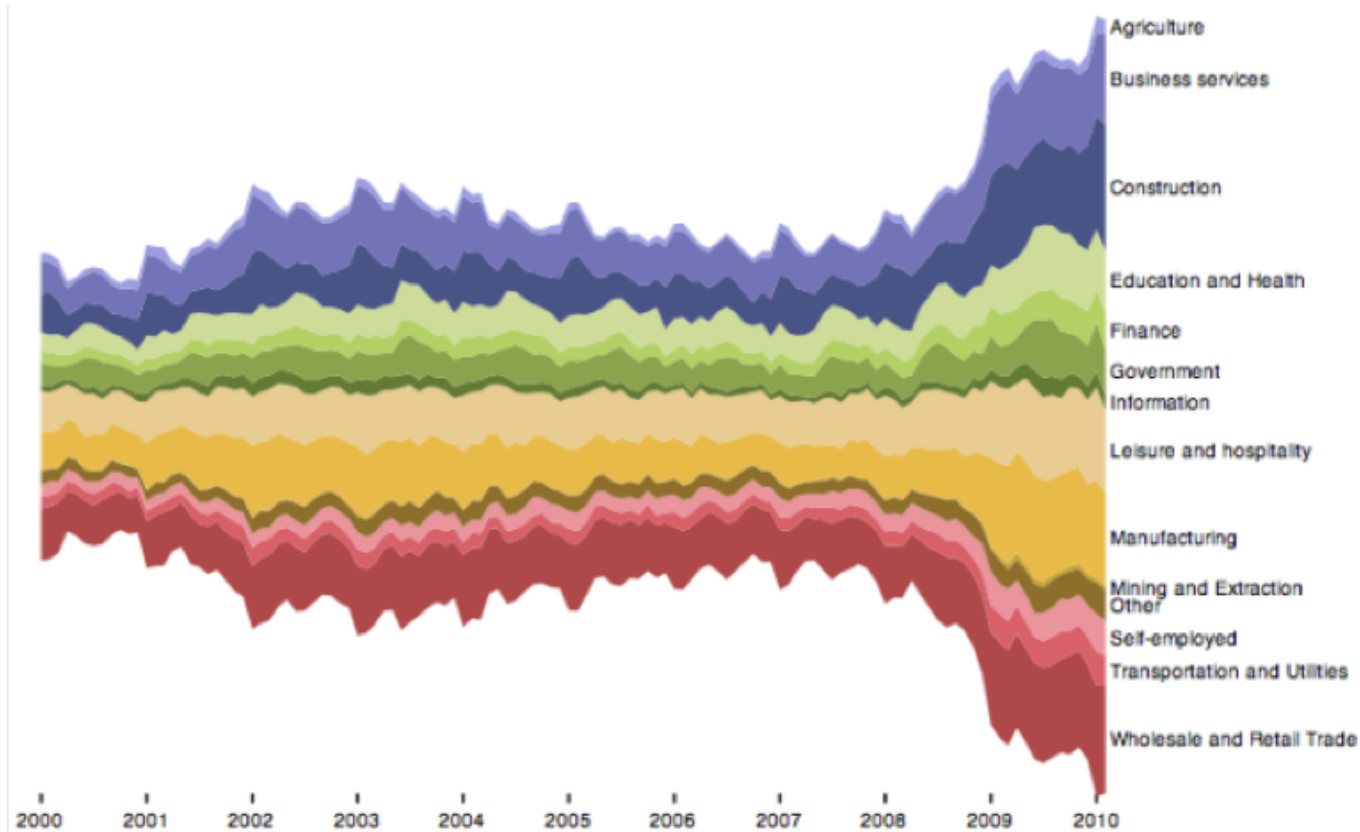
Source: Bureau of Labor Statistics

Notes: The orange band denotes a "normal" unemployment rate (4%-6%).
State code in red: unemployment rate in April 2009 is higher than the US average

Small multiples

- use the same basic graphic or chart to display difference slices of a data set
- rich, multi-dimensional data without trying to cram all that information into a single, overly-complex chart.
- Singular design reduces decoding effort.
 - E. Tufte "The Visual Display of Quantitative Information," p. 42 and "Envisioning Information," p. 29

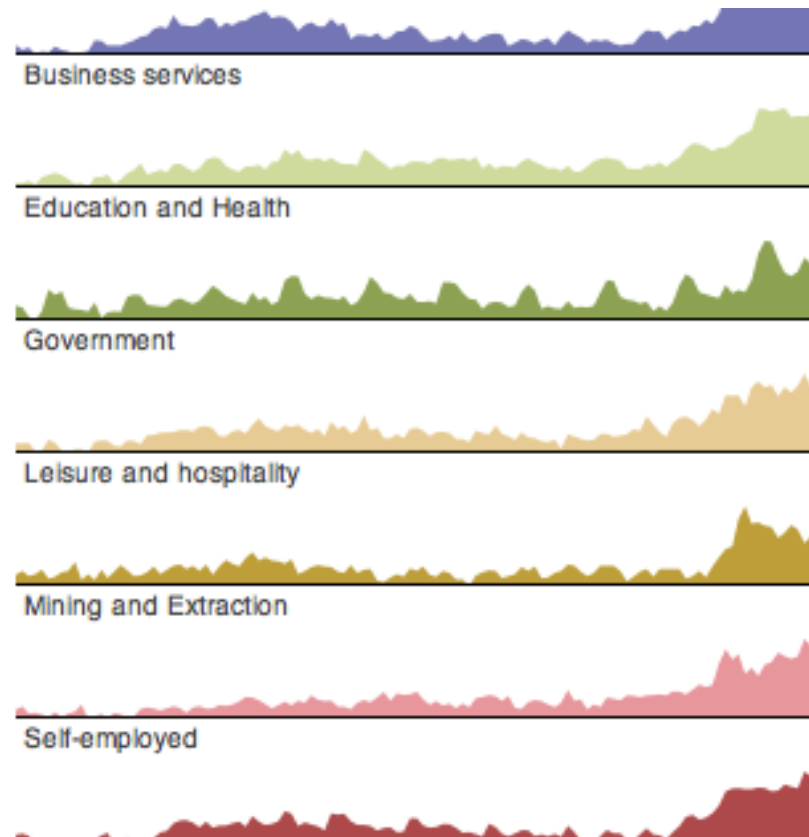
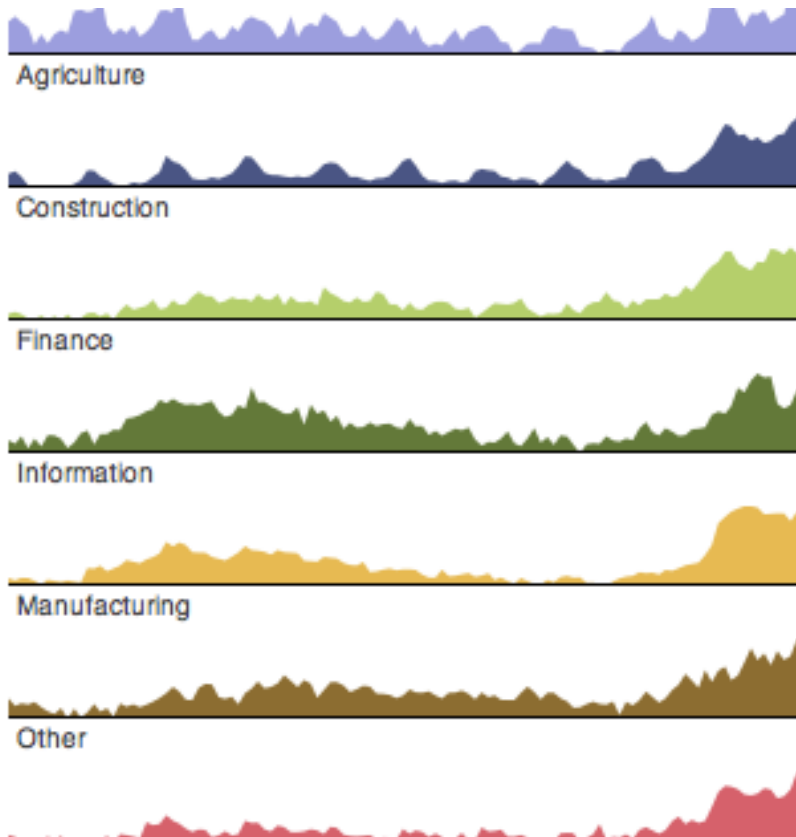
Why doesn't this work?





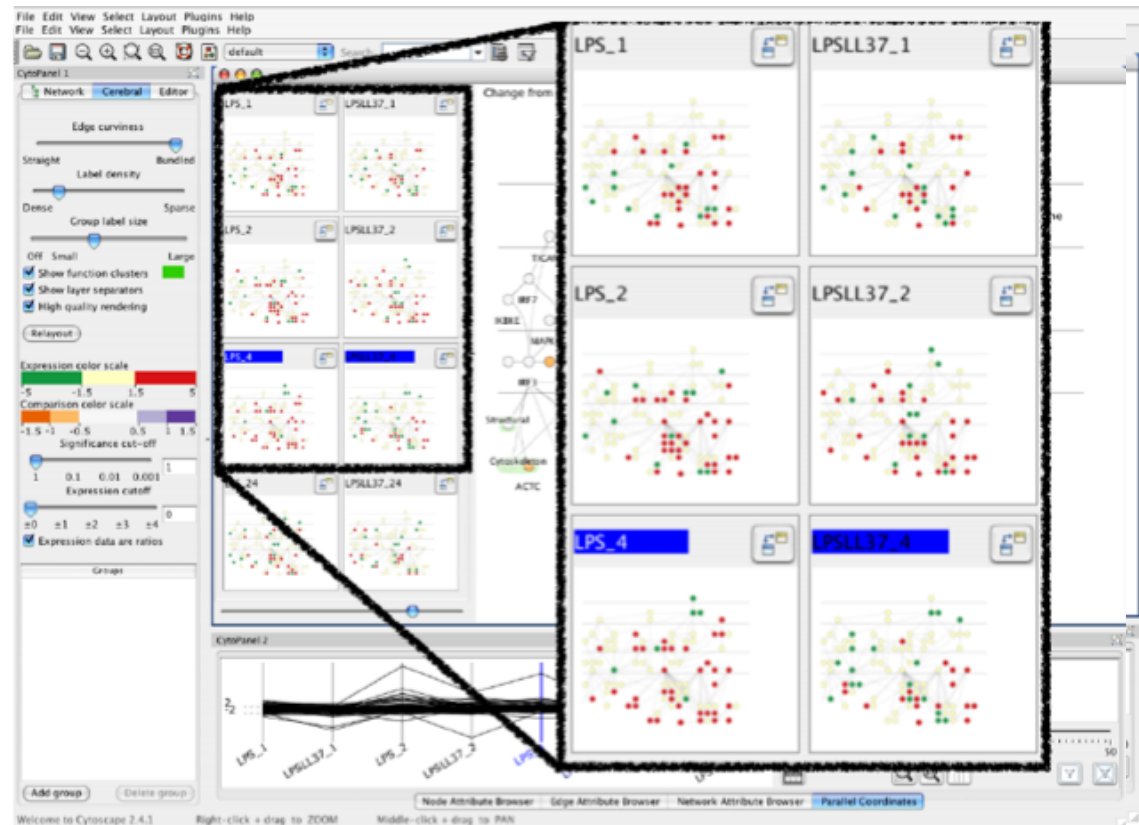
Small Multiples

Small multiples

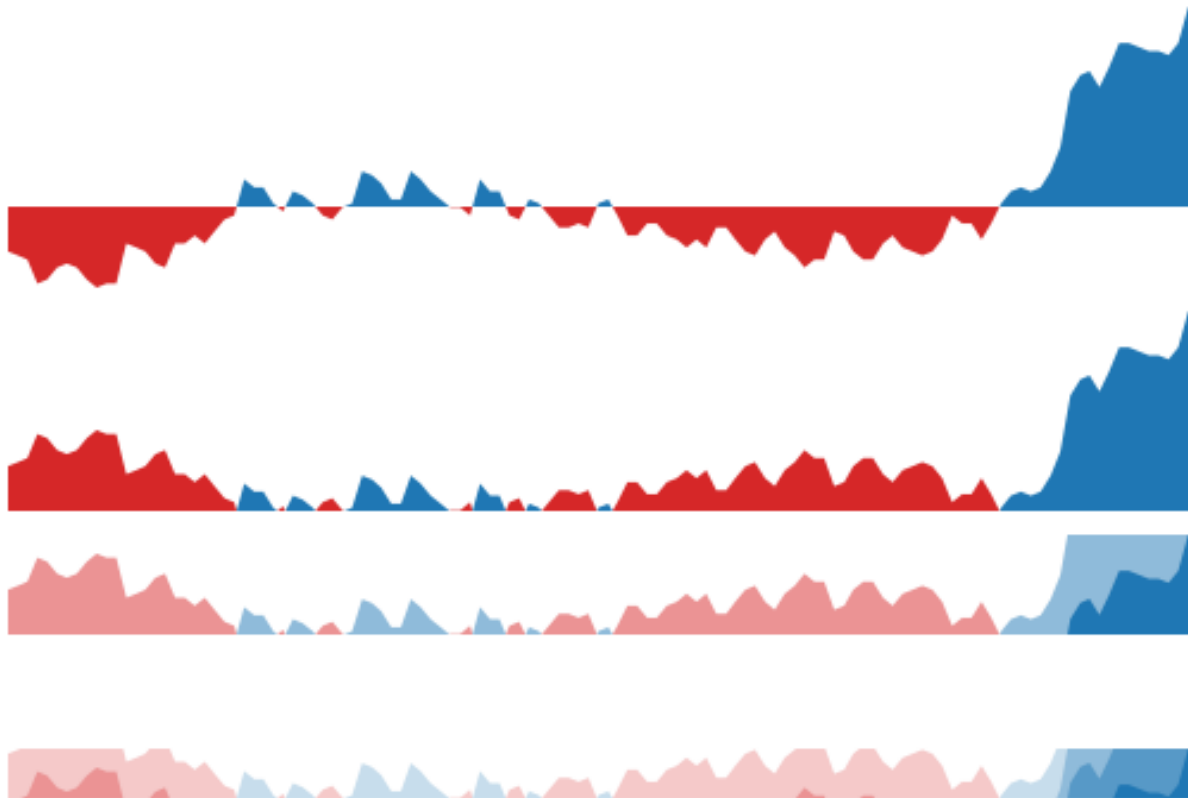


Small multiples

- Same encoding
- Data split

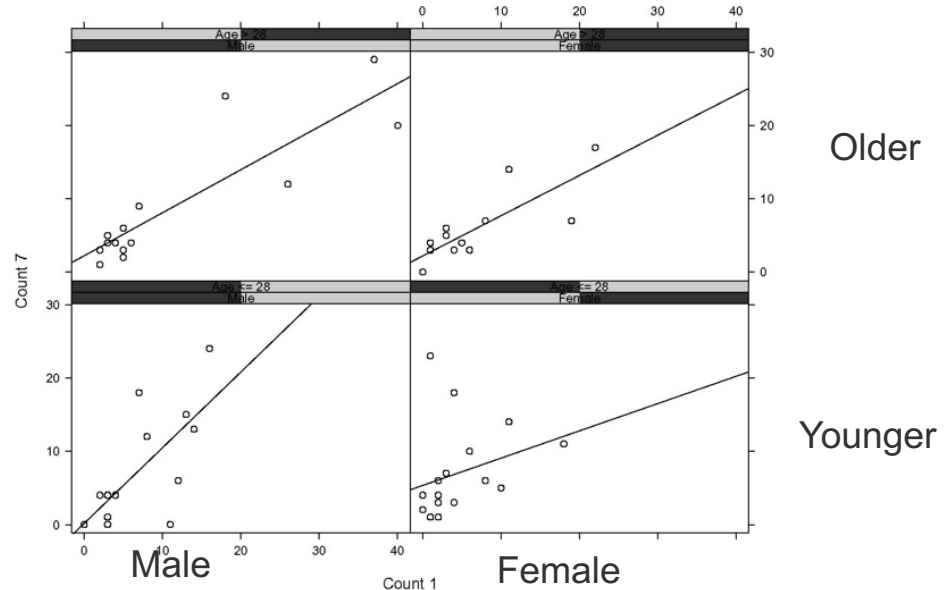


Horizon graphs



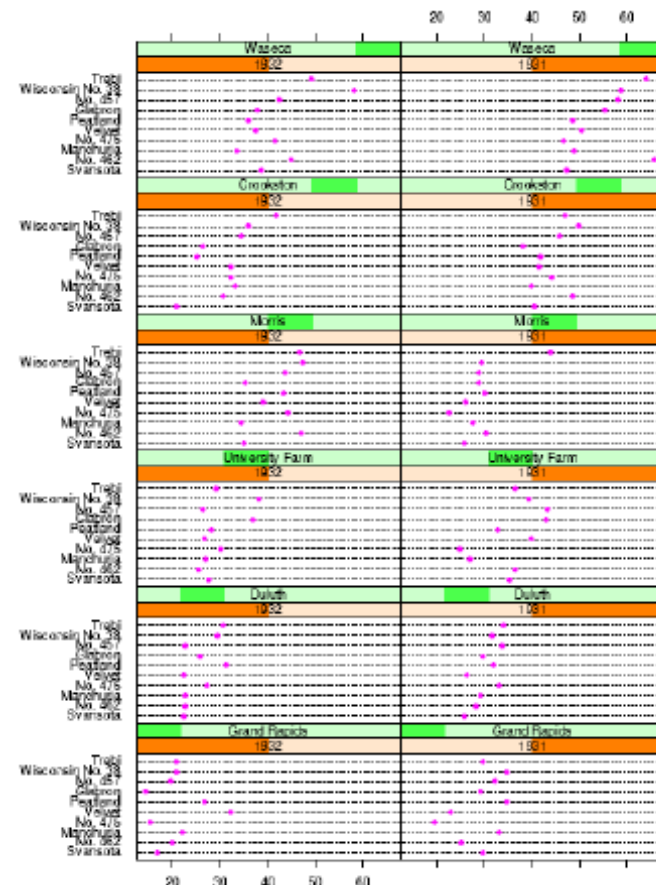
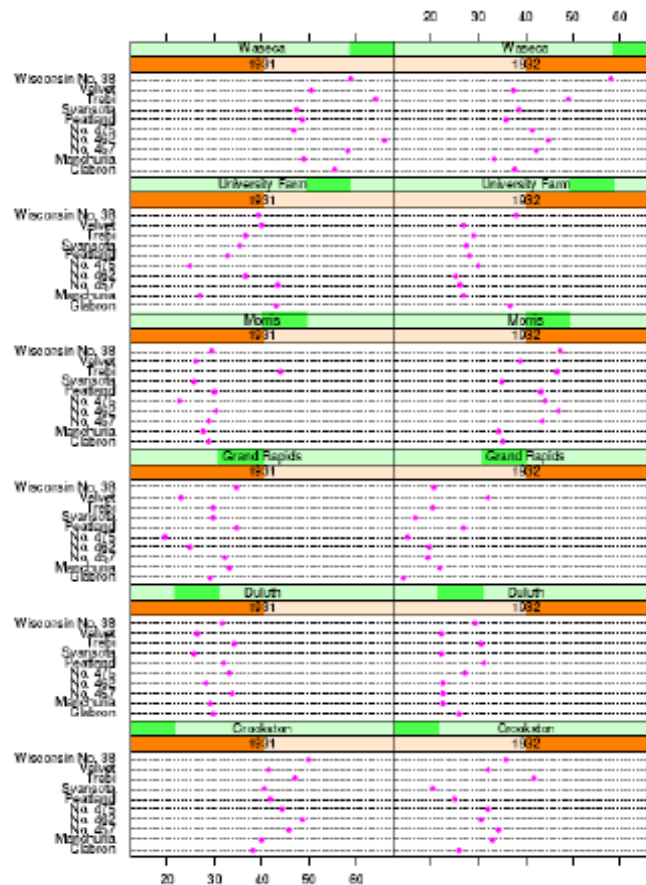
Trellis Plot

- Small multiples subsets across more than one dimensions
- Two dimensions:
- Age
- Gender



Trellis plots

- Sorting helps see different patterns in the data

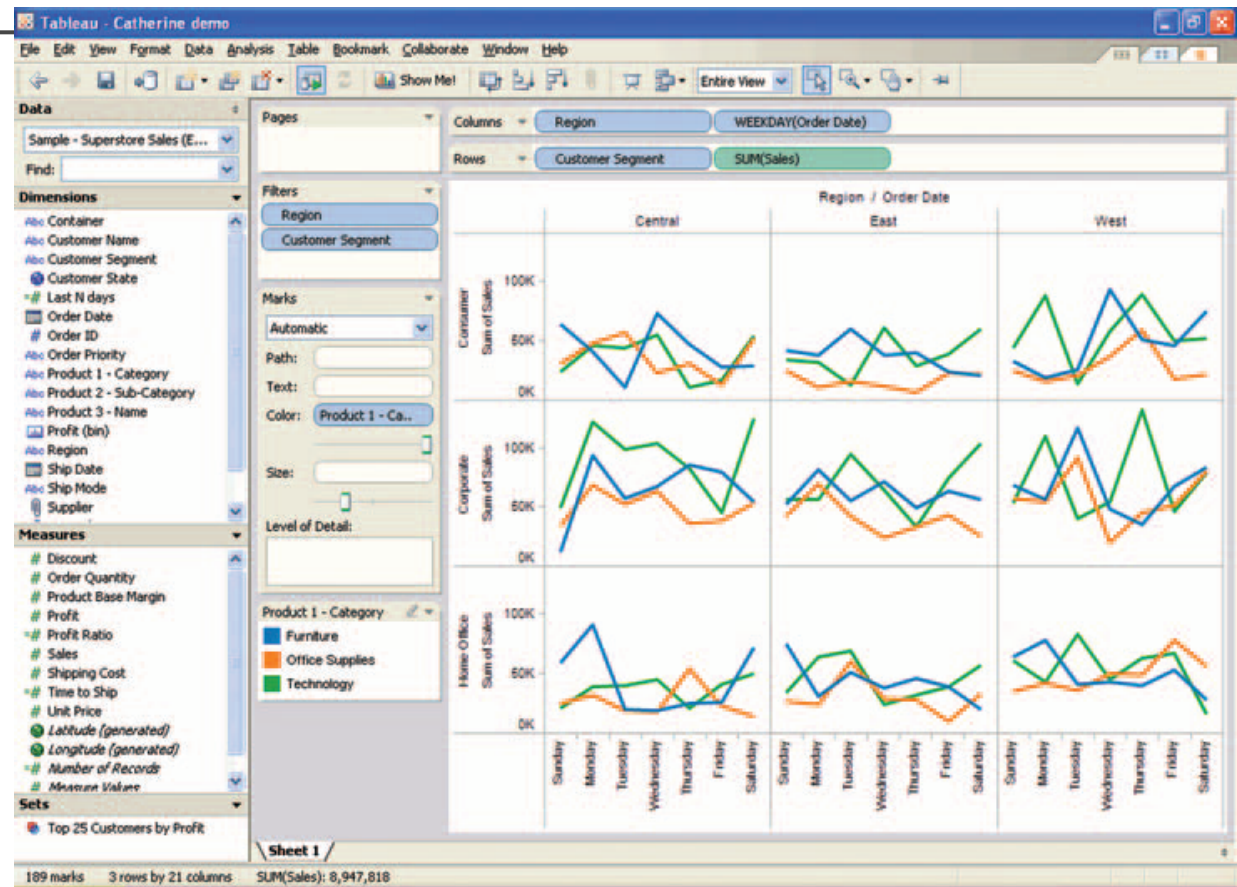


[The Visual Design and Control of Trellis Display. Becker, Cleveland, and Shyu. JCSG 5:123-155 1996]

Small multiples

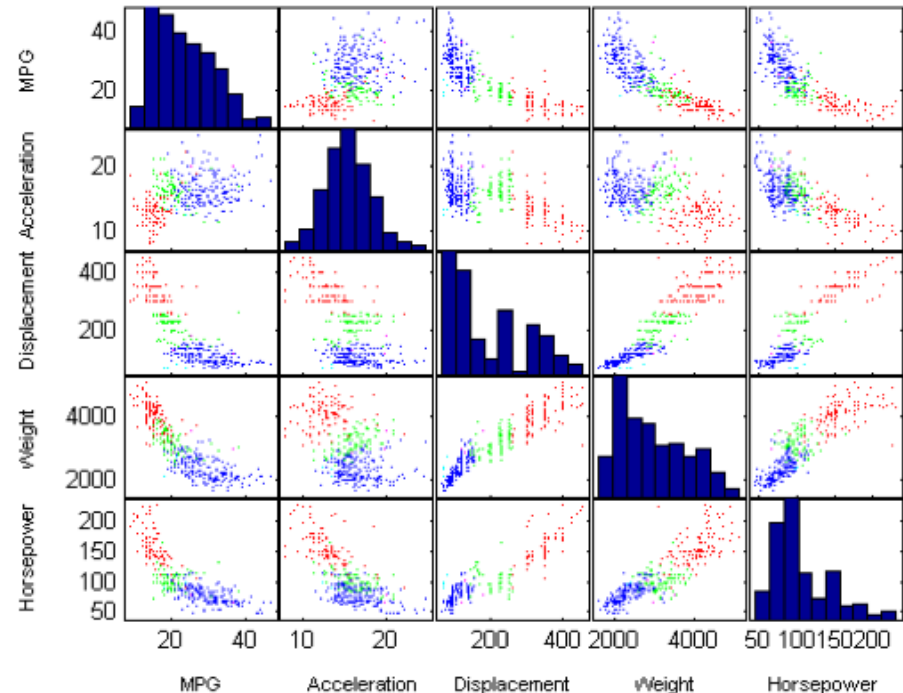
Often called most underused vis solution

- Issues/?



Small multiples matrix

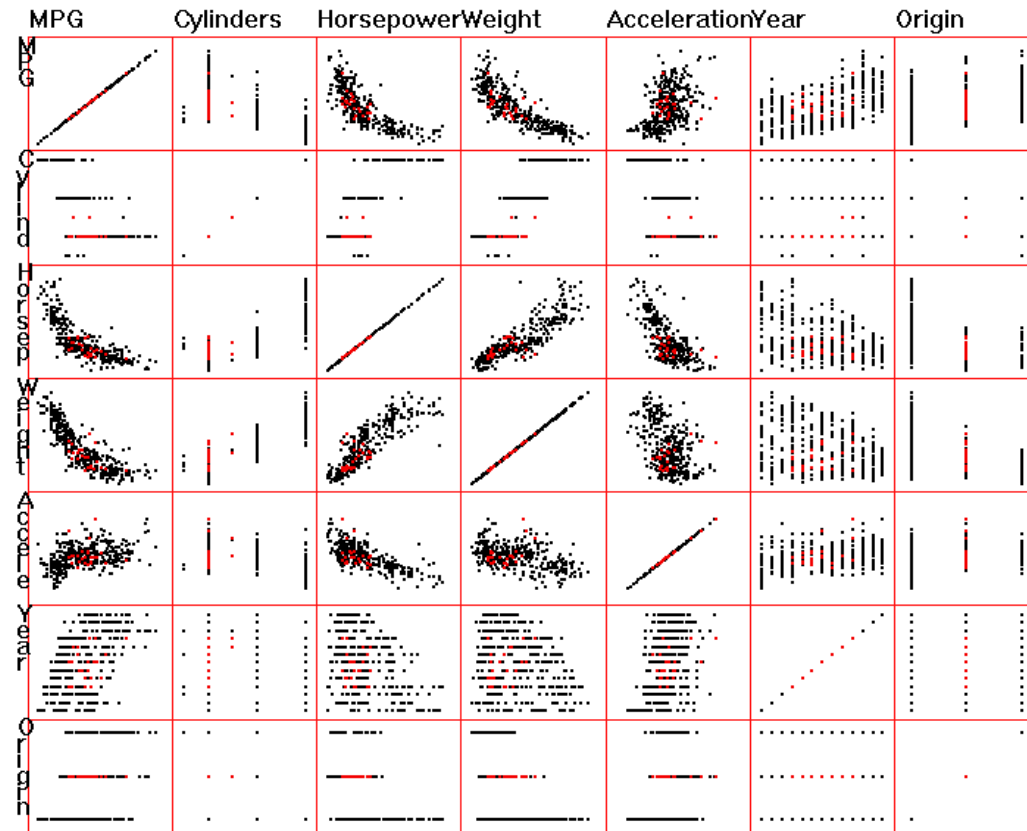
- n-dimensional data set
- subset the data into categories to compare the patterns between subsets
- Lay out small multiples in sequential or 2d ordered grid



Adapted from M. Ward

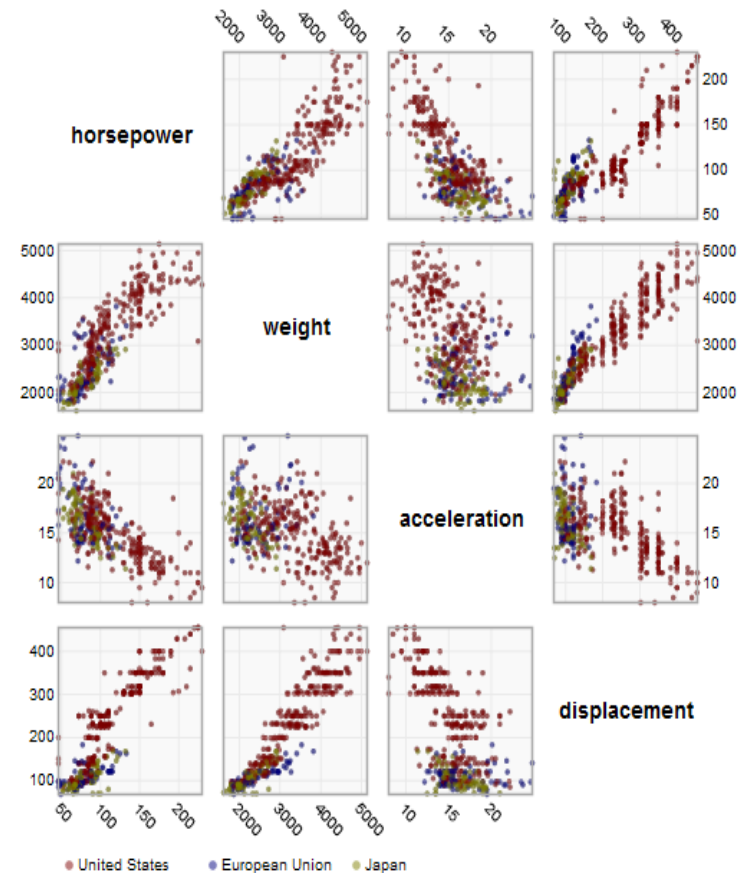
Scatterplot matrix

- all pairwise plots
- Selection allows linkage between views
- Clusters, trends, and correlations readily discerned between pairs of dimensions
- Issues?



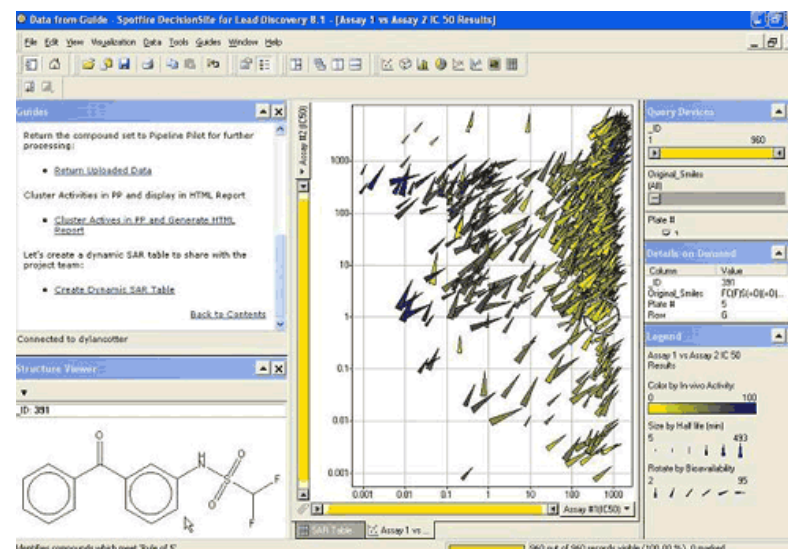
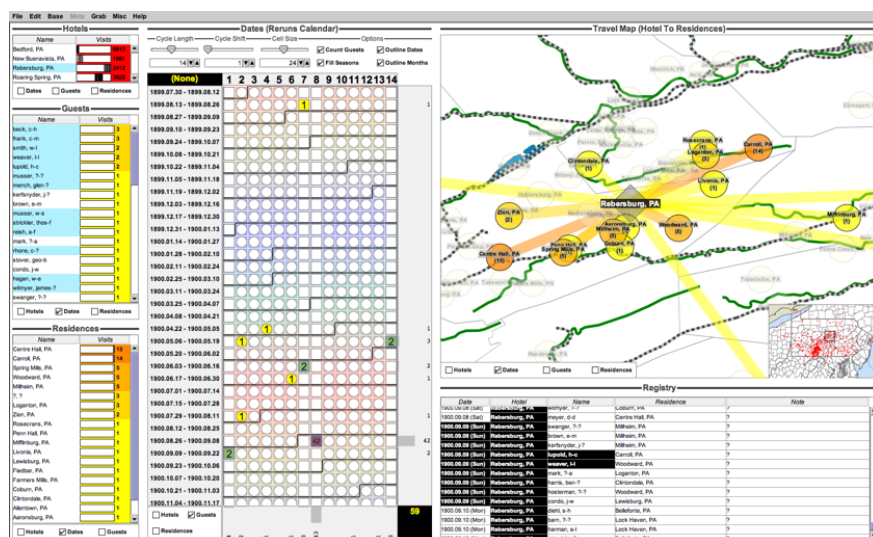
Scatter plot matrices

- More elegant layout sacrifices alignment but improves indexing
- Layout is important
- Density



Multiform views

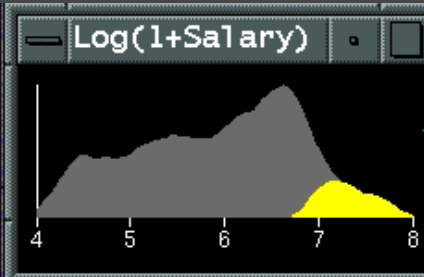
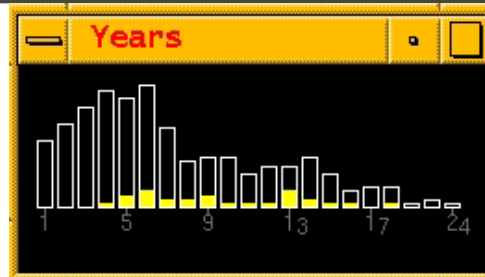
- Improvise [Weaver]
- Spotfire [Ahlberg]



Snap-together visualizations

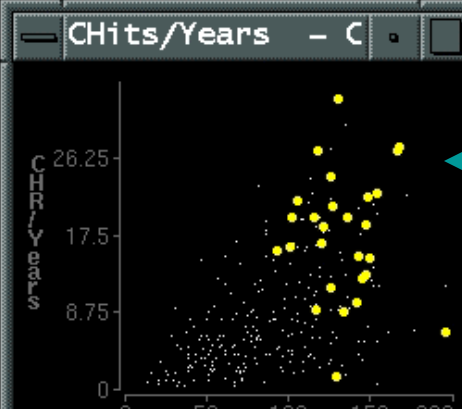
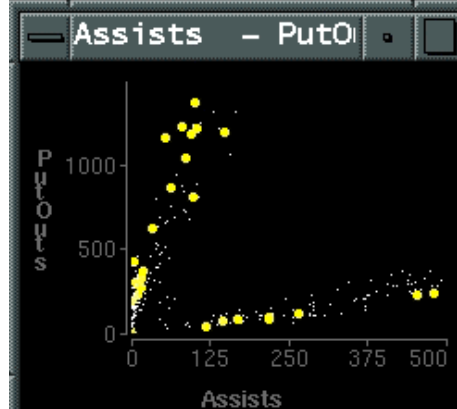
Multiform views

how long
in majors



select high
salaries

avg assists vs
avg putouts
(fielding ability)



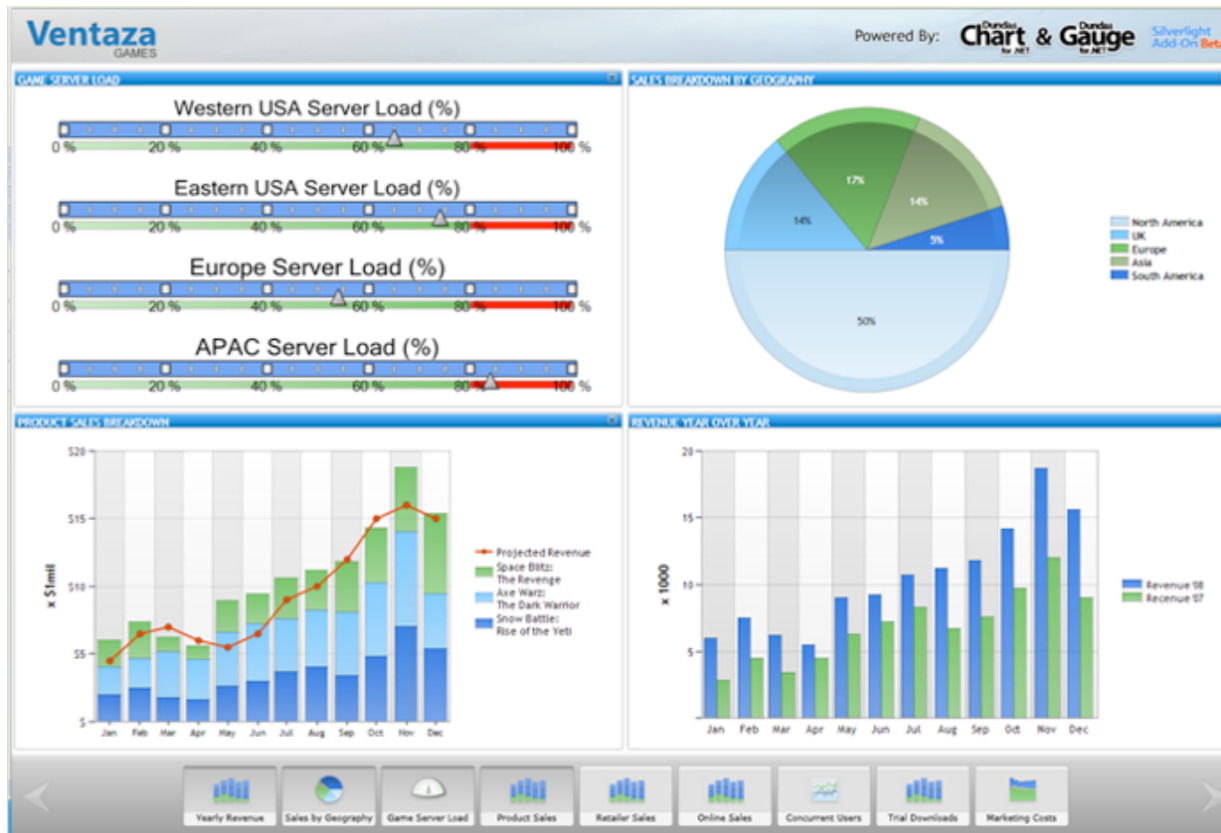
avg career
HRs vs avg
career hits
(batting ability)

distribution
of positions
played



[Eick & Wills 95]

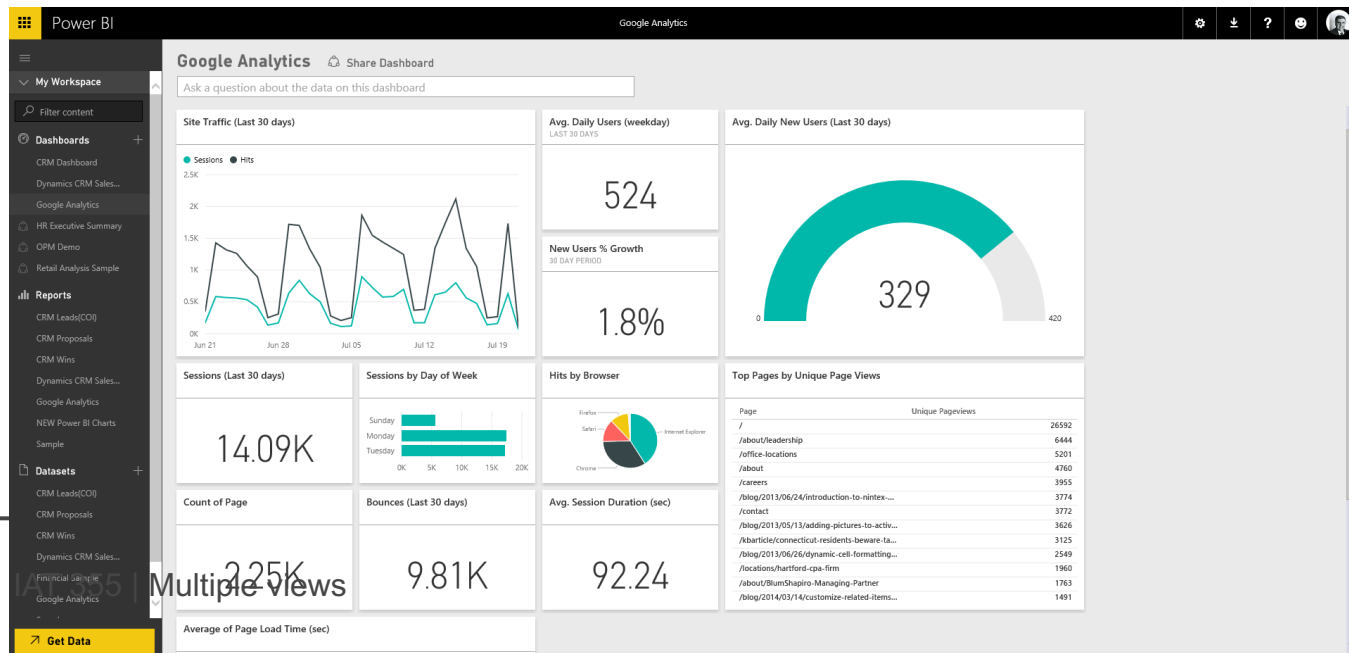
Information dashboards



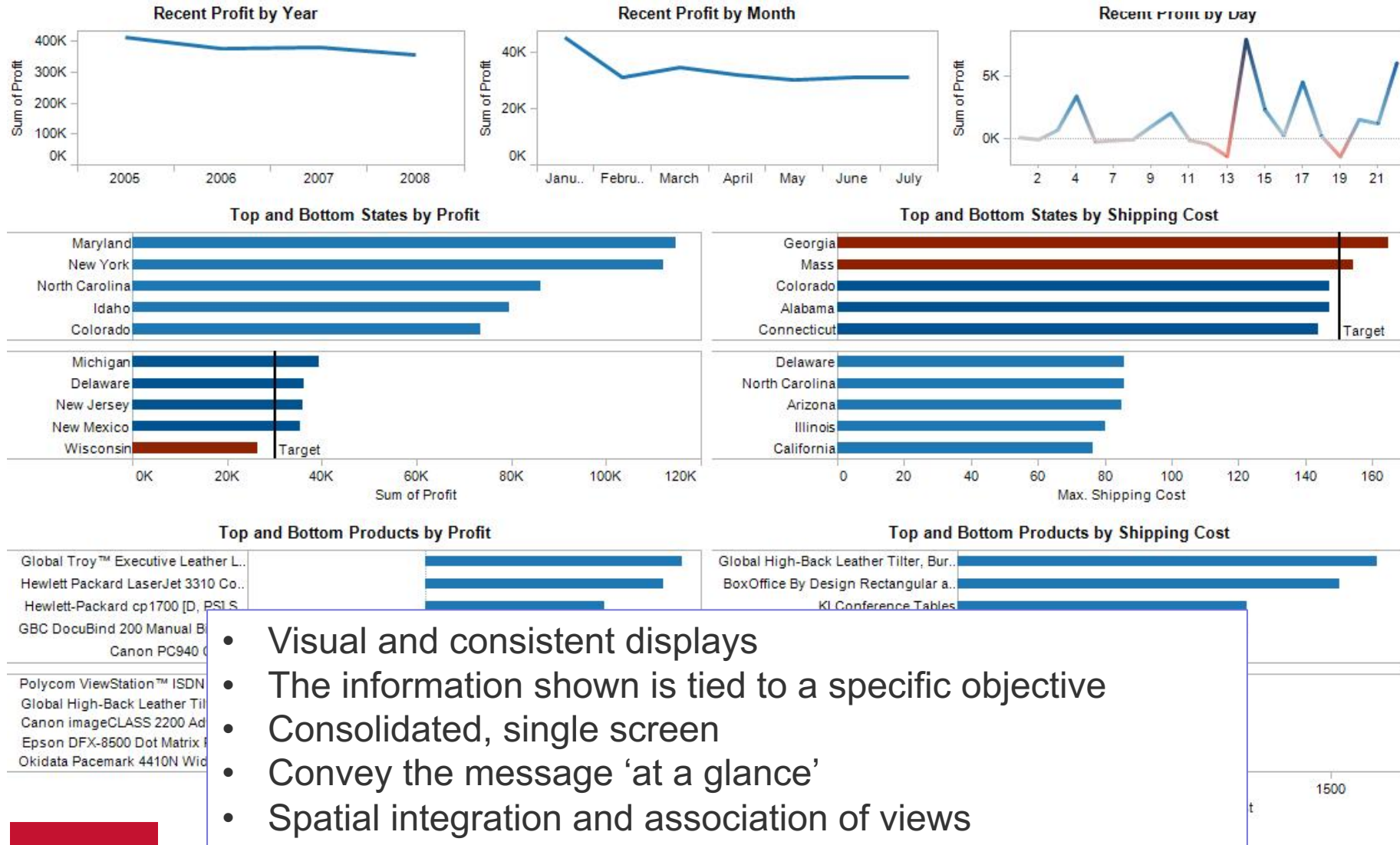
What is a “Dashboard”?

- A dashboard is a visual display of the most important information needed to achieve one or more objectives, consolidated and arranged on a single screen so the information can be monitored at a glance.*

-Stephen Few (March 20, 2004) “Dashboard Confusion” Intelligent Enterprise



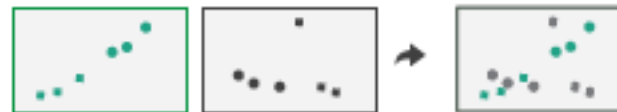
Strategic Dashboard Example



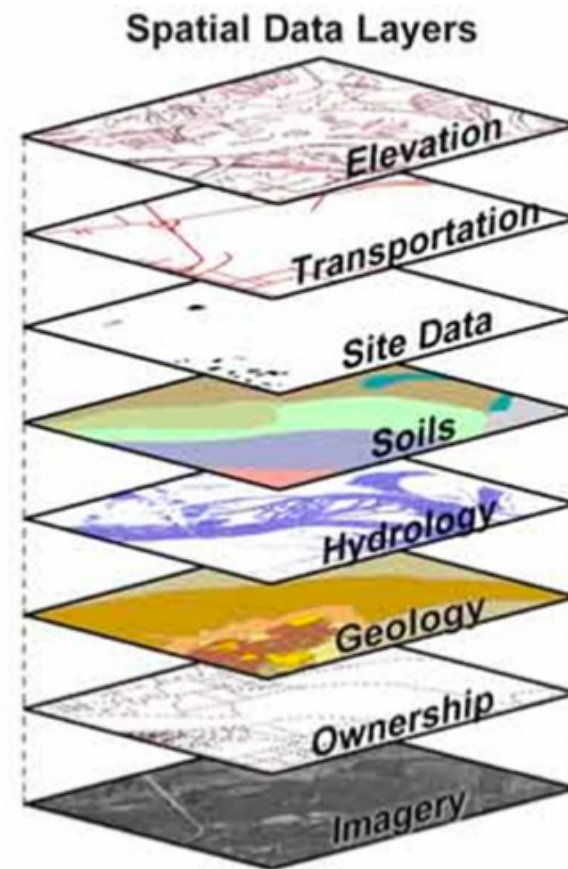
- Visual and consistent displays
- The information shown is tied to a specific objective
- Consolidated, single screen
- Convey the message 'at a glance'
- Spatial integration and association of views

Layering

➔ Superimpose Layers

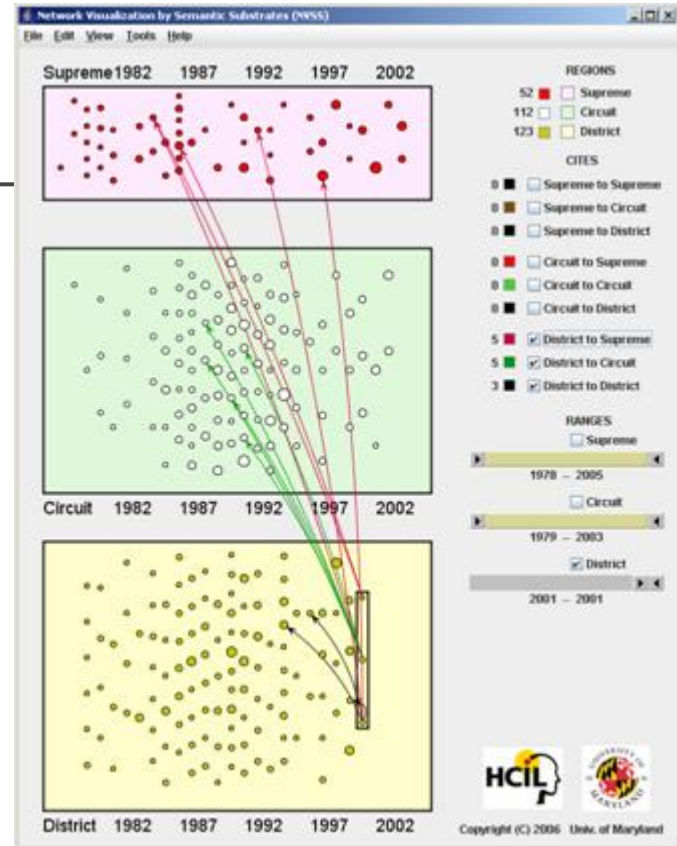
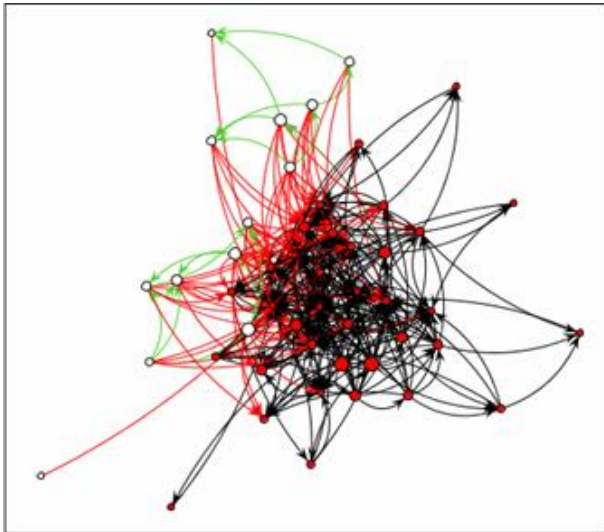


- Overlay facets
- Very common in spatial data
 - Common reference
 - Set geometry



Layering

- Multiform, small multiples

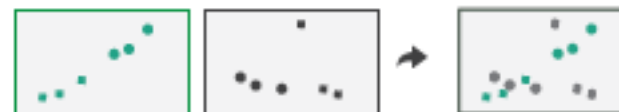


<http://hcil.cs.umd.edu/video/2006/substrates.mpg>

Shneiderman & Aris, Network Visualization by Semantic Substrates, IEEE TVCG 2006

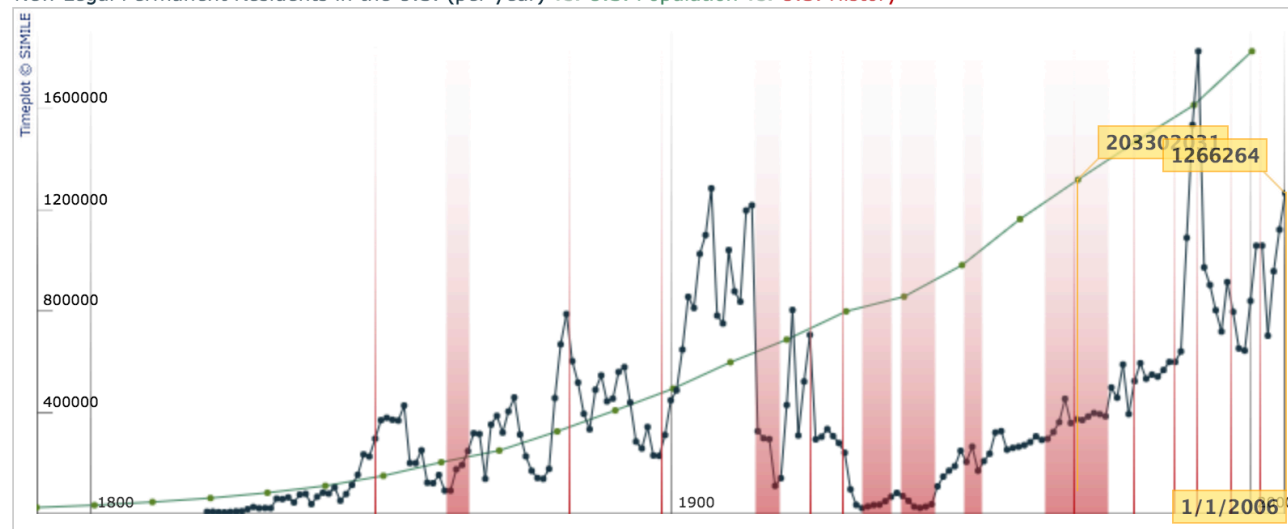
Layering

→ Superimpose Layers



- Overlay facets
- Issues?

New Legal Permanent Residents in the U.S. (per year) vs. U.S. Population vs. U.S. History

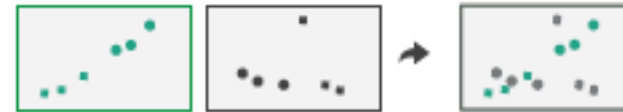


Sources: U.S. Department of Homeland Security, U.S. Census Bureau and Wikipedia

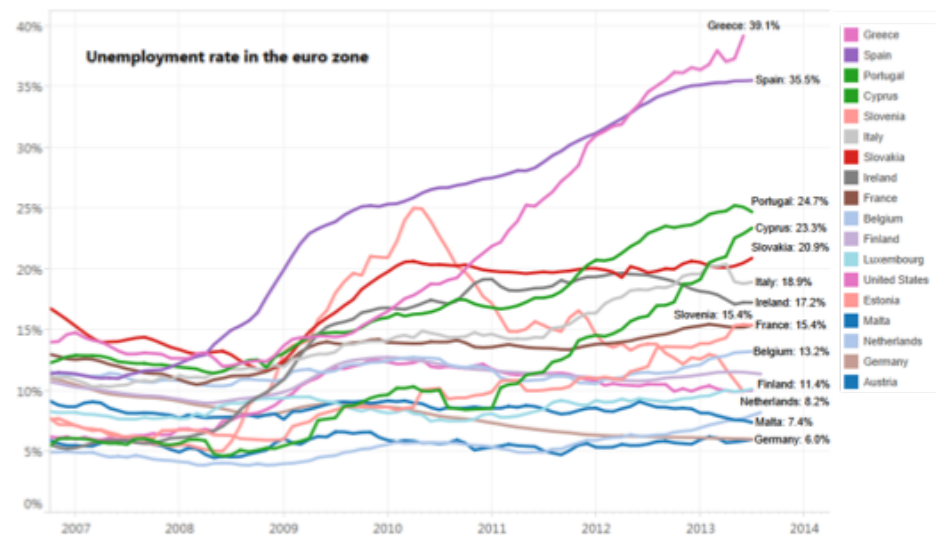
TimePlot, <http://www.simile-widgets.org/timeplot/>

Design Choices

➔ Superimpose Layers

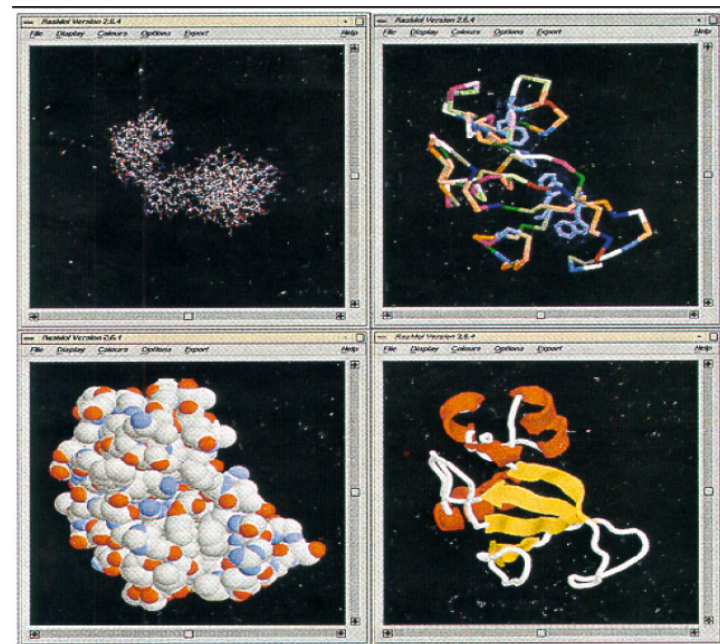
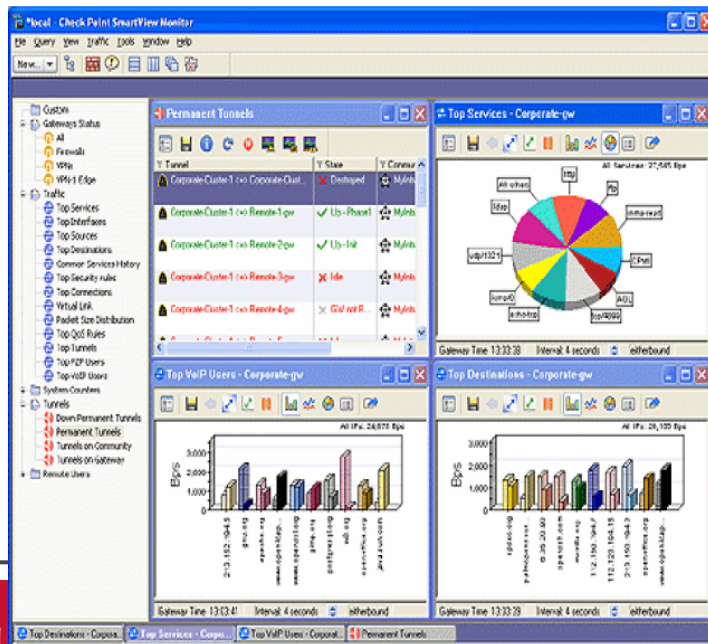


- Partitioning elements
- How many layers
 - How important to determine layers?
- How to make them distinguishable?
- Are layers static or dynamic?
 - Examples?



Multiple Views

- “Guidelines for Using Multiple Views in Information Visualization”
 - Baldonado, Woodruff and Kichinsky AVI 00



Multiple Views: 8 Guidelines

- Rule of Diversity:
 - Use multiple views when there is a diversity of attributes
- Rule of Complementarity:
 - Multiple views should bring out correlations and/or disparities
- Rule of Decomposition: “Divide and conquer”.
 - Help users visualize relevant chunks of complex data
- Rule of Parsimony:
 - Use multiple views minimally

8 Guidelines Cont'd

- Rule of Space/Time Resource
 - Optimization: Balance spatial and temporal benefits of presenting and using the views
- Rule of Self Evidence:
 - Use cues to make relationships apparent.
- Rule of Consistency:
 - Keep views and state of multiple views consistent
- Rule of attention management:
 - Use perceptual techniques to focus user attention