

# Lecture 2, InfoVis MiniCourse

**Space/Order, Depth/Occlusion,  
High Dimensionality, Interaction**

**LaBRI, University of Bordeaux  
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# Mini-Course Outline

Perception

Frameworks

Color

Space/Order

Depth/Occlusion

High Dimensionality

Interaction

Navigation/Zooming

Focus+Context

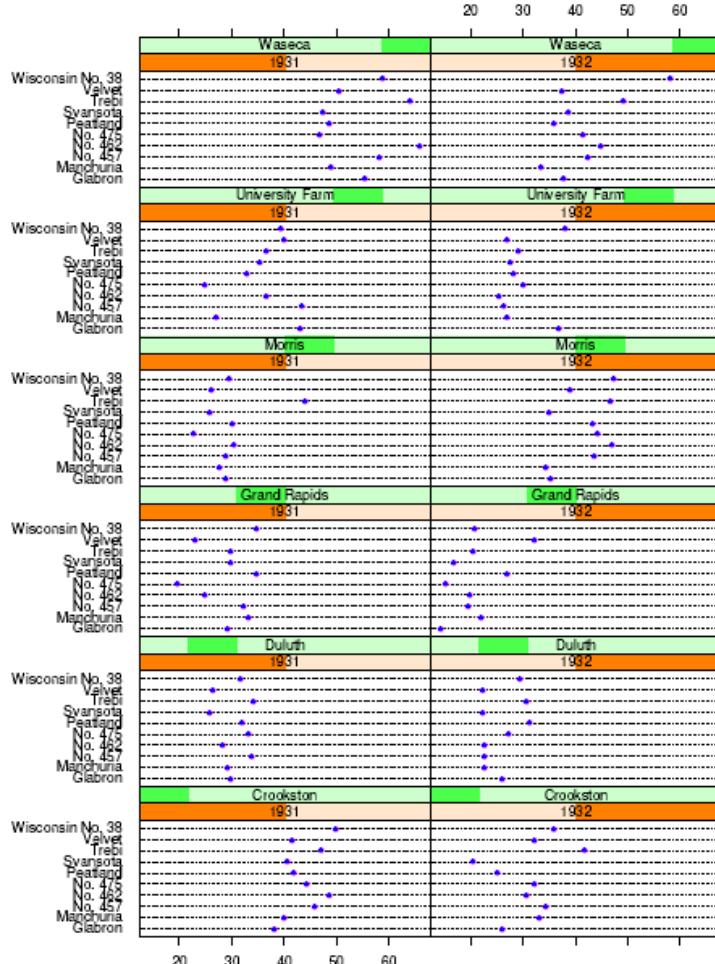
Graphs/Trees

Scalability

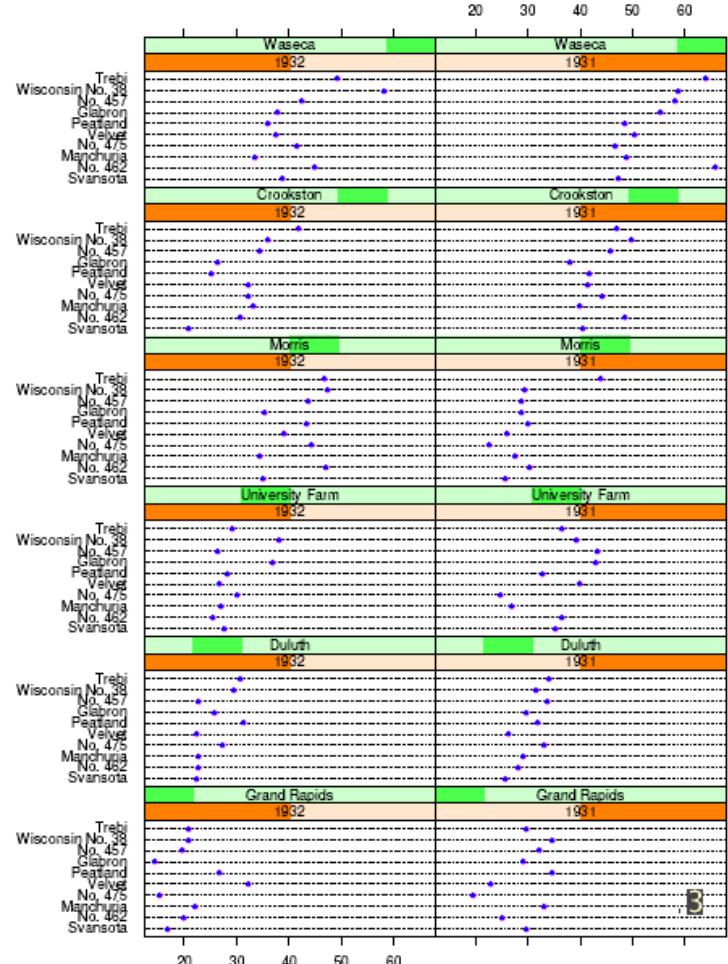
Task-Centered Design

# Finding the Right Order: Trellis

alphabetical



main-effects: sort by median value



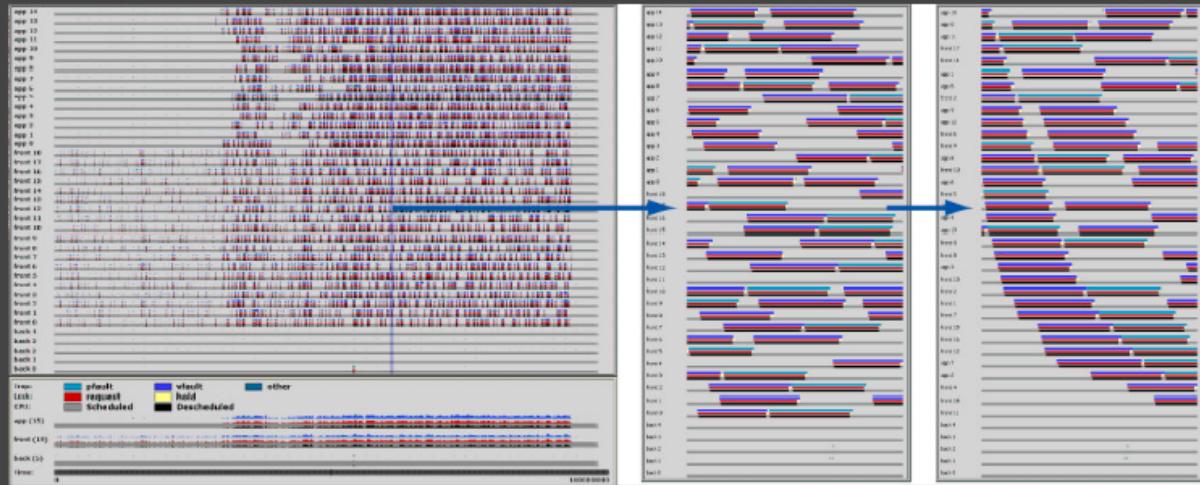
# Finding the Right Order: Table Lens

## select column to sort

- user explores

[demo: [www.tablelens.com](http://www.tablelens.com)]

# Finding the Right Order: Rivet



[Bosch, Performance Analysis and Visualization of Parallel Systems Using SimOS and Rivet: A Case Study, HPCA6, 2000.  
graphics.stanford.edu/papers/rivet\_argus]

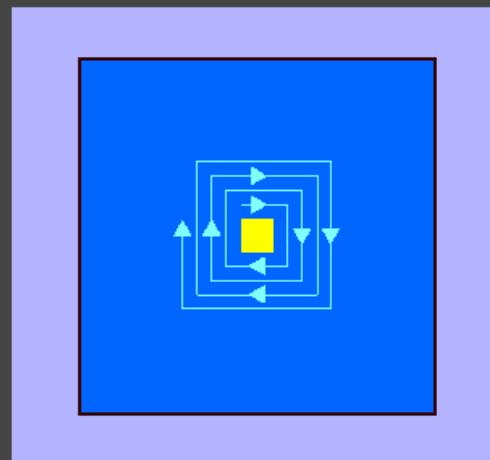
# Finding the Right Order: VisDB

each pixel represents database entry

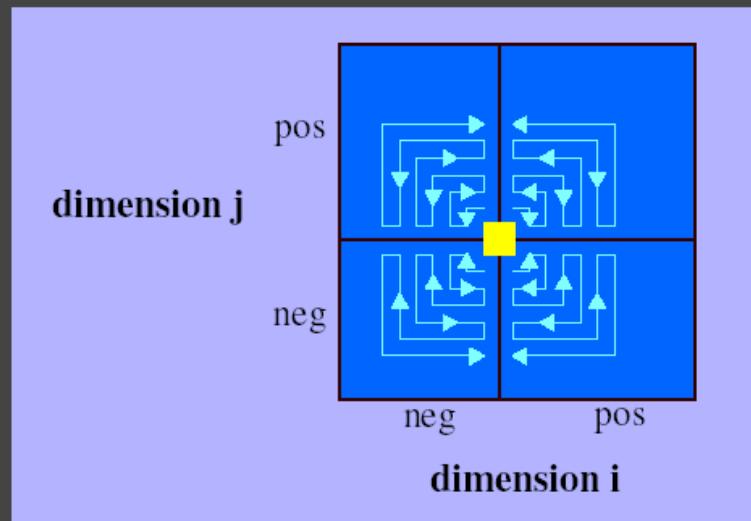
- color by query relevance

what order to sort?

spiral



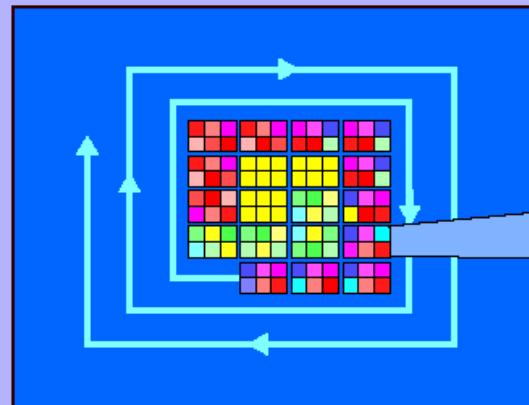
2D



# VisDB Windows

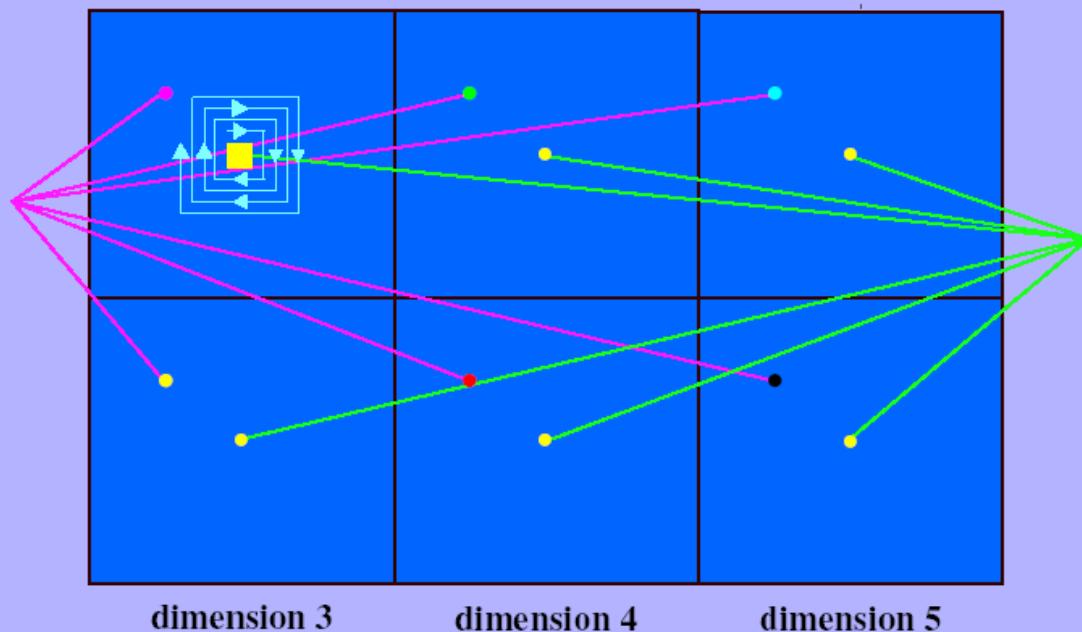
group dimensions

separate dimensions



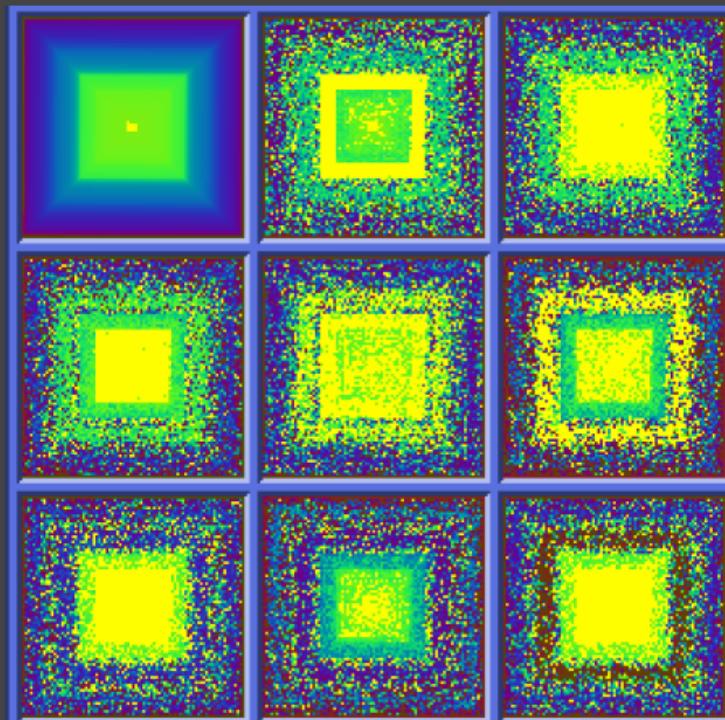
dimension 1 dimension 2  
[www.dbs.informatik.uni-muenchen.de/dbs/projekt/papers/visdb.ps](http://www.dbs.informatik.uni-muenchen.de/dbs/projekt/papers/visdb.ps)

one data item  
approximately  
fulfilling the  
query

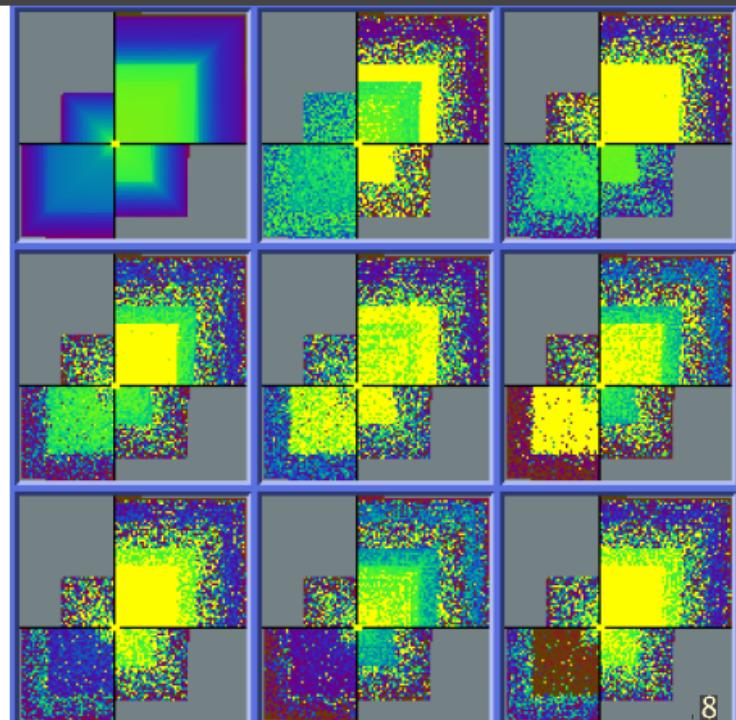


# VisDB Results: Separate Dimensions

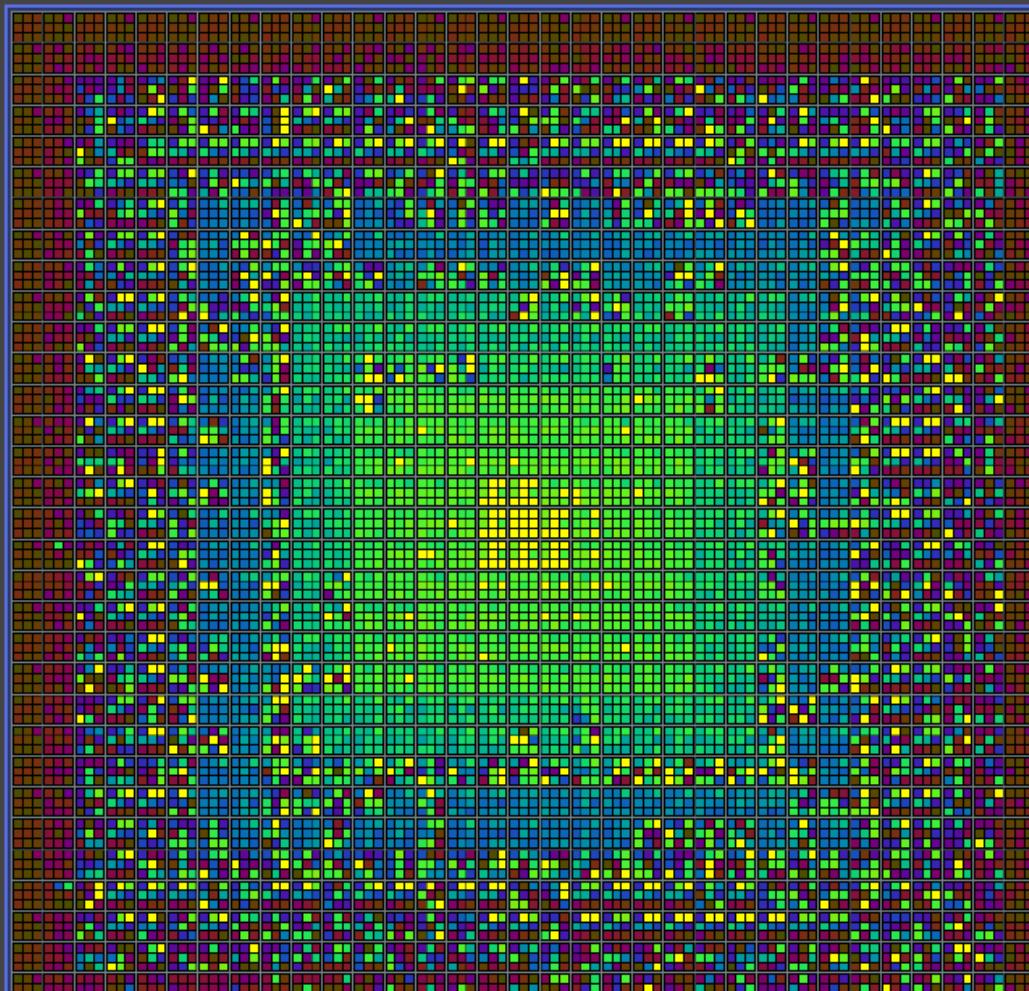
spiral



2D



# VisDB Results: Grouped Dimensions

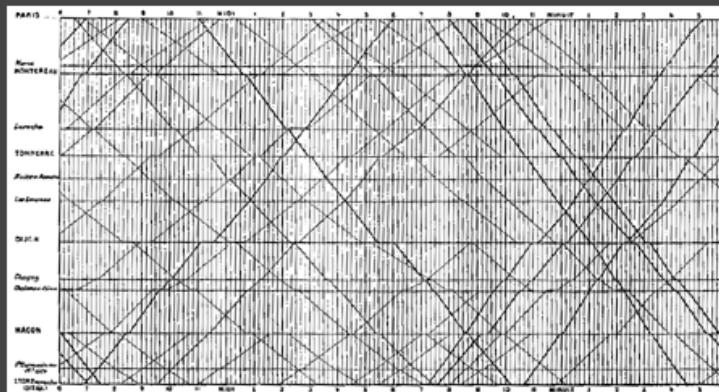


# Derived Spaces: Slope

narrative of space and time

Marey train schedule, 1885

- horizontal line length: stop length
- slope: speed
- intersection: time/place of crossing



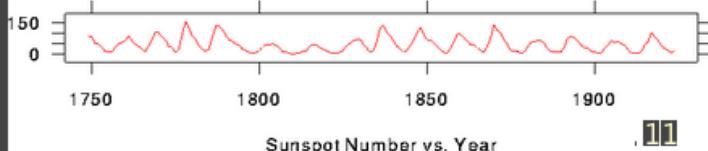
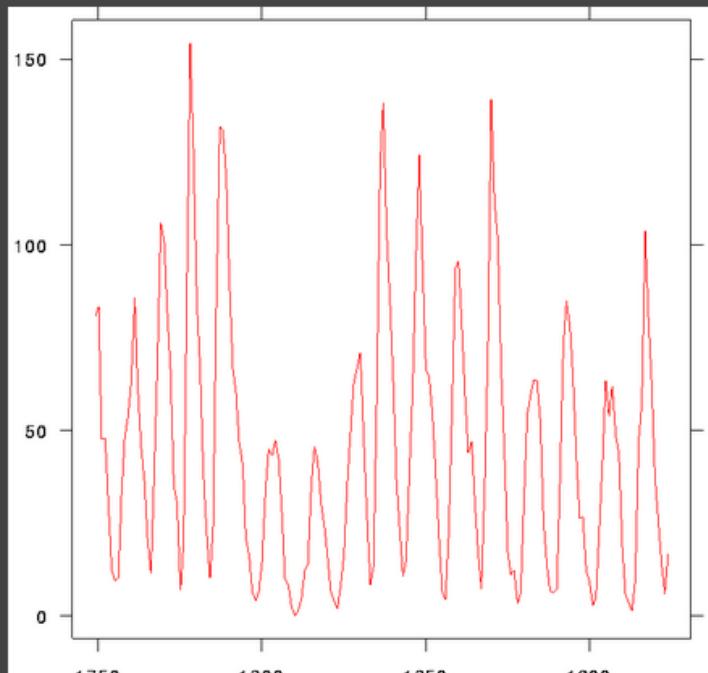
[Tufte I p 31, [www.nap.edu/html/hs\\_math/images/tl\\_f8.gif](http://www.nap.edu/html/hs_math/images/tl_f8.gif)]

# Slope: Banking to 45 Degrees

what size should a line graph be?

perceptual principle: most accurate slope judgement near 45 degrees

pick aspect ratio  
(height/width) to maximize amount of 45 degree slope



# Space vs. Time: Showing Change



animation: show time using temporal change  
good

- show process
- compare two things by blinking

bad

- directly compare between many things
  - intermediate frames "overload mental buffer"

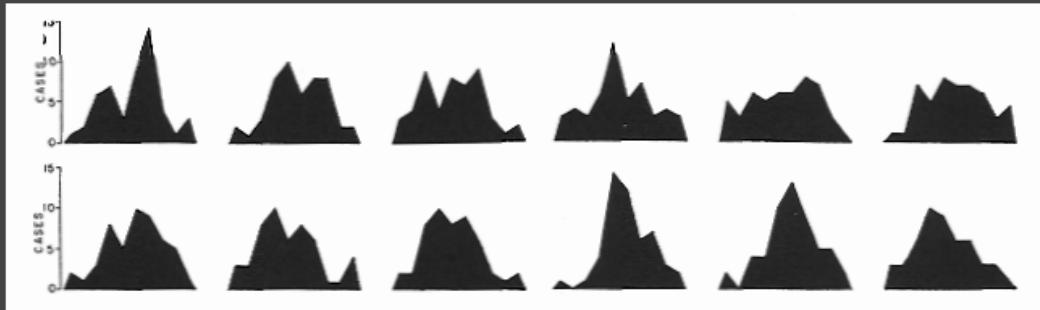


# Space vs. Time: Showing Change



small multiples: show time using space

- overview: show each time step in array
- compare: side-by-side easier than temporal  
external cognition instead of internal memory
- general technique, not just for temporal changes



[Edward Tufte. The Visual Display of Quantitative Information, p 172]

# More Reading

The Visual Design and Control of Trellis Display

R. A. Becker, W. S. Cleveland, and M. J. Shyu

Journal of Computational and Statistical Graphics, 5:123–155. (1996).

<http://cm.bell-labs.com/stat/doc/trellis.jcgs.col.ps>

Envisioning Information. Edward Tufte. Graphics Press, 1990.

Chapter 4: Small Multiples, Chapter 6: Narratives of Space and Time

The Elements of Graphing Data, William S. Cleveland, Hobart Press 1994.

VisDB: Database Exploration using Multidimensional Visualization,

Daniel A. Keim and Hans-Peter Kriegel, IEEE CG&A, 1994

<http://www.dbs.informatik.uni-muenchen.de/dbs/projekt/papers/visdb.ps>

The Table Lens: Merging Graphical and Symbolic Representations in an Interactive Focus + Context Visualization for Tabular Information

Ramana Rao and Stuart K. Card, SIGCHI '94, pp. 318–322.

<http://citeseer.ist.psu.edu/545353.html>

Performance Analysis and Visualization of Parallel Systems Using SimOS and Rivet: A Case Study. Robert Bosch, Chris Stolte, Gordon Stoll, Mendel Rosenblum, and Pat Hanrahan. In Proc. Sixth IEEE International Symposium on High-Performance Computer Architecture 2000. [http://graphics.stanford.edu/papers/rivet\\_argus/](http://graphics.stanford.edu/papers/rivet_argus/)

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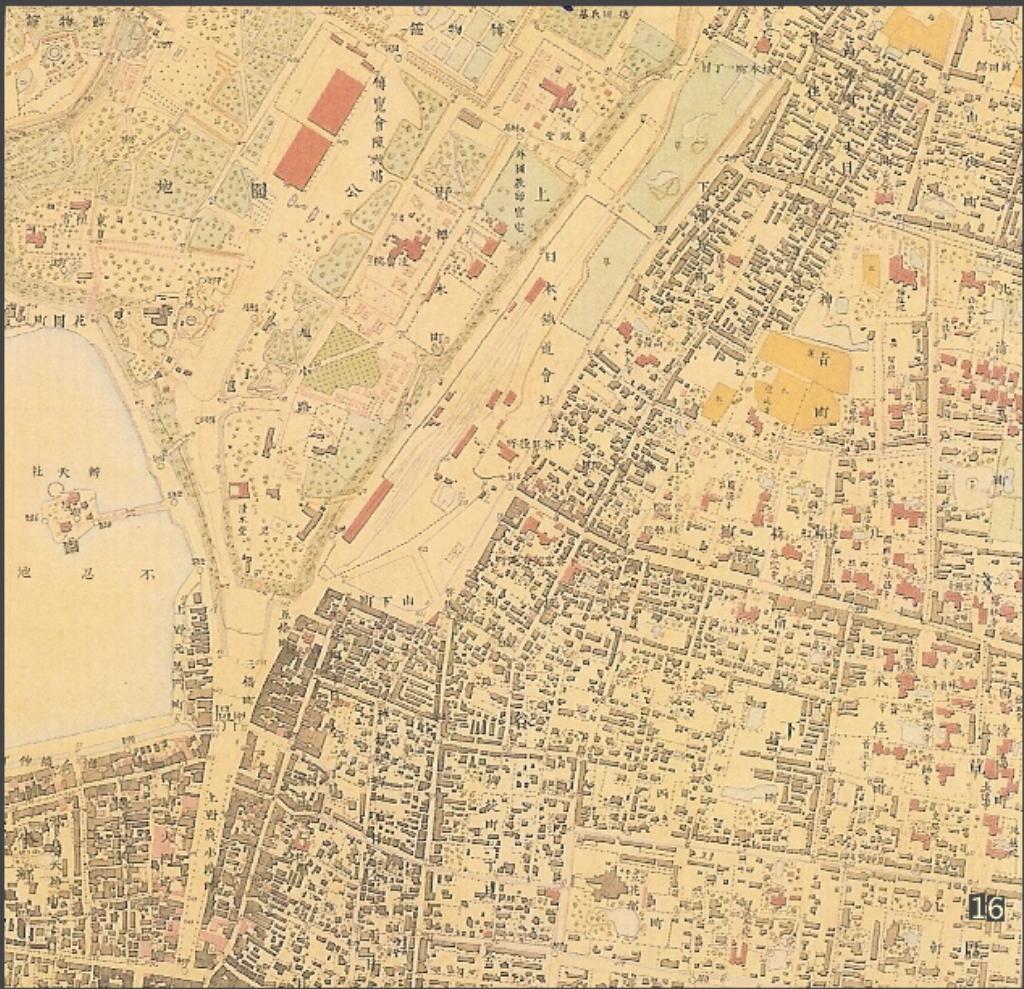
Focus+Context

Graphs/Trees

Scalability

Task-Centered Design

# Layering And Separation



# Visual Clutter

subtler background than foreground



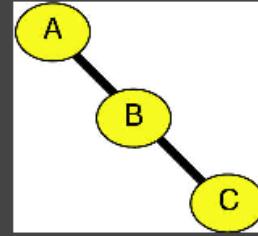
[Tufte, Envisioning Information, Chap 3]

# Visual Layering For Graphs

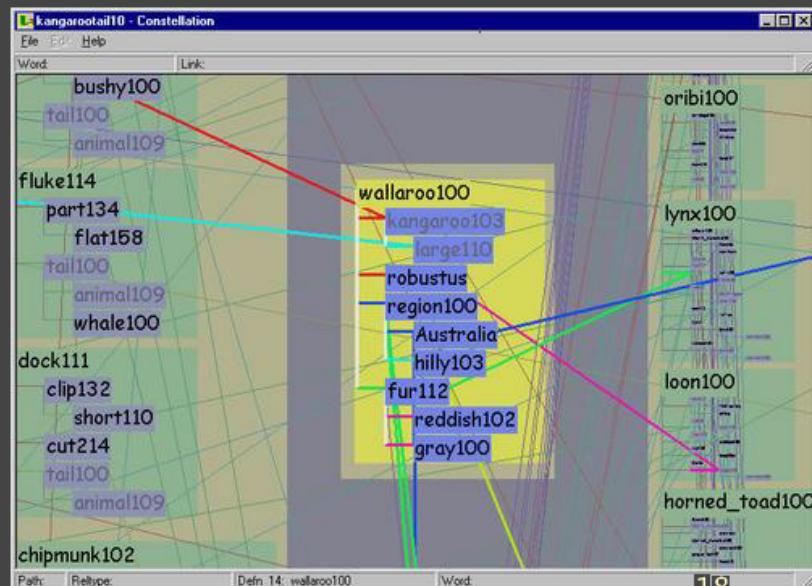
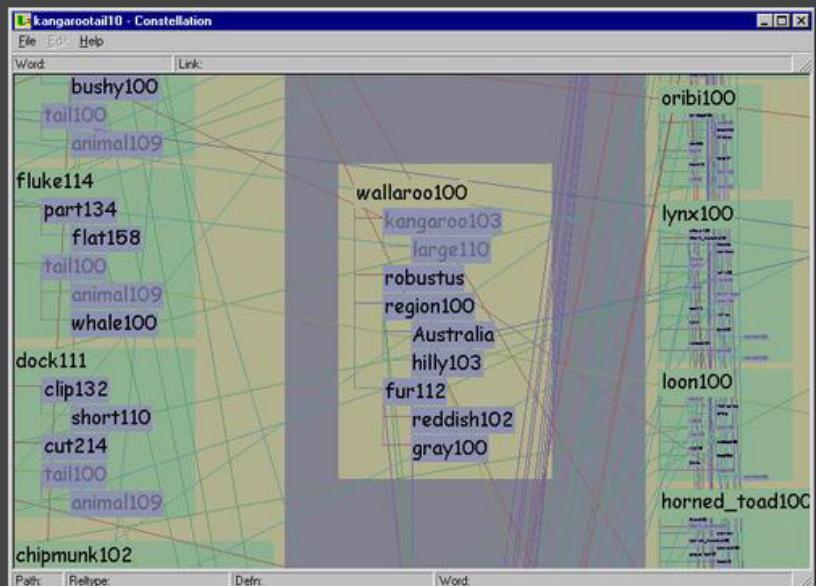
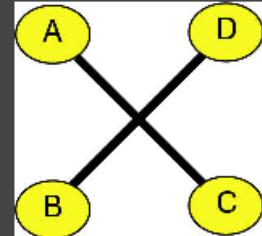
edge crossing problem  
· false attachments

layers to avoid perception  
· vs. spatial position

ambiguity



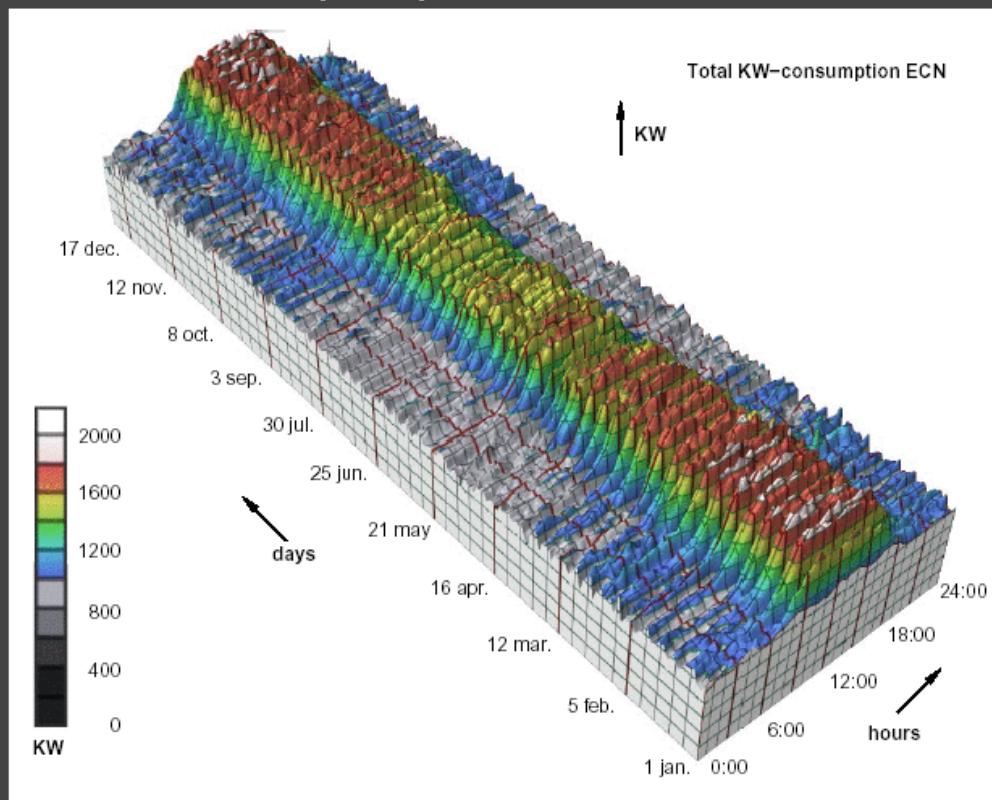
artifact salience



# 3D Time-series Data

3D extrusion pretty but not useful

- daily, weekly patterns hard to see
- occlusion hides, perspective interferes



# Time-series Data Analysis

data: N pairs of (value, time)

- N large: 50K

tasks

- find standard day patterns
- find how patterns distributed over year, week, season
- find outliers from standard daily patterns
- want overview first, then detail on demand

possibilities

- predictive mathematical models
  - details lost, multiscale not addressed
- scale-space approaches (wavelet, fourier, fractal)
  - hard to interpret, known scales lost
- 3D mountain: x hours, y value, z days
  - occlusion hides, perspective interferes

# Hierarchical Clustering

start with all  $M$  day patterns

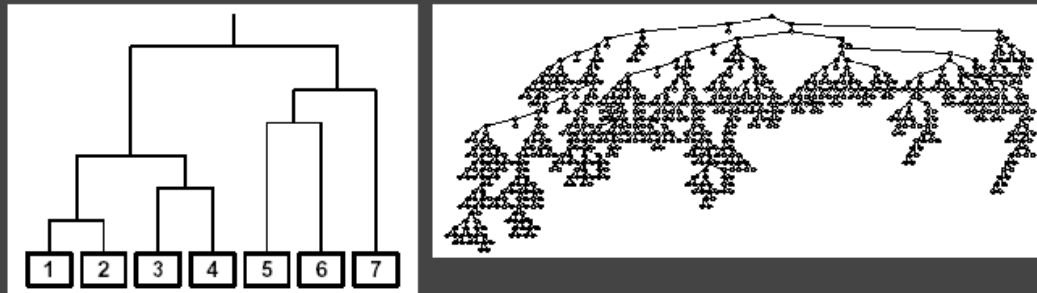
- compute mutual differences, merge most similar
- continue up to 1 root cluster

result: binary hierarchy of clusters

- choice of distance metrics

dendrogram display common

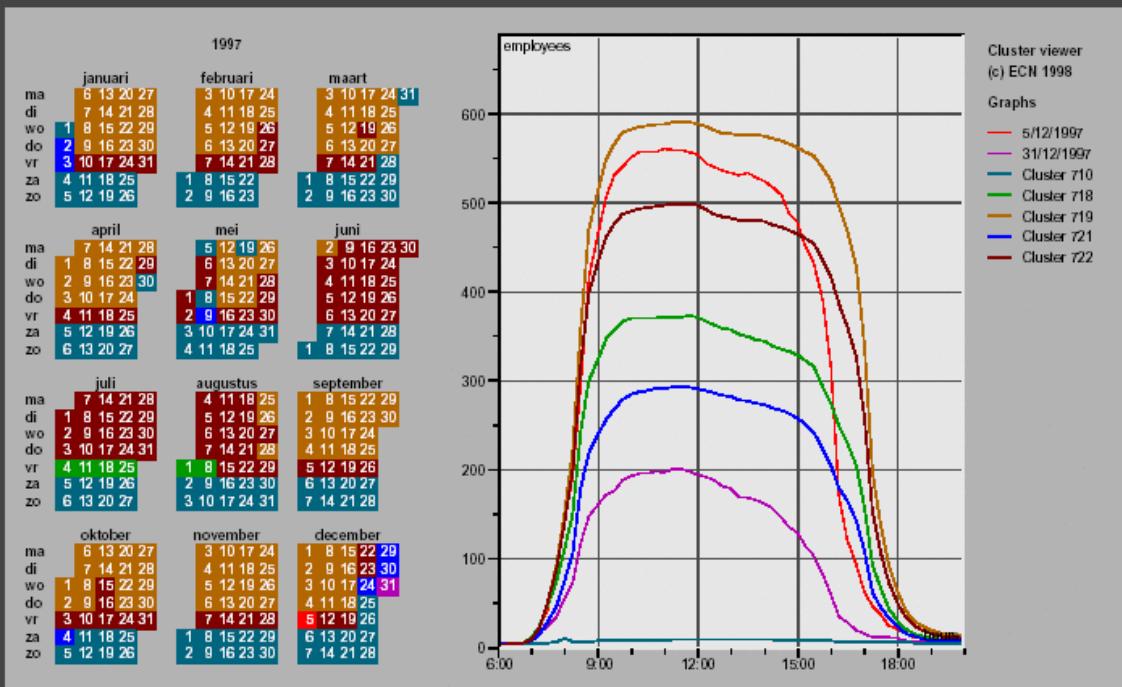
- shows structure of hierarchy
- does not solve pattern finding problem!



# Link Clusters and Calendar

2D linked clusters-calendars shows patterns

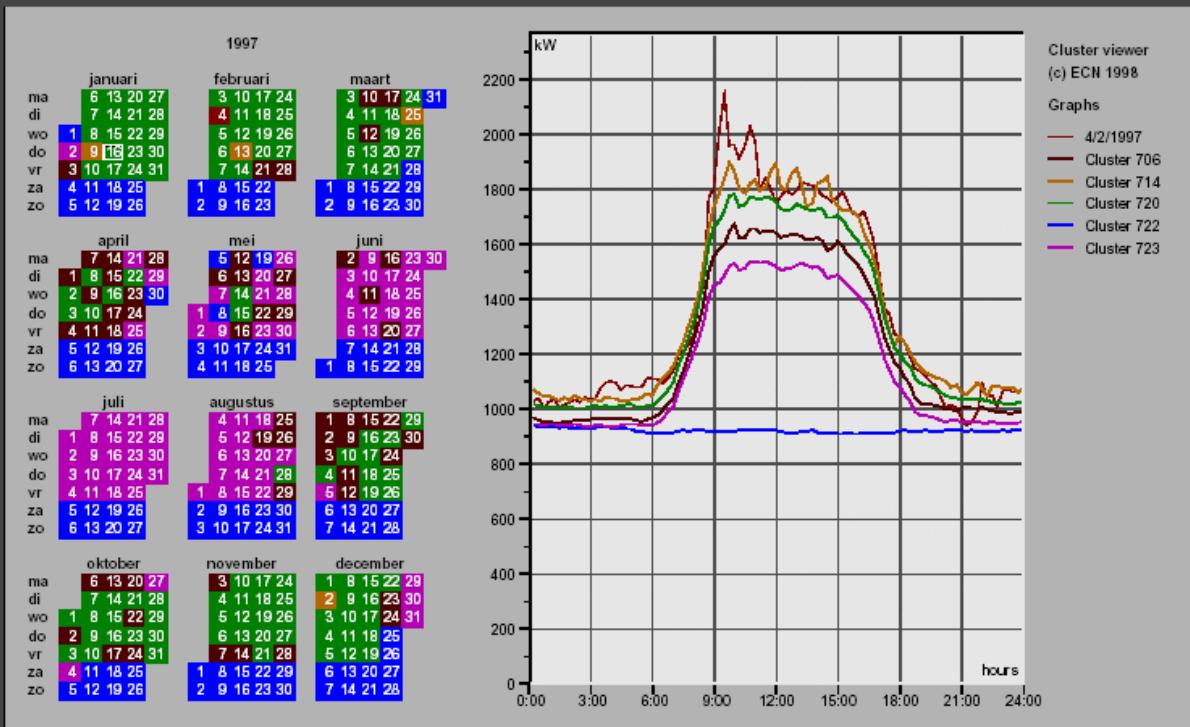
- number of employees:
- office hours, fridays in/and summer, school break
- weekend/holidays, post-holiday, santa claus



[van Wijk and van Selow, Cluster and Calender based Visualization of Time Series Data, InfoVis99,

Figure 4, citeseer.nj.nec.com/vanwijk99cluster.html]

# Power Consumption



[van Wijk and van Selow, Cluster and Calender based Visualization of Time Series Data, InfoVis99,  
Figure 5, citeseer.nj.nec.com/vanwijk99cluster.html]

# van Wijk Lessons

derived space: clusters

visual representation of time: calendar

- linked display
- interactive exploration

clear task analysis guided choices

- reject standard 3D extrusion
- reject standard dendrogram

critique

- pro: great design study, problem solved!
- con: some chosen colors not discriminable

# 3DPS

1: 2D displace+magnify

2: 3D displace+magnify

3: 2D displace only

4: 3D displace only

5: visual access distortion

Extending Distortion Viewing  
Techniques from 2D to 3D Data  
Carpendale et al 1997

|   | Stretch Orthogonal | Non-Linear Orthogonal | Non-Linear Radial | Step Orthogonal |
|---|--------------------|-----------------------|-------------------|-----------------|
| 1 |                    |                       |                   |                 |
| 2 |                    |                       |                   |                 |
| 3 |                    |                       |                   |                 |
| 4 |                    |                       |                   |                 |
| 5 |                    |                       |                   |                 |

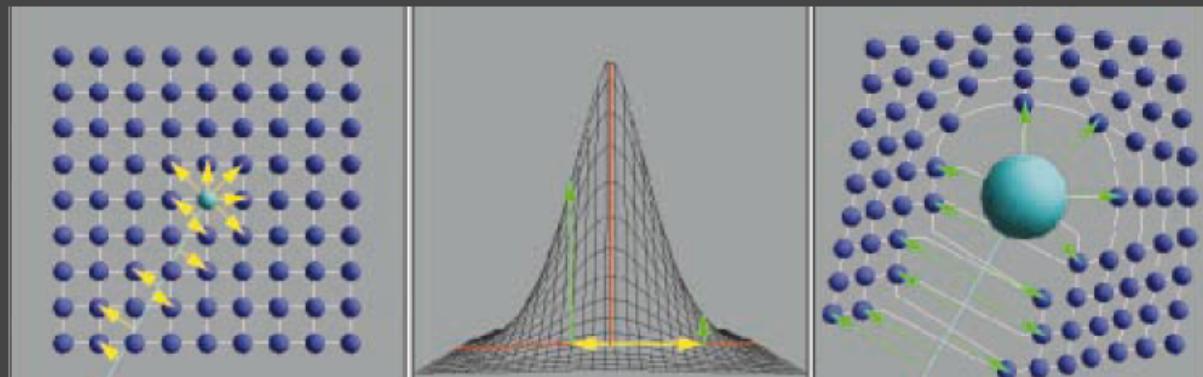
# Visual Access Distortion

naive 2D → 3D extension yields occlusion

- same problem as van Wijk

graph-based solution

- move geometry according to viewpoint
- magnify focus only
- introduce curves into formerly straight lines

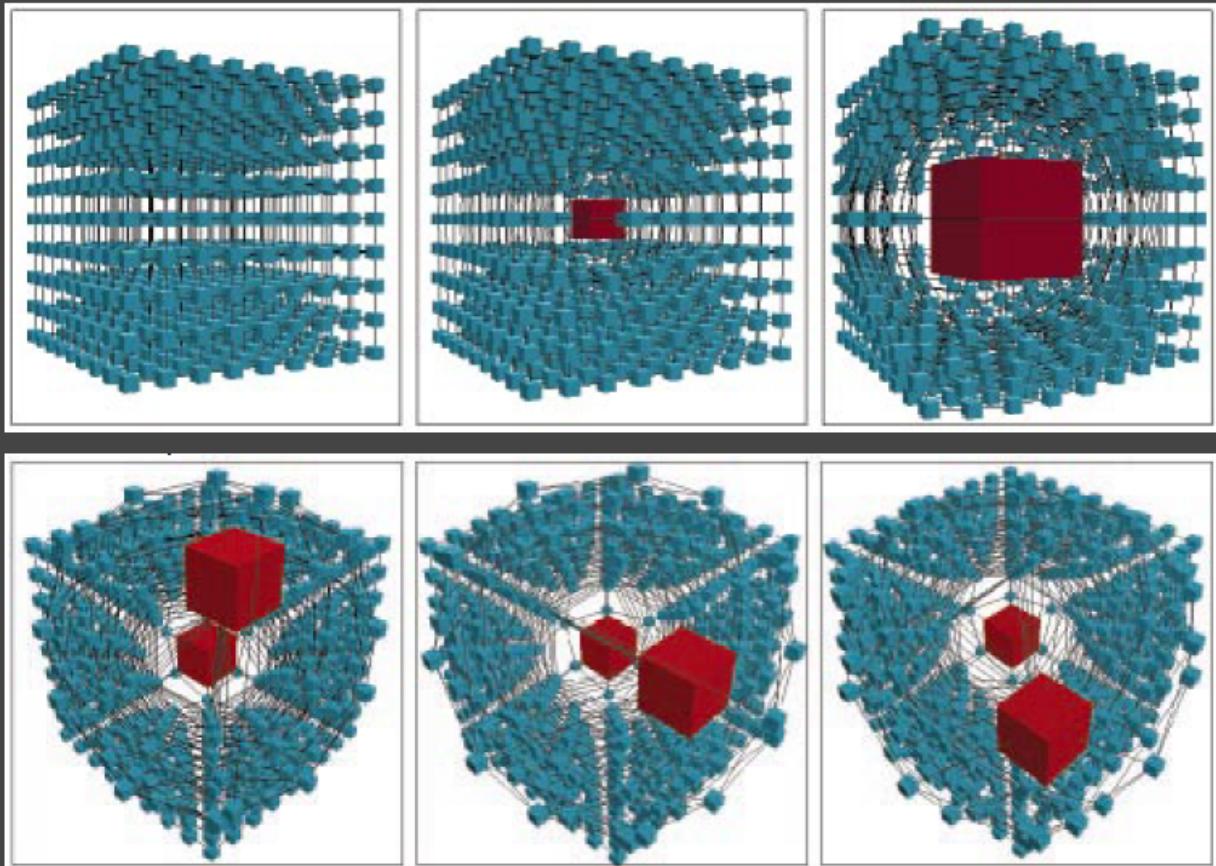


focus+context issues discussed later

[[pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/cga97.pdf](http://pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/cga97.pdf)]

# Results

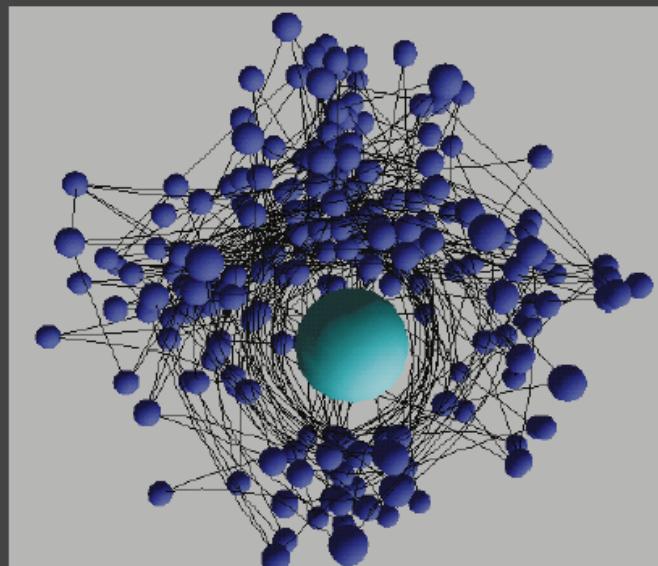
single, multiple foci



# Results

randomly positioned nodes instead of grid

- closer to real dataset



[[pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/cga97.pdf](http://pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/cga97.pdf)]

# Critique

sophisticated way to navigate 3D graphs

nice technique paper

- not a design study

interesting discussion I'd like to see

- more analysis of why 3D necessary
- cites Ware 3x improvement
- occlusion workaround vs. occlusion avoidance

never shown on real data

- hard to draw conclusions from toy datasets

# EdgeLens

interactive control over edge occlusion

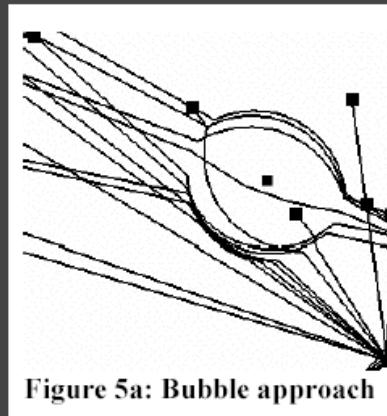
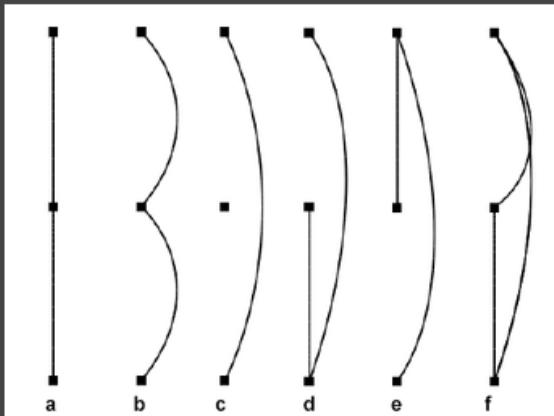


Figure 5a: Bubble approach

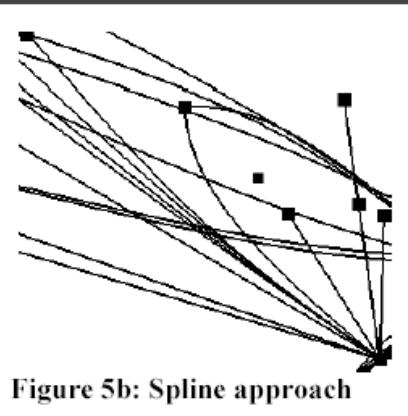
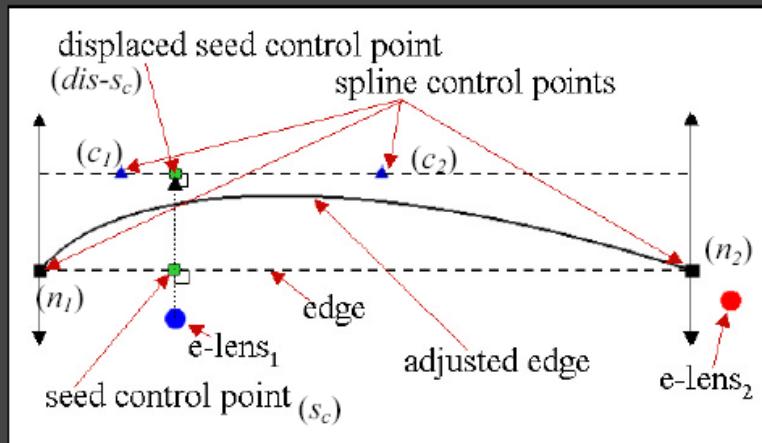


Figure 5b: Spline approach

user study: spline better than bubble

[EdgeLens: An Interactive Method for Managing Edge Congestion in Graphs  
Nelson Wong, M. Sheelagh T. Carpendale, Saul Greenberg, Proc. InfoVis03, pp 51–58,  
[pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/2003/wong-carp-infovis03-submit.pdf](http://pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/2003/wong-carp-infovis03-submit.pdf)]<sup>30</sup>

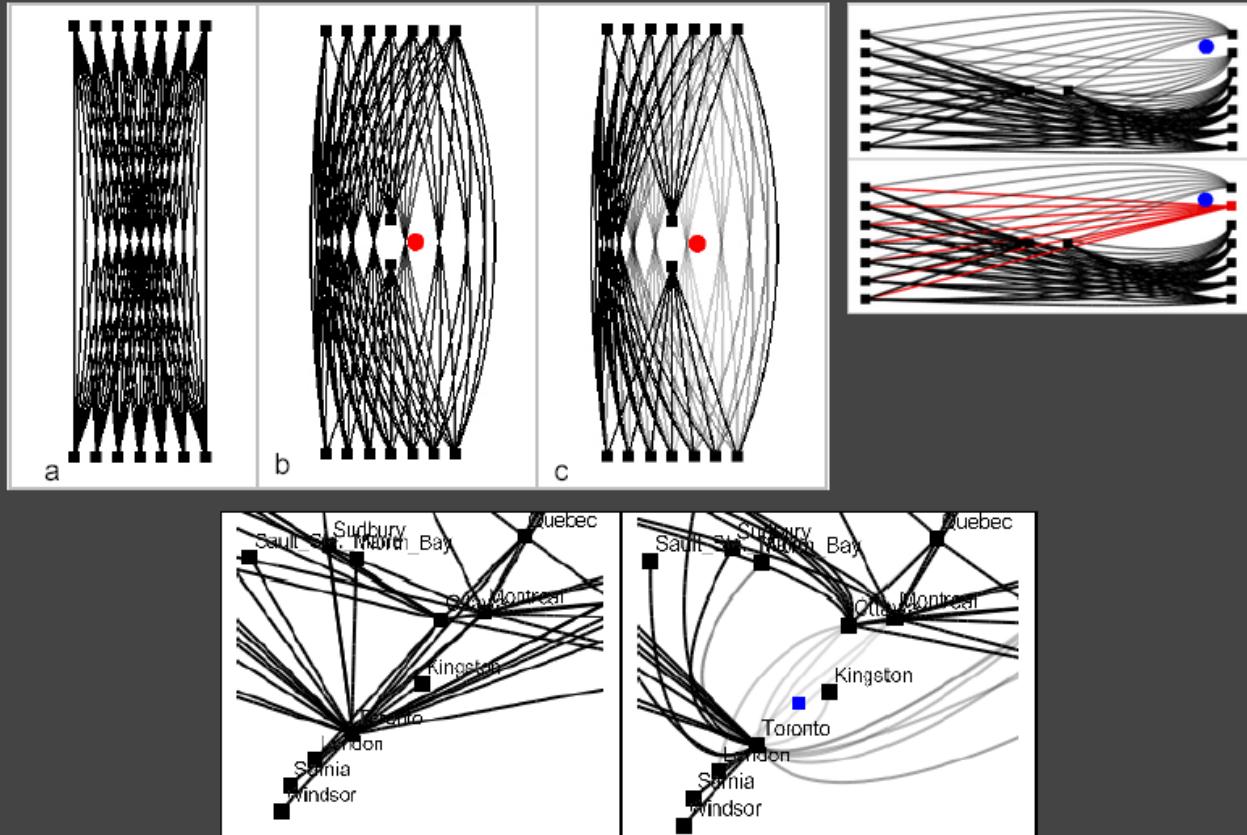
# EdgeLens Final Algorithm



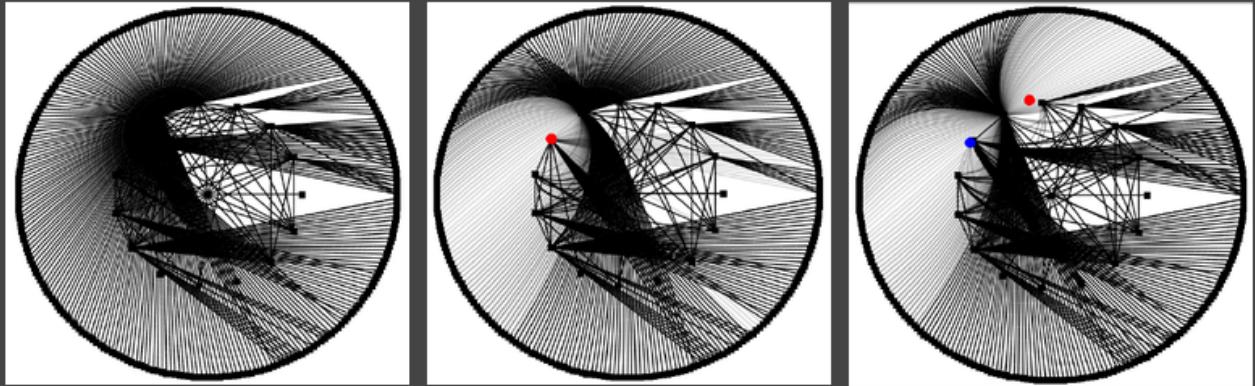
decide which edges affected  
calculate displacements  
calculate spline control points  
draw curves

# EdgeLens Techniques

transparency, color



# EdgeLens Results



[[pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/2003/wong-carp-infovis03-submit.pdf](http://pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/2003/wong-carp-infovis03-submit.pdf)]

## critique

- very nice technique
- compelling need
- shown on real data

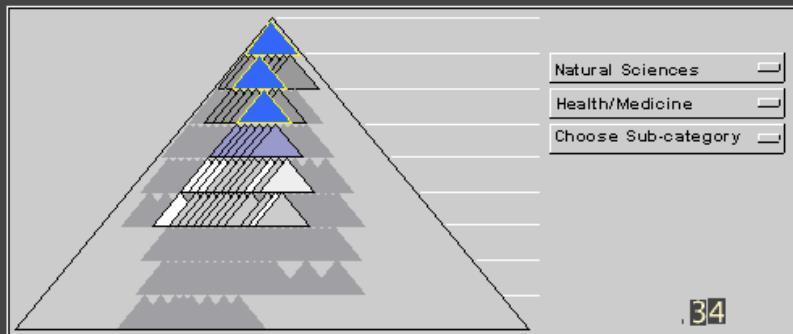
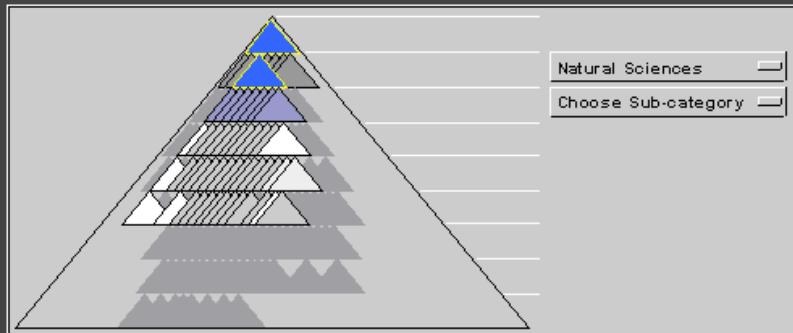
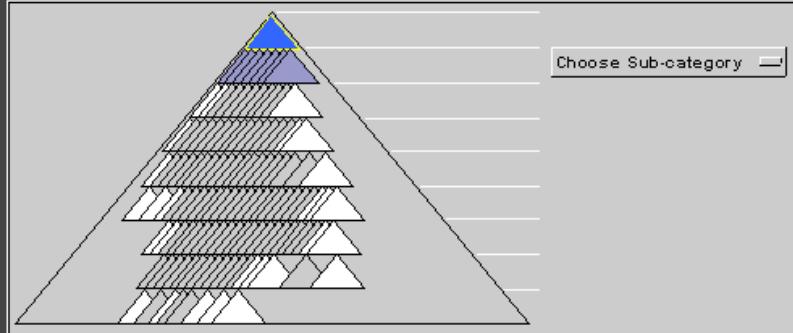
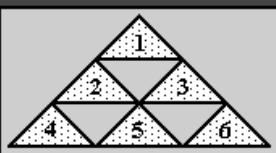
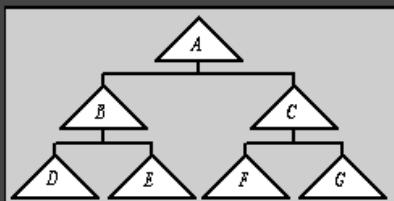
# Cheops

compact

show paths through tree

extreme occlusion  
deliberately

browsing/exploration, not  
topological analysis



# Cheops Critique

pro

- tiny footprint  
suitable when main user focus is other task
- interaction techniques investigated  
informal usability

con

- relatively hard to understand
- singular nodes very salient, but not so important

# More Reading

Envisioning Information. Edward Tufte. Graphics Press, 1990.  
Chapter 3: Layering and Separation

Tamara Munzner, Interactive Visualization of Large Graphs and Networks  
Chapter 5, Constellation: Linguistic Semantic Networks  
PhD thesis, Stanford University, 2000, pp 87–122  
[http://graphics.stanford.edu/papers/munzner\\_thesis/html/node10.html](http://graphics.stanford.edu/papers/munzner_thesis/html/node10.html)

Cluster and Calendar based Visualization of Time Series Data  
Jarke J. van Wijk Edward R. van Selow, Proc InfoVis 99.  
<http://citeseer.nj.nec.com/vanwijk99cluster.html>

Extending Distortion Viewing Techniques from 2D to 3D Data  
M.S.T. Carpendale, David J. Cowperthwaite, and F. David Fracchia, IEEE CG&A, 17(4), pp 42 – 51, July 1997.  
<http://pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/cga97.pdf>

EdgeLens: An Interactive Method for Managing Edge Congestion in Graphs  
Nelson Wong, M. Sheelagh T. Carpendale, Saul Greenberg, Proc. InfoVis03, pp 51–58.  
<http://pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/2003/wong-carp-infovis03-submit.pdf>

Cheops: A Compact Explorer For Complex Hierarchies  
Luc Beaudoin, Marc-Antoine Parent, Louis C. Vroomen, Proc.  
Visualization 1996, pp 87–92. <http://www.istop.com/~maparent/paper.html>

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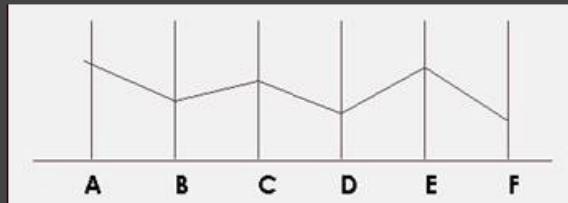
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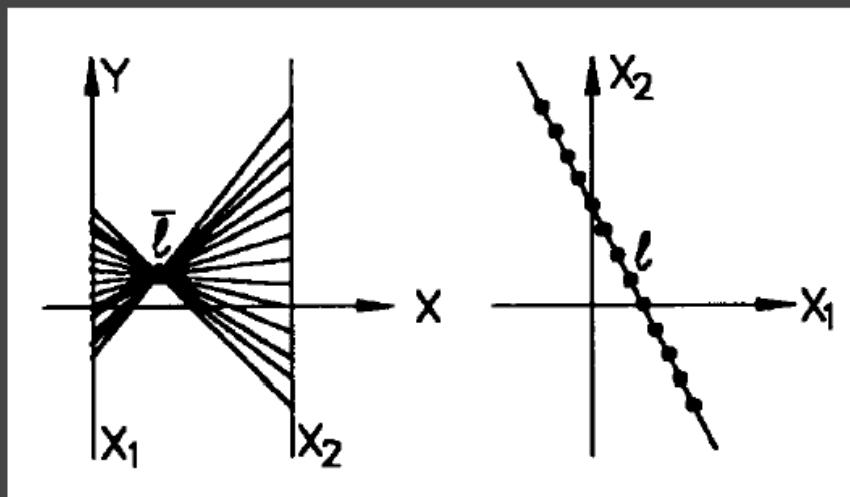
Task-Centered Design

# Parallel Coordinates

only 2 orthogonal axes in the plane  
instead, use parallel axes!



point-line duality



# Parallel Coords: Axis Ordering

geometric interpretations

- hyperplane, hypersphere: points have intrinsic order

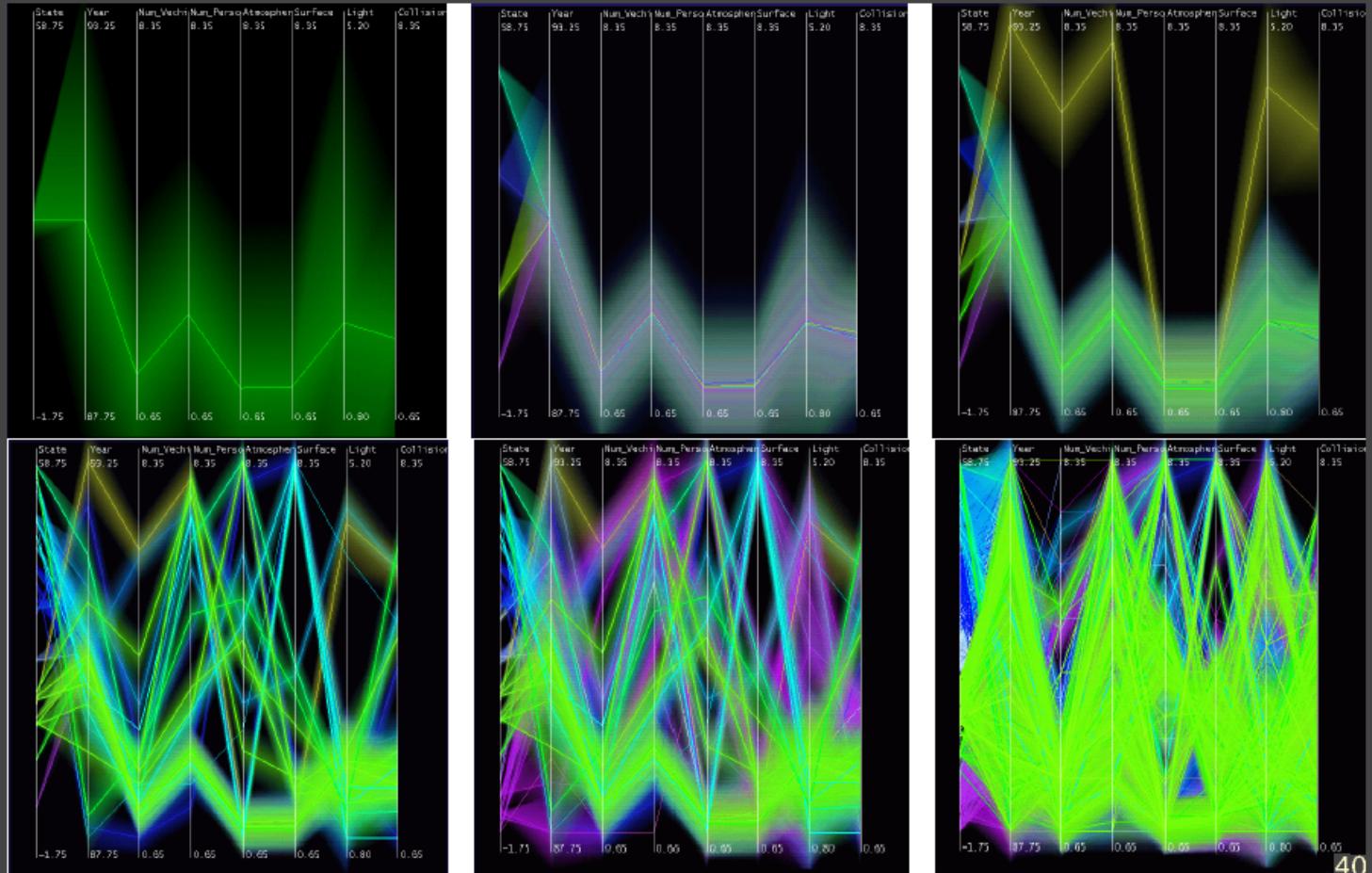
infovis

- no intrinsic order, what to do?
- indeterminate/arbitrary order
  - weakness of many techniques
  - downside: human-powered search
  - upside: powerful interaction technique
- most implementations
  - user can interactively swap axes

Automated Multidimensional Detective

- [Inselberg 99]
- machine learning approach

# Hierarchical Parallel Coords: LOD

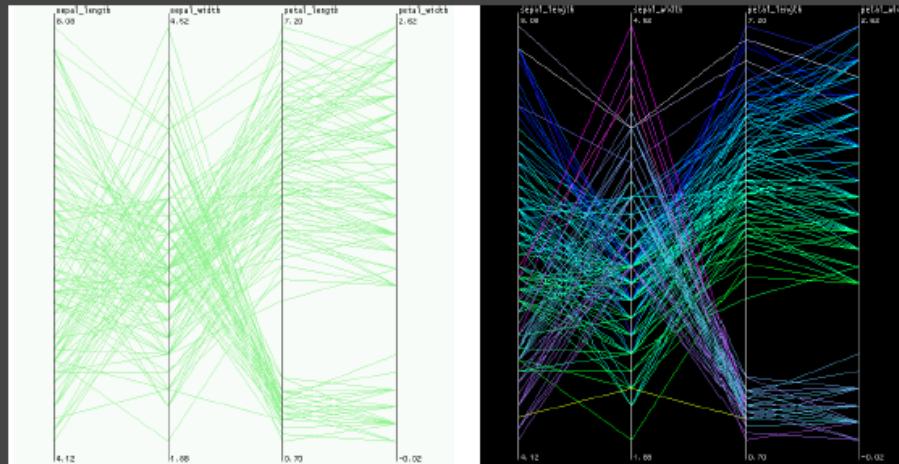


[Hierarchical Parallel Coordinates for Visualizing Large Multivariate Data Sets]

Eua, Ward, and Rundensteiner, IFFF Vis '99, davis.wpi.edu/~xmdv/docs/vis99\_HPC.pdf

# Hierarchical Clustering

proximity-based coloring



[Hierarchical Parallel Coordinates for Visualizing Large Multivariate Data Sets  
Fua, Ward, and Rundensteiner. IEEE Vis '99, davis.wpi.edu/~xmdv/docs/vis99\_HPC.pdf]

# Dimensionality Reduction

## mapping

- Q high dims
- P low dims (2 or 3)
- n points
- map  $Q \rightarrow P$
- minimize error of low-dim distances wrt high-dim dist

## distance measures

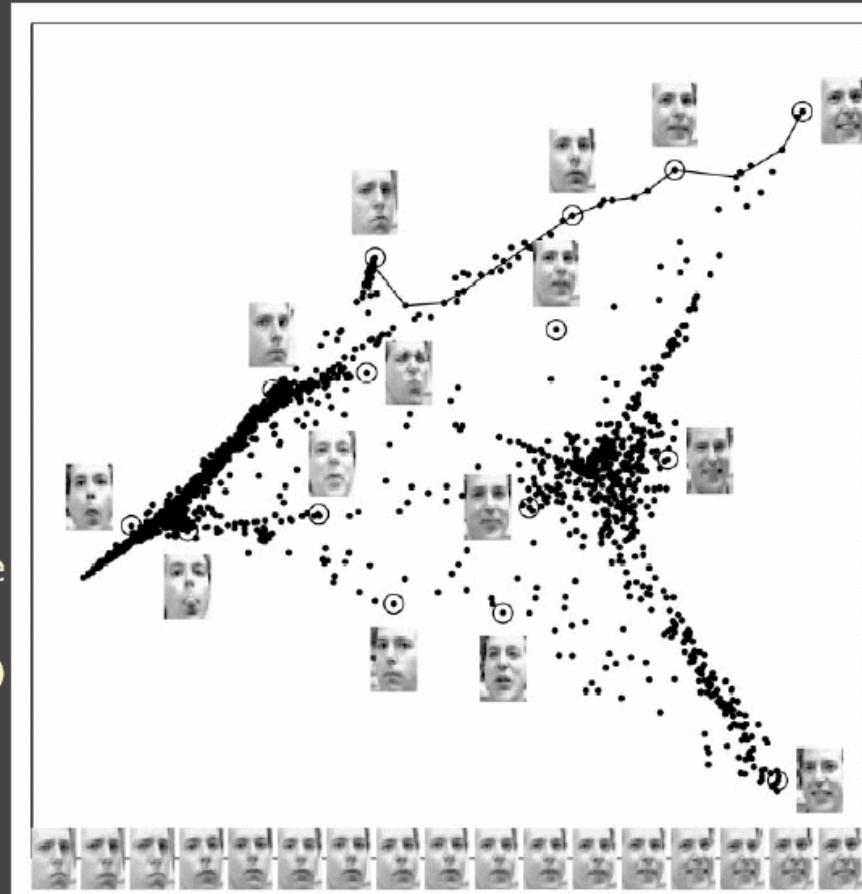
- pairwise distance matrix between points
- metric between points in space

## methods

- MDS (multidimensional scaling)
- LLE (locally linear embedding)
- IsoMap, charting
- PCA, SOM

## complexity

- naive  $O(n^3)$ , many  $O(n^2)$ ,
- best  $O(n \sqrt{n})$



[LLE, Roweis and Saul, [www.cs.toronto.edu/~roweis/papers/2323.pdf](http://www.cs.toronto.edu/~roweis/papers/2323.pdf)]

# True Dimensionality: Linear

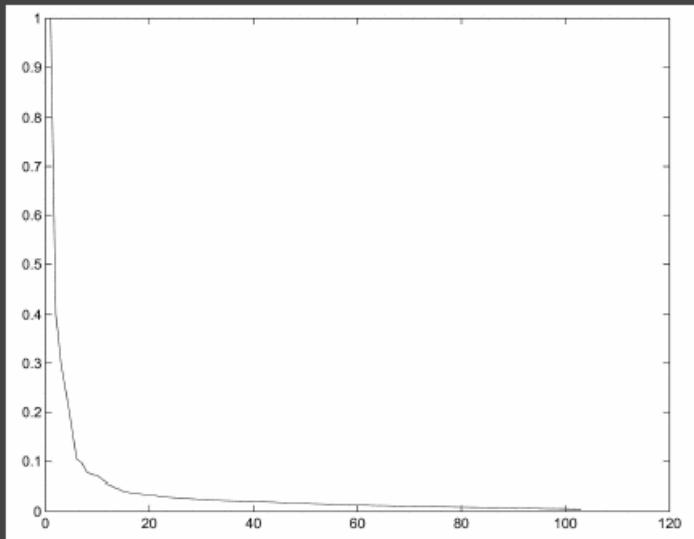
how many dimensions is enough? > 2 or 3?

- knee in error curve

example: measured materials from graphics

linear PCA: 25

- can get physically impossible intermediate points



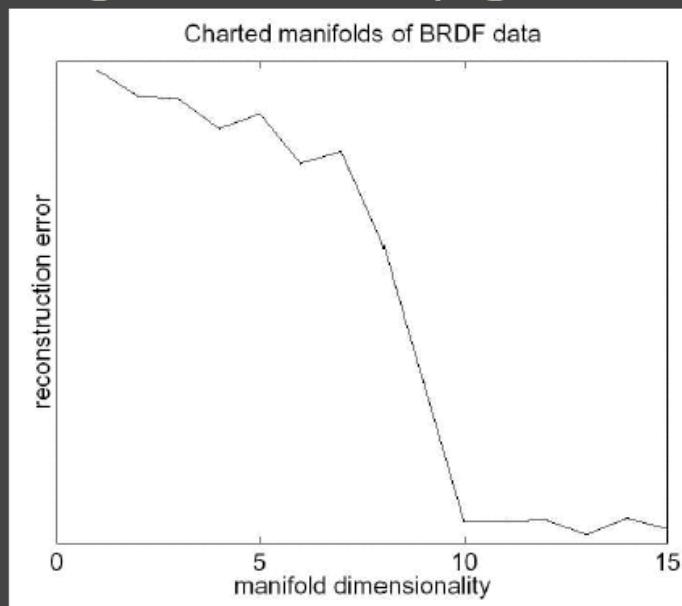
# True Dimensionality: Nonlinear

nonlinear MDS: 10–15

- all intermediate points possible

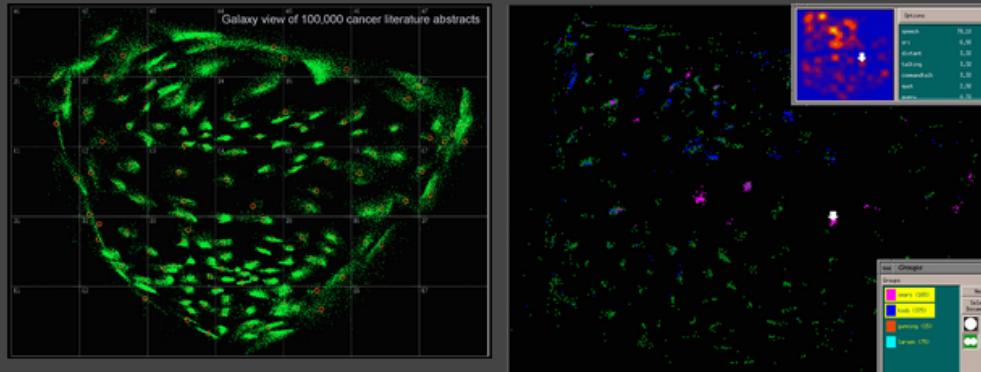
categorizable by people

- red, green, blue, specular, diffuse, glossy, metallic,
- plastic-y, roughness, rubbery, greasiness, dustiness...

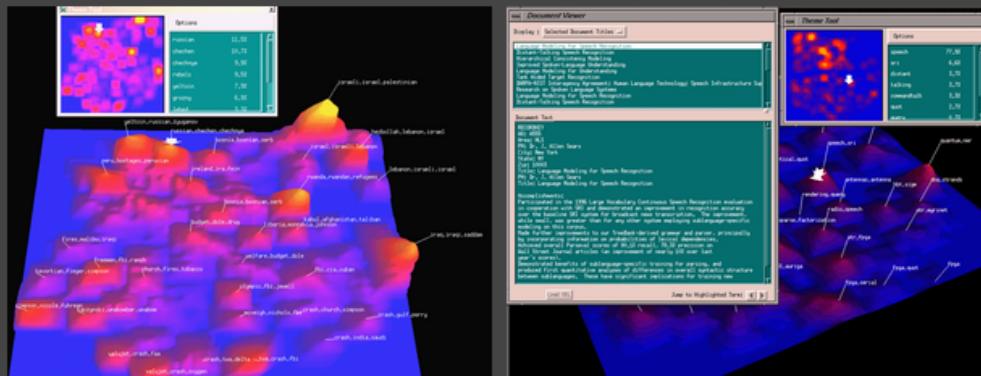


# Themescapes/Galaxies

MDS output: beyond just drawing points  
· galaxies: aggregation



· themescapes: terrain/landscapes



# Cluster Stability

display

- also terrain metaphor

underlying computation

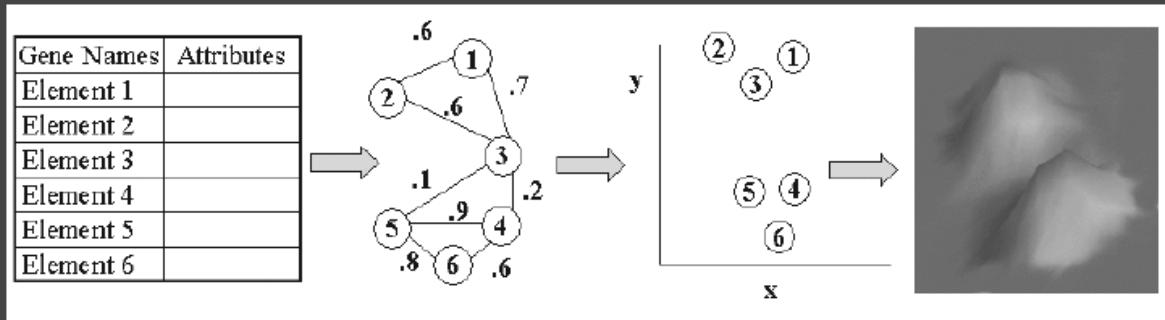
- energy minimization (springs) vs. MDS
- weighted edges

do same clusters form with different random start points?

"ordination"

- spatial layout of graph nodes

# Approach



normalize within each column

similarity metric

- discussion: Pearson's correlation coefficient

threshold value for marking as similar

- discussion: finding critical value

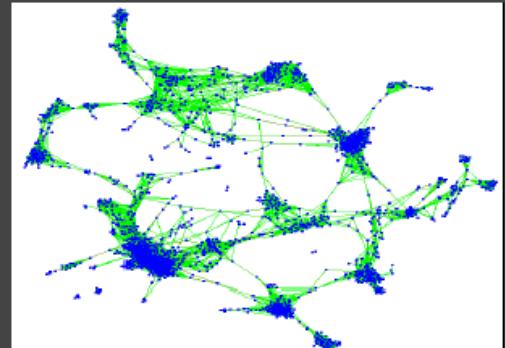
# Graph Layout

## criteria

- distance in layout matching graph-theoretic distance
  - vertices one hop away close
  - vertices many hops away far
- insensitive to random starting positions
  - major problem with previous work!
- tractable computation

## force-directed placement

- discussion: energy minimization
- others: gradient descent, etc
- discussion: termination criteria

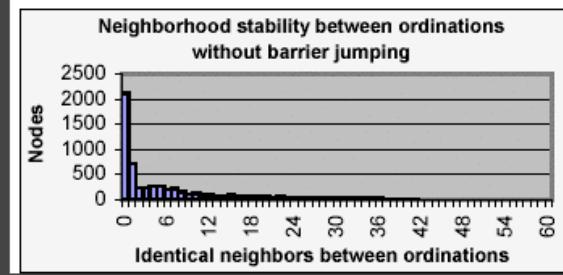
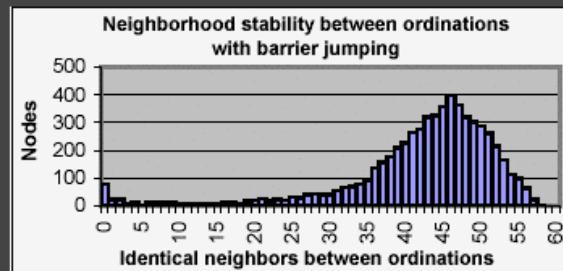
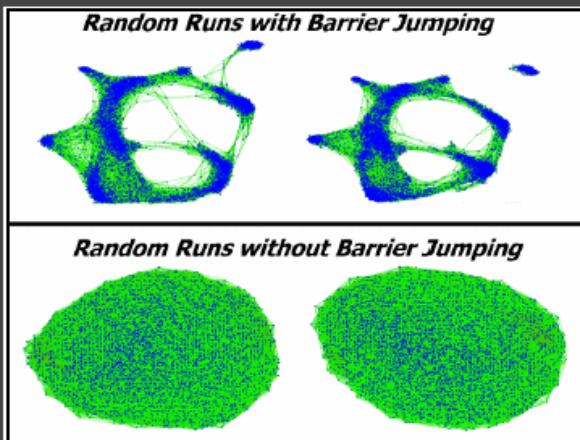
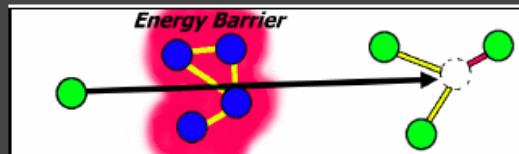


# Barrier Jumping

same idea as simulated annealing

- but compute directly
- just ignore repulsion for fraction of vertices

solves start position sensitivity problem



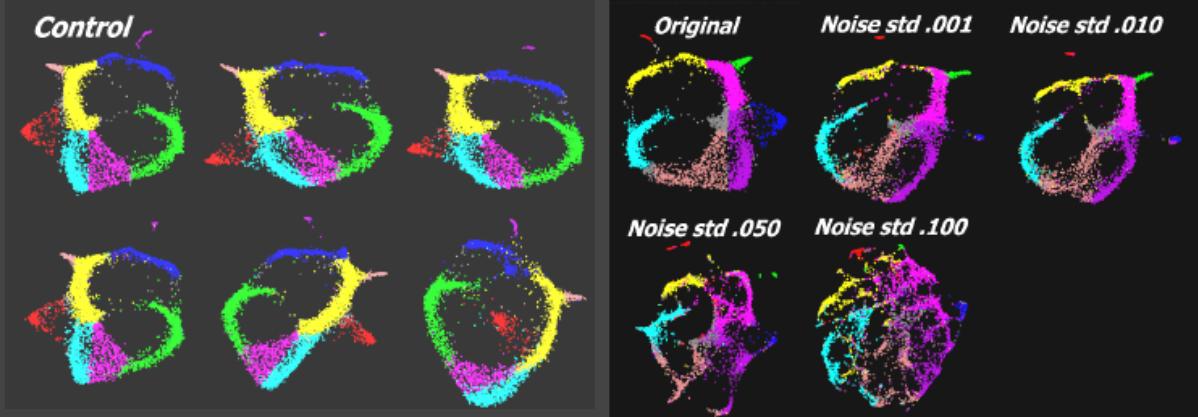
# Results

efficiency

- naive approach:  $O(V^2)$
- approximate density field:  $O(V)$

good stability

- rotation/reflection can occur



# Critique

real data

- suggest check against subsequent publication!

give criteria, then discuss why solution fits

visual + numerical results

- convincing images plus benchmark graphs

detailed discussion of alternatives at each stage

specific prescriptive advice in conclusion

# More Reading

Parallel Coordinates: A Tool for Visualizing Multi-Dimensional Geometry.  
Alfred Inselberg and Bernard Dimsdale, IEEE Visualization '90, 1990.

On Some Graphical Representations of Multivariate Data.  
M. Bolorfoush and E. J. Wegman. Proc. of Interface '88, Amer. Stat. Assoc.

Hierarchical Parallel Coordinates for Visualizing Large Multivariate Data Sets  
Fua, Ward, and Rundensteiner. IEEE Visualization '99,  
[http://davis.wpi.edu/~xmdv/docs/vis99\\_HPC.pdf](http://davis.wpi.edu/~xmdv/docs/vis99_HPC.pdf)

Nonlinear dimensionality reduction by locally linear embedding.  
Sam Roweis & Lawrence Saul. Science v.290 no.5500, Dec.22, 2000. pp.2323--2326.  
<http://www.cs.toronto.edu/~roweis/papers/2323.pdf>

Fast Multidimensional Scaling through Sampling, Springs and Interpolation  
Alistair Morrison, Greg Ross, Matthew Chalmers  
Information Visualization 2(1) March 2003, pp. 68–77.  
<http://www.dcs.gla.ac.uk/~matthew/papers/JInfoVis.pdf>

Cluster Stability and the Use of Noise in Interpretation of Clustering  
George S. Davidson, Brian N. Wylie, Kevin W. Boyack, Proc InfoVis 2001.  
<http://citeseer.nj.nec.com/davidson01cluster.html>

# Mini-Course Outline

Perception

Frameworks

Color

Space/Order

Depth/Occlusion

High Dimensionality

Interaction

Navigation/Zooming

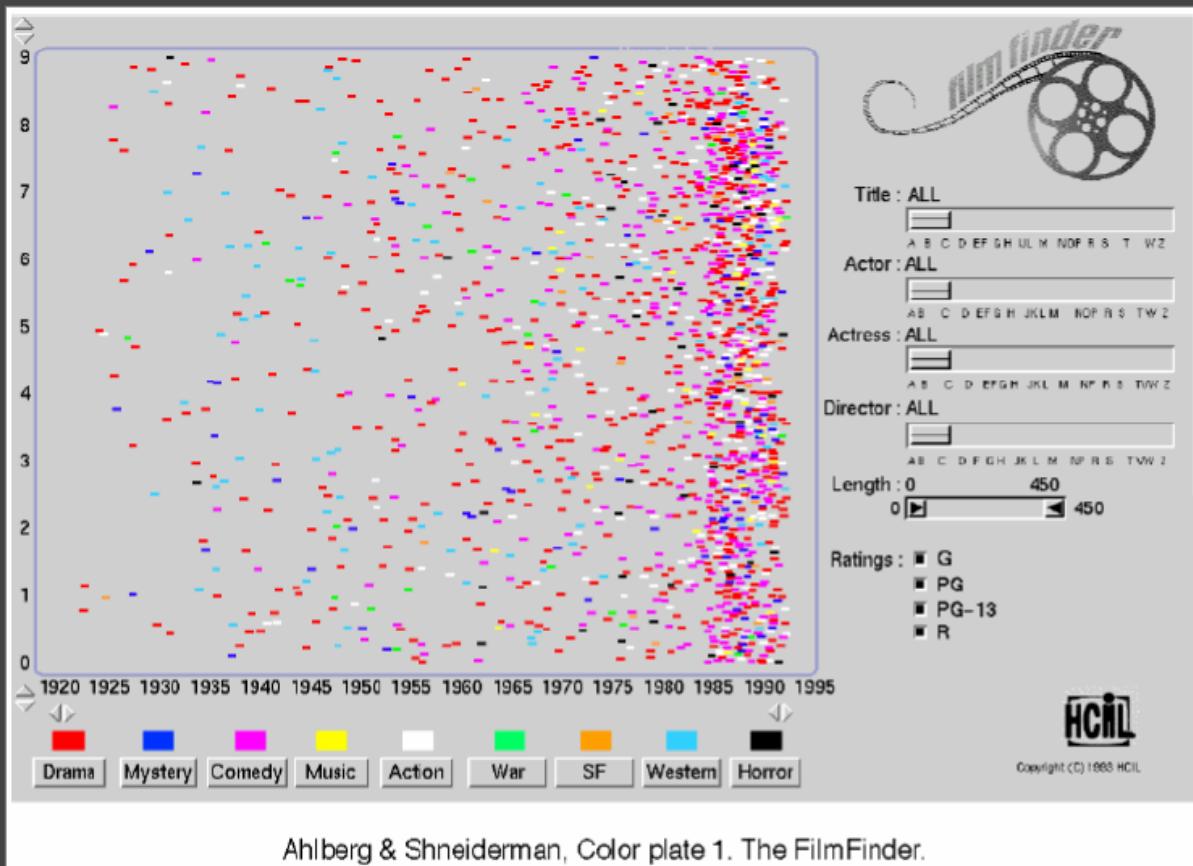
Focus+Context

Graphs/Trees

Scalability

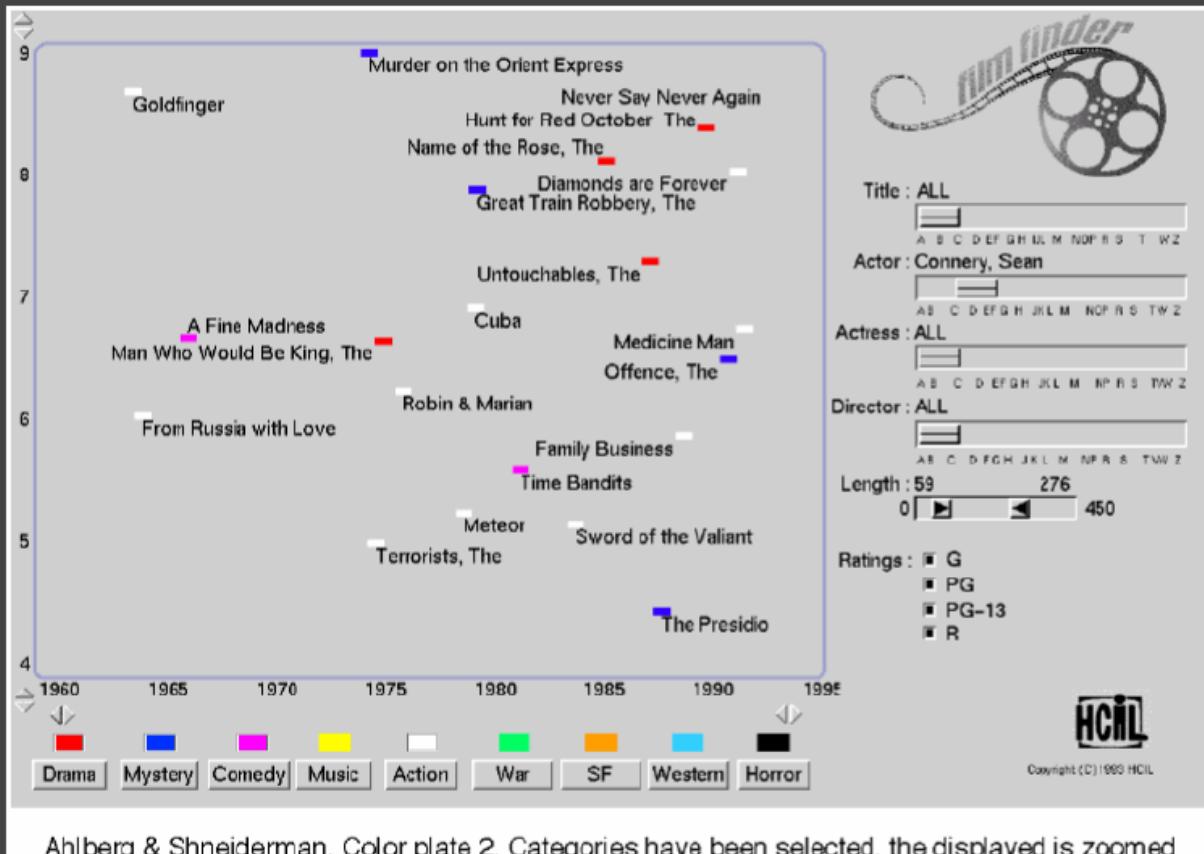
Task-Centered Design

# Dynamic Queries



[Visual Information Seeking: Tight Coupling of Dynamic Query Filters with Starfield Displays. Ahlberg and Shneiderman, Proc SIGCHI '94. citeseer.ist.psu.edu/ahlberg94visual.html]

# Dynamic Queries

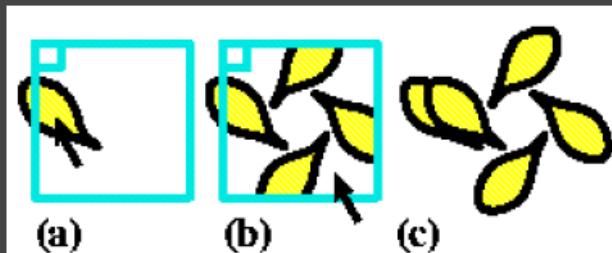


Ahlberg & Shneiderman, Color plate 2. Categories have been selected, the displayed is zoomed  
[Visual Information Seeking: Tight Coupling of Dynamic Query Filters  
with Starfield Displays. Ahlberg and Shneiderman, Proc SIGCHI  
'94, citeseer.ist.psu.edu/ahlberg94visual.html]

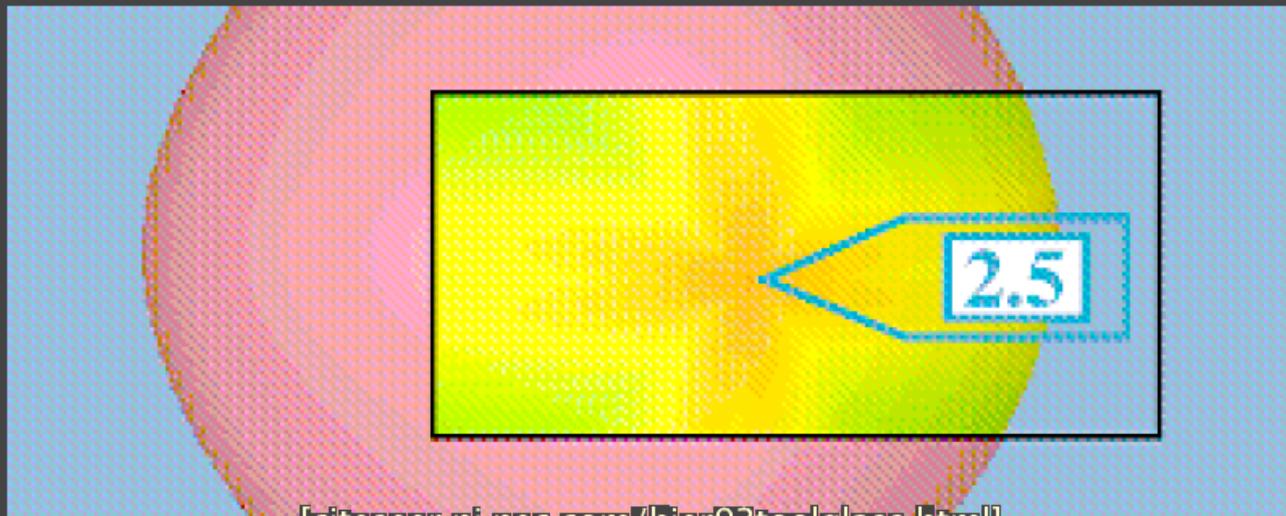
# Toolglass and Magic Lens

- see-through
- two-handed

symmetry glass



curvature lens



# Linked Views

linked highlighting/brushing/

- extremely powerful technique
- Brushing Scatterplots, Becker and Cleveland,  
Technometrics 1987 vol 29, pp 127-142

coordinated views

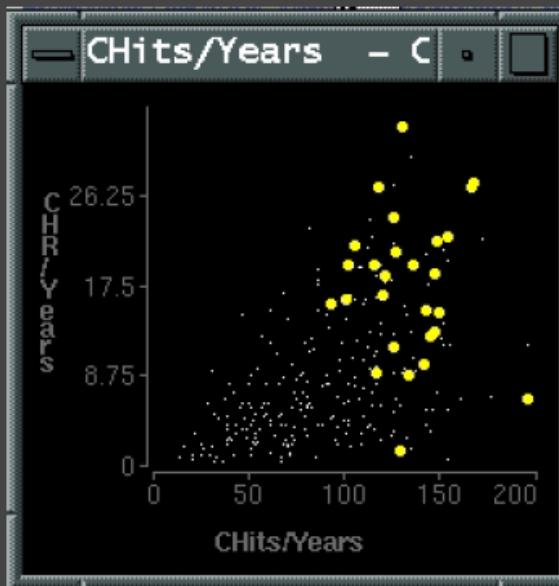
- linked navigation
- CMV: International Conference on Coordinated & Multiple Views in Exploratory Visualization

example: Exploratory Data Visualizer

- Visual Exploration of Large Structured Databases,  
Graham J. Wills, in New Techniques and Trends in  
Statistics, pp 237-246, IOS Press 1995.

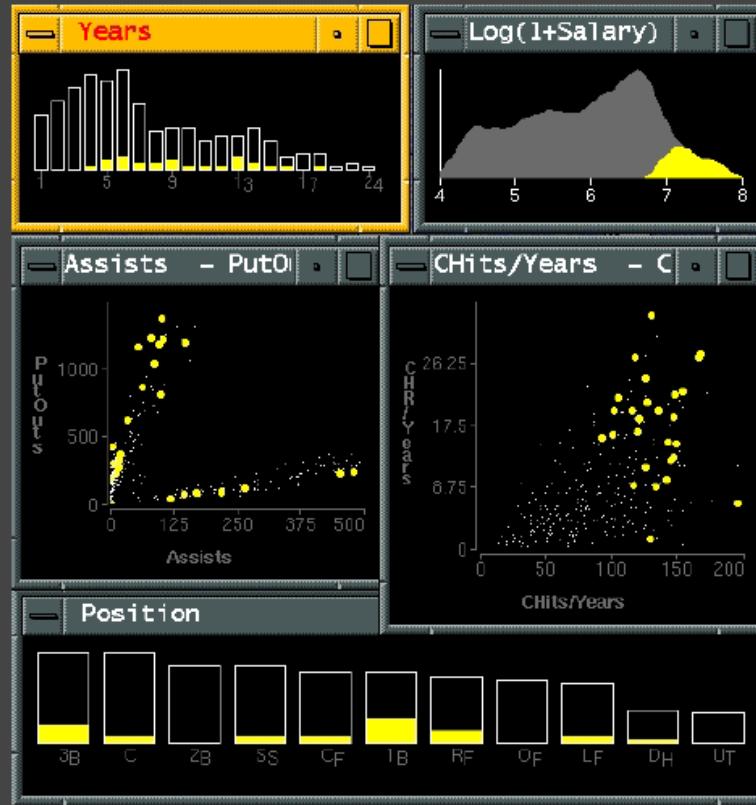
# Highlighting (Focusing)

Focus user attention on a subset of the data within one graph (from Wills 95)

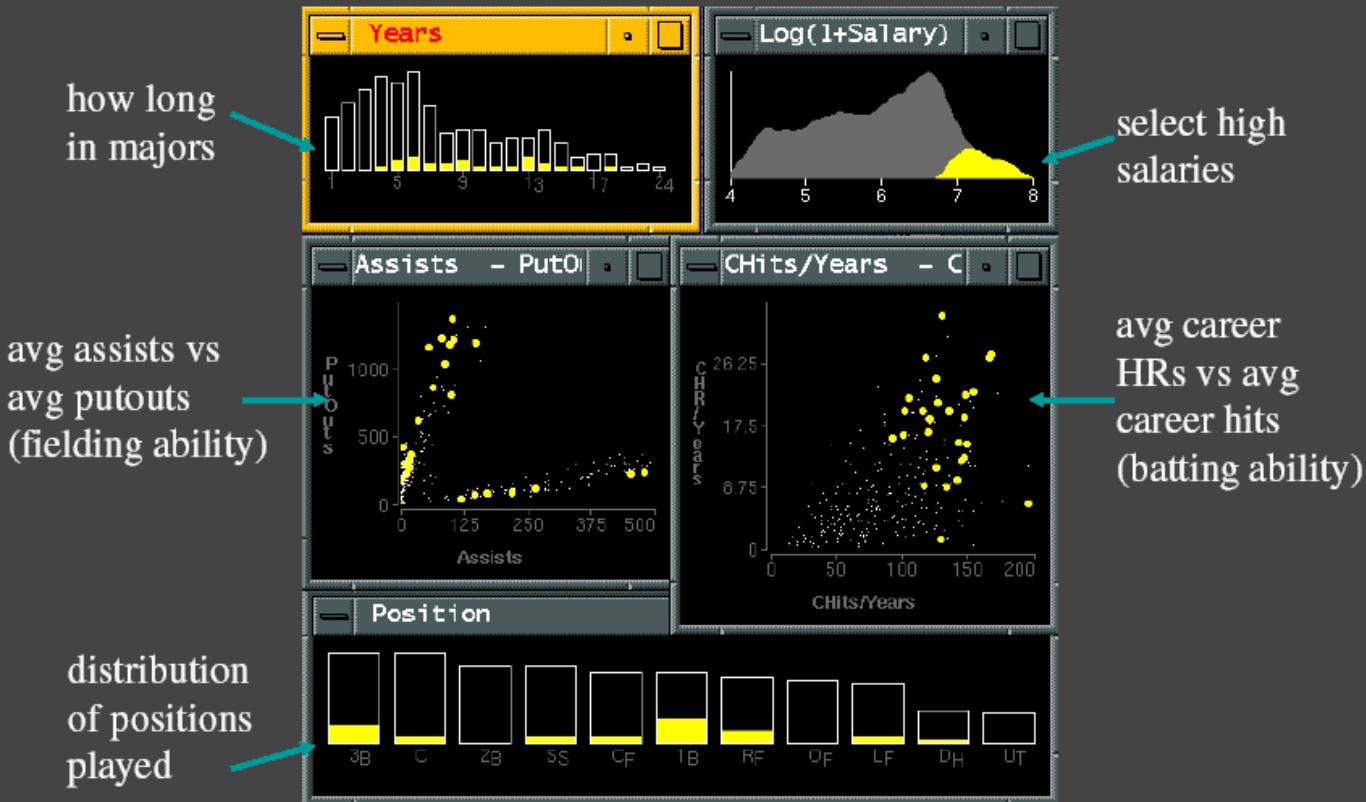


# Link different types of graphs: Scatterplots and histograms and bars

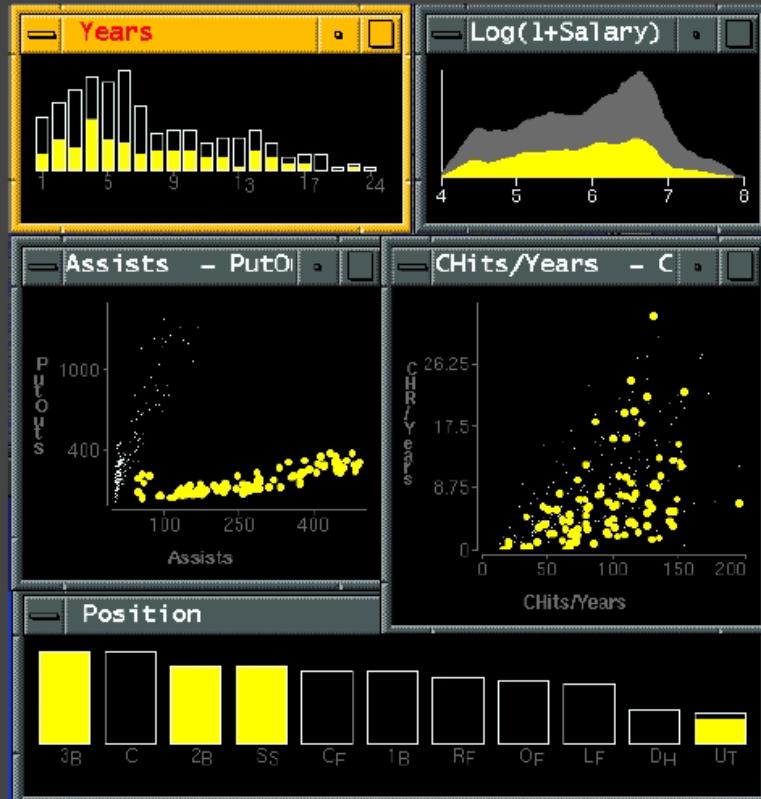
(from Wills 95)



# Baseball data: Scatterplots and histograms and bars (from Wills 95)



# Linking types of assist behavior to position played (from Wills 95)



# More Reading

Visual information seeking: Tight coupling of dynamic query filters with starfield displays. Chris Ahlberg and Ben Shneiderman, Proc SIGCHI '94, pages 313–317. <http://citeseer.ist.psu.edu/ahlberg94visual.html>

Toolglass and magic lenses: the see-through interface  
Eric A. Bier, Maureen C. Stone, Ken Pier, William Buxton, and Tony D. DeRose, Proc. SIGGRAPH'93, pp. 73–76. <http://citeseer.nj.nec.com/bier93toolglass.html>

Brushing Scatterplots, Becker and Cleveland  
Technometrics, vol 29, pp 127–142, 1987  
Reprinted in Dynamic Graphics for Data Analysis, edited by W. S. Cleveland and M. E. McGill, Chapman and Hall, New York, (1988)

Visual Exploration of Large Structured Databases, Graham J. Wills, in New Techniques and Trends in Statistics, pp 237–246, IOS Press 1995.