

# Lecture 2, InfoVis MiniCourse

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

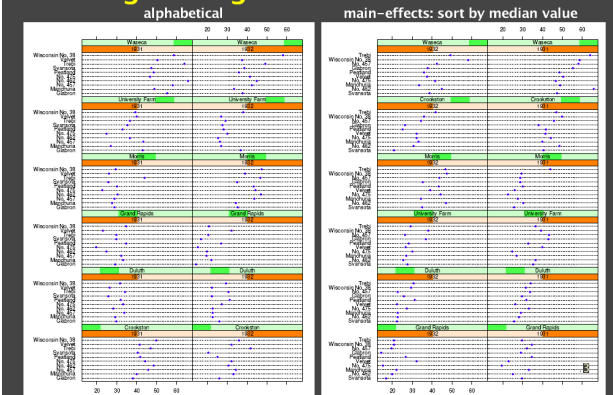
LaBRI, University of Bordeaux  
16 June 2004

Tamara Munzner

### Mini-Course Outline

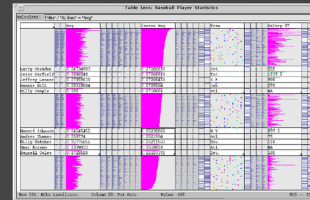
- Perception
- Frameworks
- Color
- Space/Order
- Depth/Occlusion
- High Dimensionality
- Interaction
- Navigation/Zooming
- Focus+Context
- Graphs/Trees
- Scalability
- Task-Centered Design

### Finding the Right Order: Trellis



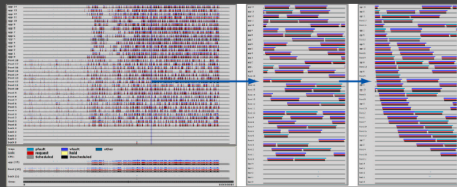
### Finding the Right Order: Table Lens

- select column to sort
- user explores



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

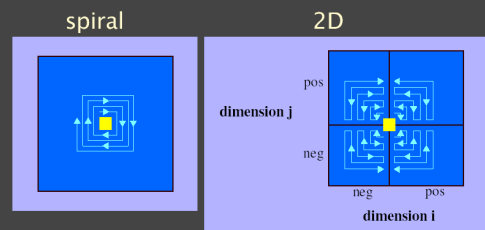
### Finding the Right Order: Rivet



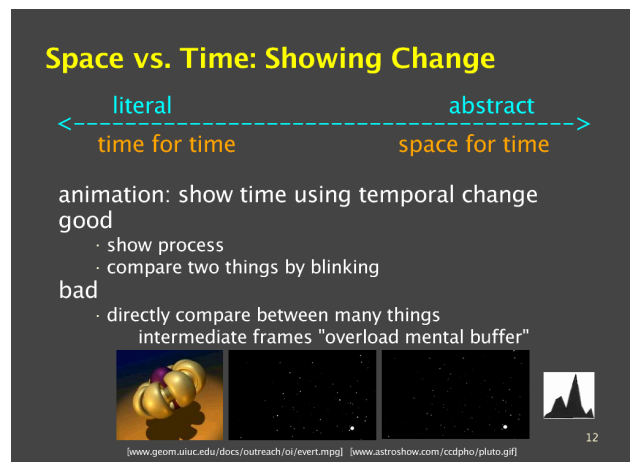
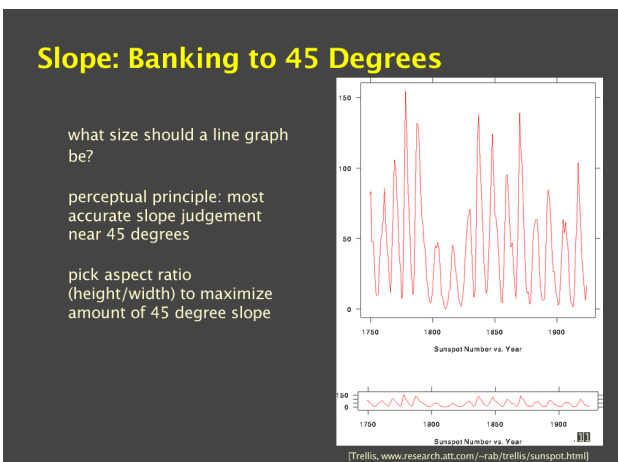
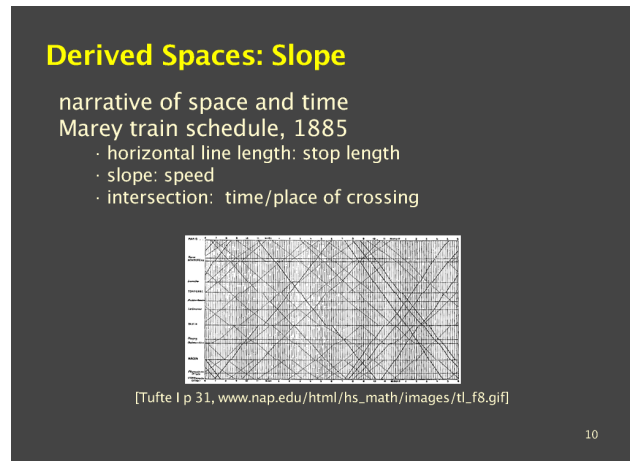
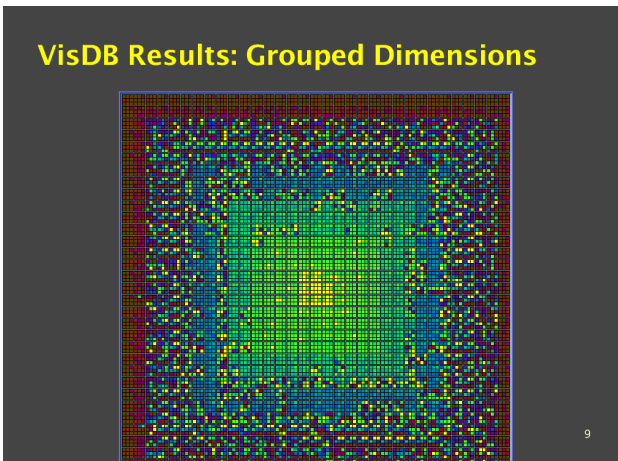
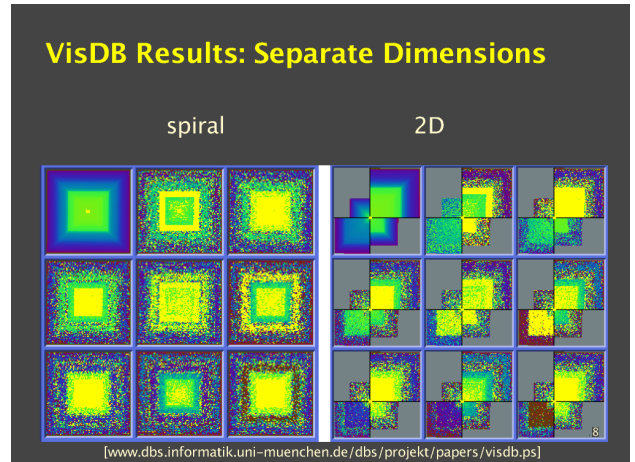
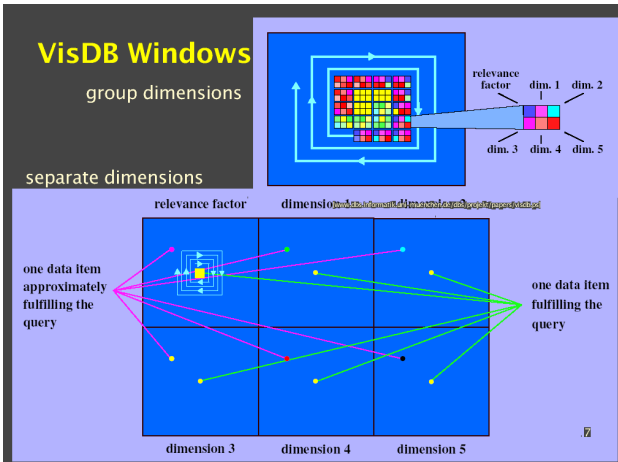
[Bosch, Performance Analysis and Visualization of Parallel Systems Using SimOS and Rivet: A Case Study, HPCA6, 2000, [graphics.stanford.edu/papers/rivet\\_argus/](http://graphics.stanford.edu/papers/rivet_argus/)]

### Finding the Right Order: VisDB

- each pixel represents database entry
- color by query relevance
- what order to sort?



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

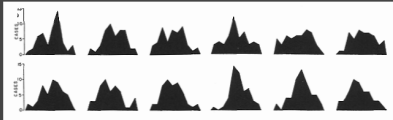


## Space vs. Time: Showing Change



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]

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## 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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## Layering And Separation



## Visual Clutter

subtler background than foreground

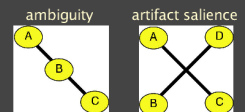


[Tufte, Envisioning Information, Chap 3]

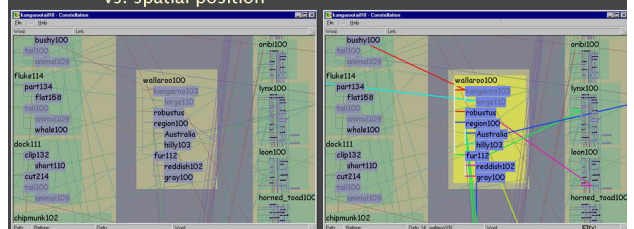
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## Visual Layering For Graphs

edge crossing problem  
· false attachments



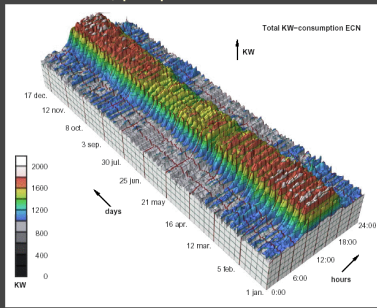
layers to avoid perception  
· vs. spatial position



[Munzner et al, Constellation, [graphics.stanford.edu/papers/const](http://graphics.stanford.edu/papers/const)]

### 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

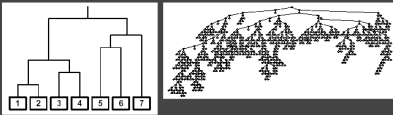
- predictive mathematical models
  - details lost, multiscale not addressed
- scale-space approaches (wavelet, fourier, fractal)
  - hard to interpret, known scales lost
- 3D mountain: x hours, y value, z days
  - occlusion hides, perspective interferes

### 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

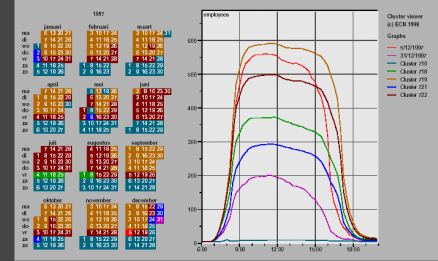
- shows structure of hierarchy
- does not solve pattern finding problem!



### Link Clusters and Calendar

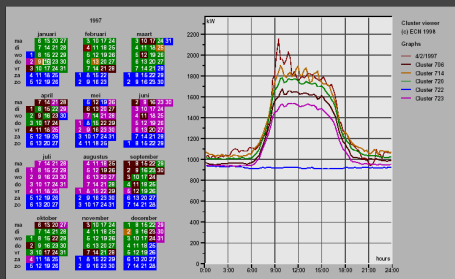
2D linked clusters–calendars shows patterns

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



[van Wijk and van Selow, Cluster and Calendar based Visualization of Time Series Data, InfoVis99, Figure 4, citeseer.nj.nec.com/vanwijk99cluster.html]

### Power Consumption



[van Wijk and van Selow, Cluster and Calendar 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



	Stretch Orthogonal	Non-Linear Orthogonal	Non-Linear Radial	Step Orthogonal
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

### Results

single, multiple foci

[pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/cqa97.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

### 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/cqa97.pdf]

### Results

randomly positioned nodes instead of grid

- closer to real dataset

[pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/cqa97.pdf]

### EdgeLens

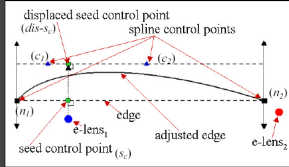
interactive control over edge occlusion

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]

## EdgeLens Final Algorithm

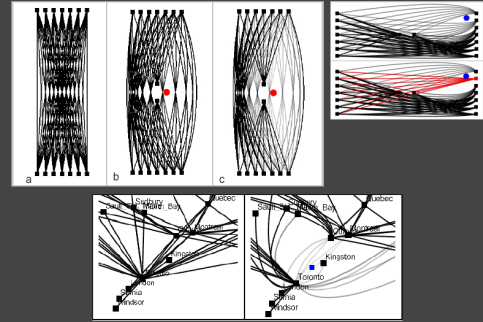


- decide which edges affected
- calculate displacements
- calculate spline control points
- draw curves

[pages.cpsc.ucalgary.ca/~sheelagh/personal/pubs/2003/wong-carp-infovis03-submit.31f]

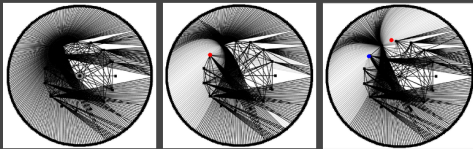
## EdgeLens Techniques

transparency, color



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

## EdgeLens Results



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

critique

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

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## Cheops

compact

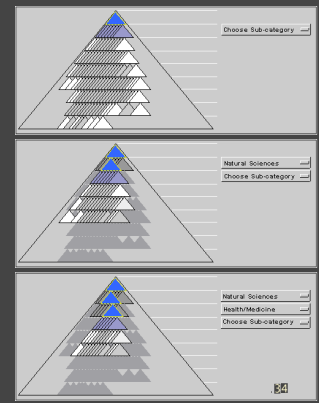
show paths through tree

extreme occlusion  
deliberately

browsing/exploration, not  
topological analysis



[Cheops, Beaudoin/Parent/Vroomen, www.istop.com/~maparent/paper.html]



## 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

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## 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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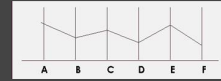
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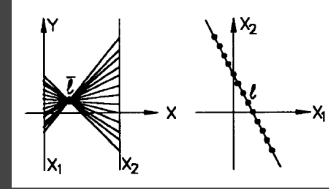
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## Parallel Coordinates

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



point-line duality



[Parallel Coordinates: A Tool for Visualizing Multi-Dimensional Geometry.

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## Parallel Coords: Axis Ordering

geometric interpretations

- hyperplane, hypersphere: points have intrinsic order

infovis

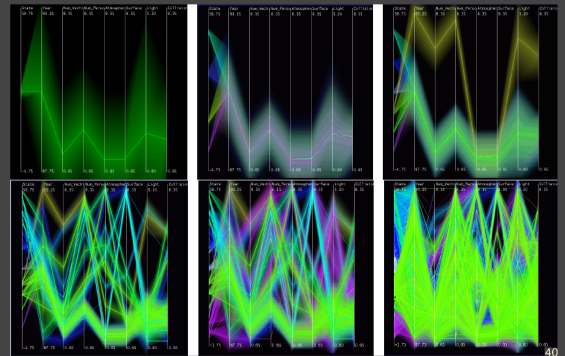
- no intrinsic order, what to do?
- indeterminate/arbitrary order
  - weakness of many techniques
  - downside: human-powered search
  - upside: powerful interaction technique
- most implementations
- user can interactively swap axes

Automated Multidimensional Detective

- [Inselberg 99]
- machine learning approach

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## Hierarchical Parallel Coords: LOD

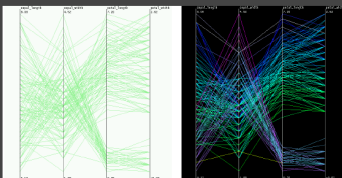


[Hierarchical Parallel Coordinates for Visualizing Large Multivariate Data Sets  
Fua, Ward, and Rundensteiner. IEEE Vis '99, davis.wpi.edu/~xmdv/docs/vis99\_HPC.pdf]

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## 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]

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## 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

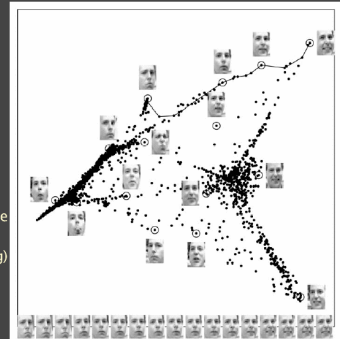
- pairwise distance matrix between points
- metric between points in space

methods

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

complexity

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



[LLE, Roweis and Saul, www.cs.toronto.edu/~roweis/papers/lle23.pdf]

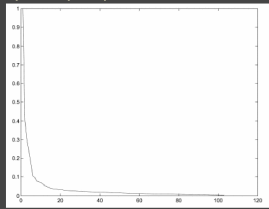
## True Dimensionality: Linear

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



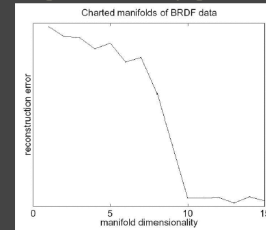
[A Data-Driven Reflectance Model, SIGGRAPH 2003, W Matusik, H. Pfister M. Brand and L. McMillan, graphics.lcs.mit.edu/~wojciech/pubs/sig2003.pdf] 43

## 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...

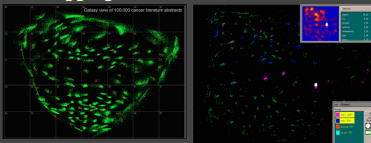


[A Data-Driven Reflectance Model, SIGGRAPH 2003, W Matusik, H. Pfister 44

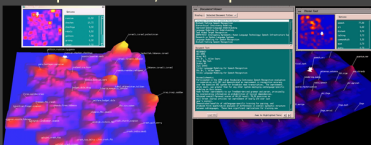
## Themescapes/Galaxies

MDS output: beyond just drawing points

- galaxies: aggregation



- themescapes: terrain/landscapes



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## Cluster Stability

display

- also terrain metaphor

underlying computation

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

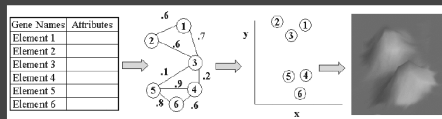
do same clusters form with different random start points?

"ordination"

- spatial layout of graph nodes

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## Approach



normalize within each column

similarity metric

- discussion: Pearson's correlation coefficient

threshold value for marking as similar

- discussion: finding critical value

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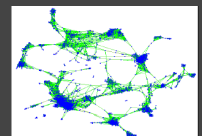
## Graph Layout

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

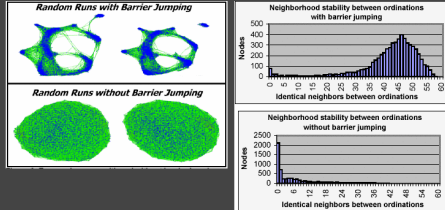
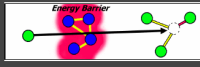
- discussion: energy minimization
- others: gradient descent, etc
- discussion: termination criteria



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## Barrier Jumping

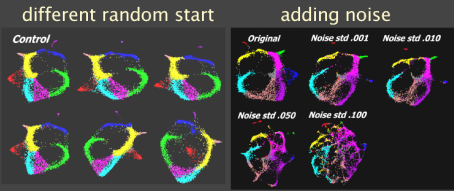
- same idea as simulated annealing
  - but compute directly
  - just ignore repulsion for fraction of vertices
- solves start position sensitivity problem



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## Results

- efficiency
  - naive approach:  $O(V^2)$
  - approximate density field:  $O(V)$
- good stability
  - rotation/reflection can occur



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## 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

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## 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/InfoVis.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>

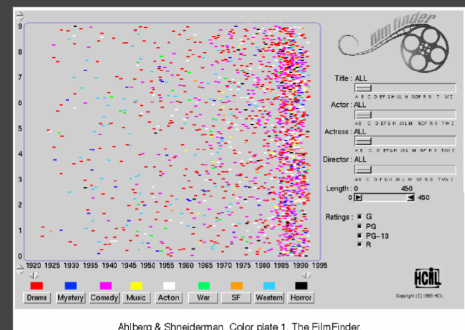
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## Dynamic Queries

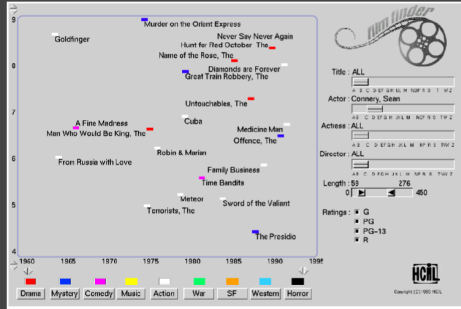


Ahlberg & Shneiderman, Color plate 1. The FilmFinder. [Visual Information Seeking: Tight Coupling of Dynamic Query Filters with Starfield Displays. Ahlberg and Shneiderman, Proc SIGCHI '94. [citeseer.ist.psu.edu/ahlberg94visual.html](http://citeseer.ist.psu.edu/ahlberg94visual.html)]

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## 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]

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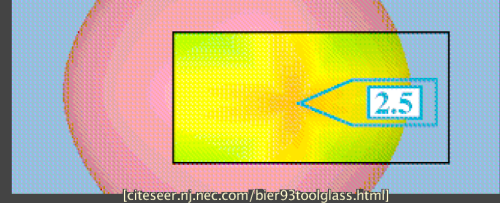
## Toolglass and Magic Lens

- see-through
- two-handed

symmetry glass



curvature lens



[citeseer.ist.psu.edu/bier93toolglass.html]

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## 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

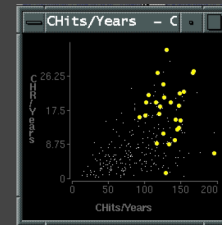
example: Exploratory Data Visualizer

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

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## Highlighting (Focusing)

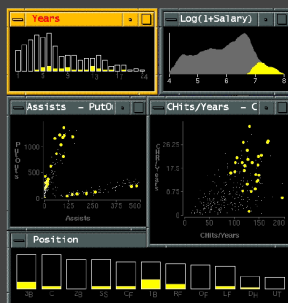
Focus user attention on a subset of the data within one graph (from Wills 95)



[www.sims.berkeley.edu/courses/is247/s02/lectures/Lecture3.ppt]

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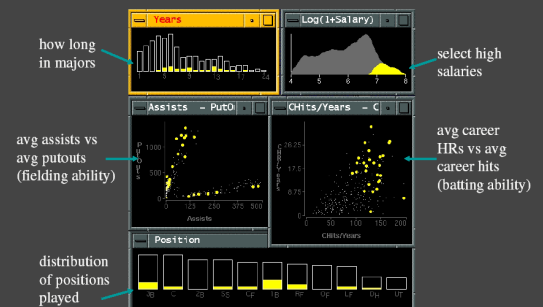
## Link different types of graphs: Scatterplots and histograms and bars (from Wills 95)



[www.sims.berkeley.edu/courses/is247/s02/lectures/Lecture3.ppt]

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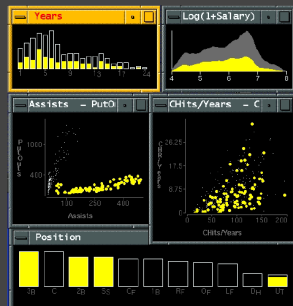
## Baseball data: Scatterplots and histograms and bars (from Wills 95)



[www.sims.berkeley.edu/courses/is247/s02/lectures/Lecture3.ppt]

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



[www.sims.berkeley.edu/courses/is247/s02/lectures/Lecture3.ppt]

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## 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.