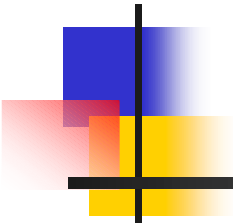
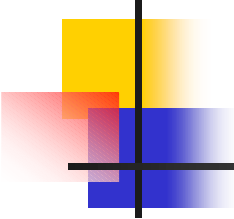


Visualization Frameworks to Detect the Presence of Unusual Structures in High-Dimensional Data



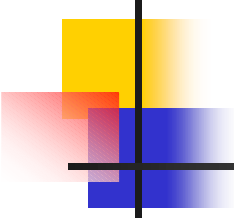
Jeff Solka
NSWDD and George Mason University
February 19, 2004

DIMACS Working Group on
Adverse Event/Disease Reporting,
Surveillance, and Analysis II



Acknowledgements

- **Ed Wegman**
- **Dan Carr**
- **Helpful discussions, pictures, software**



Benefits of Visualization

- **Software debugging**
- **Algorithmic understanding**
- **New applications of algorithms**
- **Data mining and exploration**
- **Model mining and exploration**

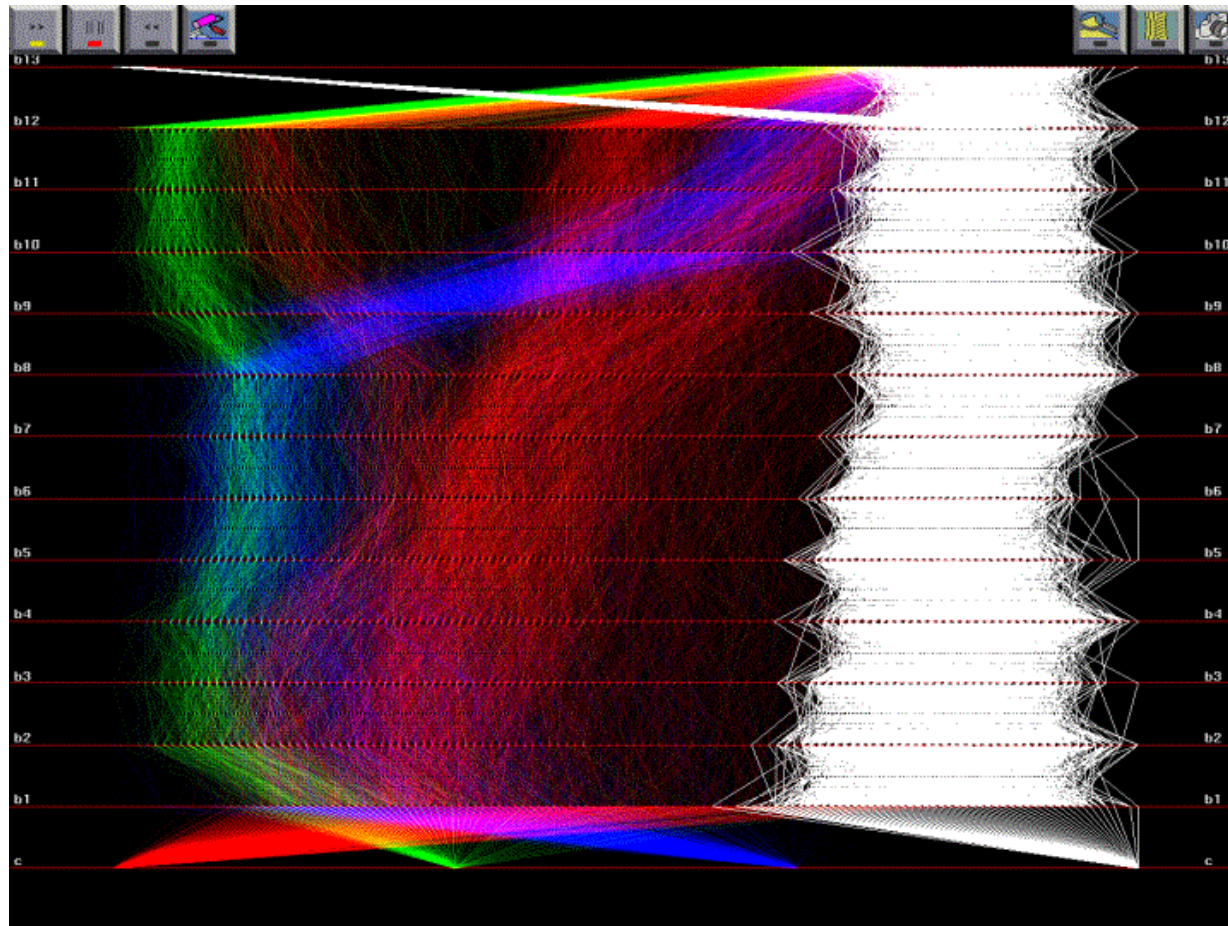
Biosurveillance Needs and Visualization Strategies

- **High-dimensional Data Visualization**
- **Outlier/Anomaly Detection**
- **Cluster Detection**
- **Methods of Hypothesis Generation and Testing**
- **Spatial Statistics Smoothing Methods**
- **Disparate Data Fusion Methods**
- **Meta Data Access and Organization**
- **Modeling**
- **Collaborative Environments**
- **Denoising**

- **Parallel Coordinates**
- **Brushing on Linked Plots**
- **Grand Tour**
- **Dotplots**
- **Distributional Summaries**
- **Micromaps**
- **Ordered Data Images**
- **Ordered Interpoint Distance Images**
- **Integrated Visualization Environments**
 - **Overview and drill-down**
 - **Data analysis management**

SALAD Data

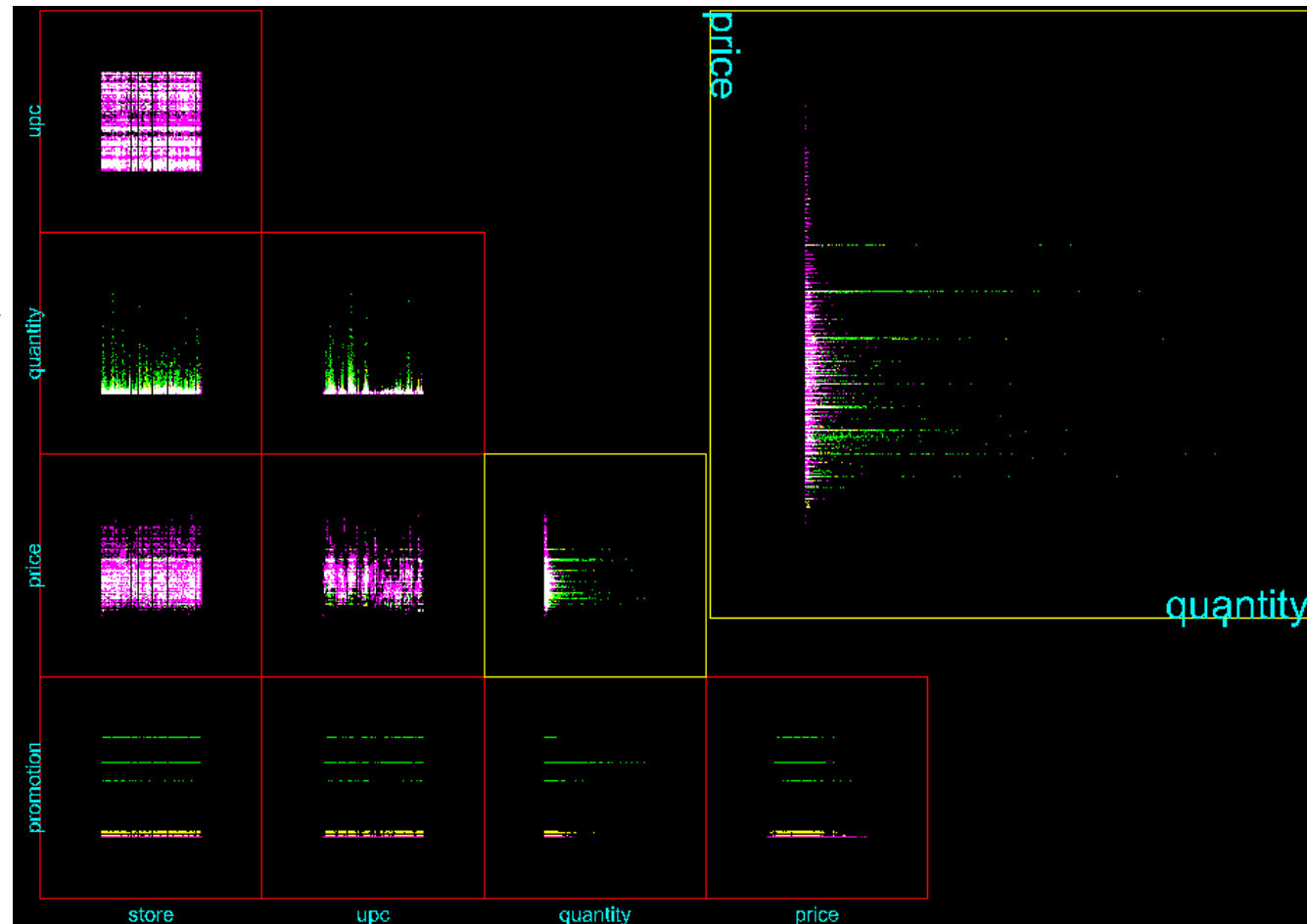
*Parallel
coordinates for
cluster
identification and
discriminant
analysis
facilitation*



Adverse Event/Disease Reporting,
Surveillance, and Analysis II

Scanner Data for Breakfast Cereals (April 1999)

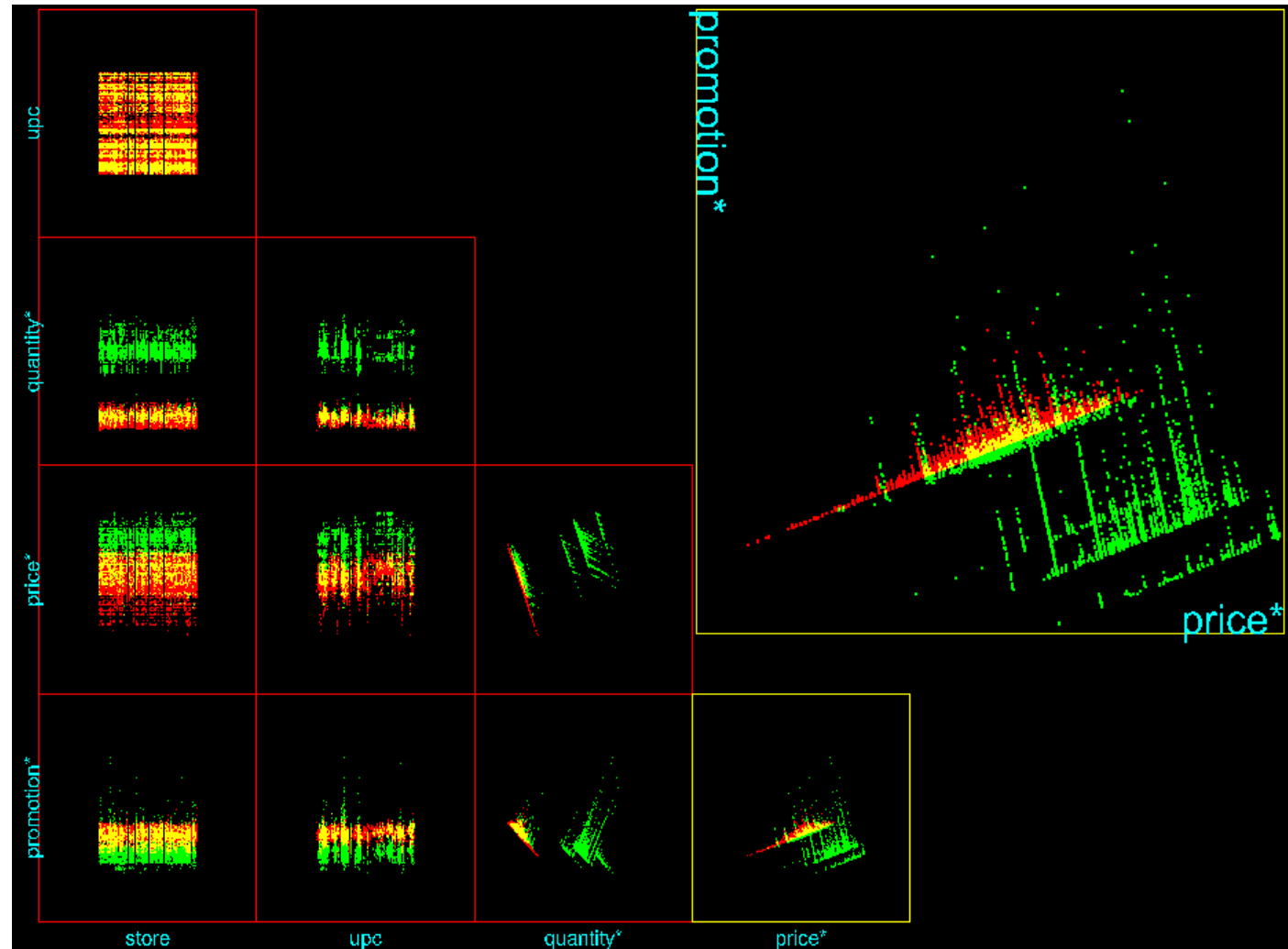
*Reprinted with
permission from
"Visualizing Cereal
World,"
Ed Wegman,
George Mason
University
(ewegman@gmu.
edu)*



Promotion has huge impact on sales volume

Scanner Data for Breakfast Cereals (April 1999)

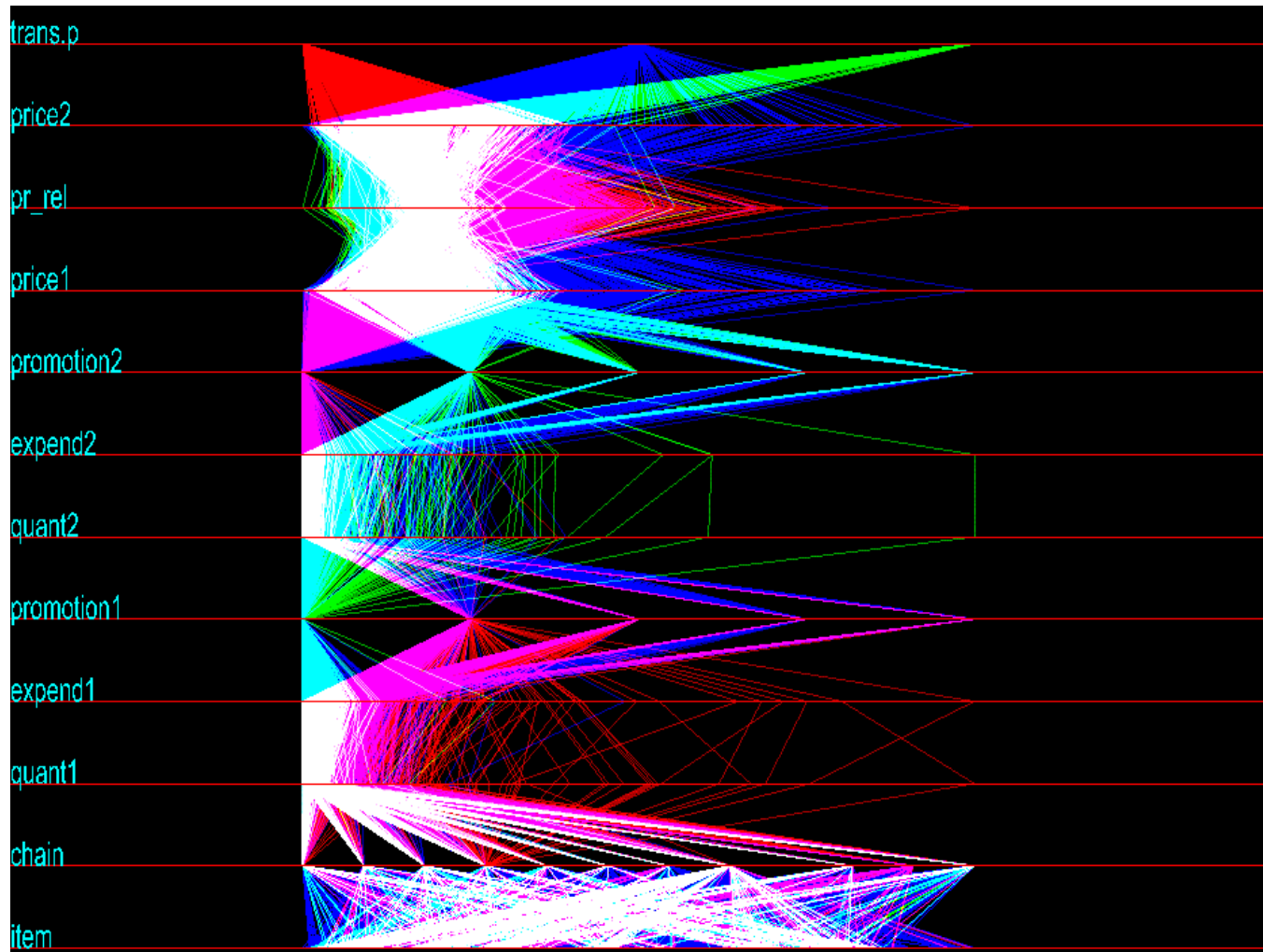
*Reprinted with
permission from
"Visualizing Cereal
World,"
Ed Wegman,
George Mason
University
(ewegman@gmu.
edu)*



Aggressive promotion pays

Scanner Data for Breakfast Cereals

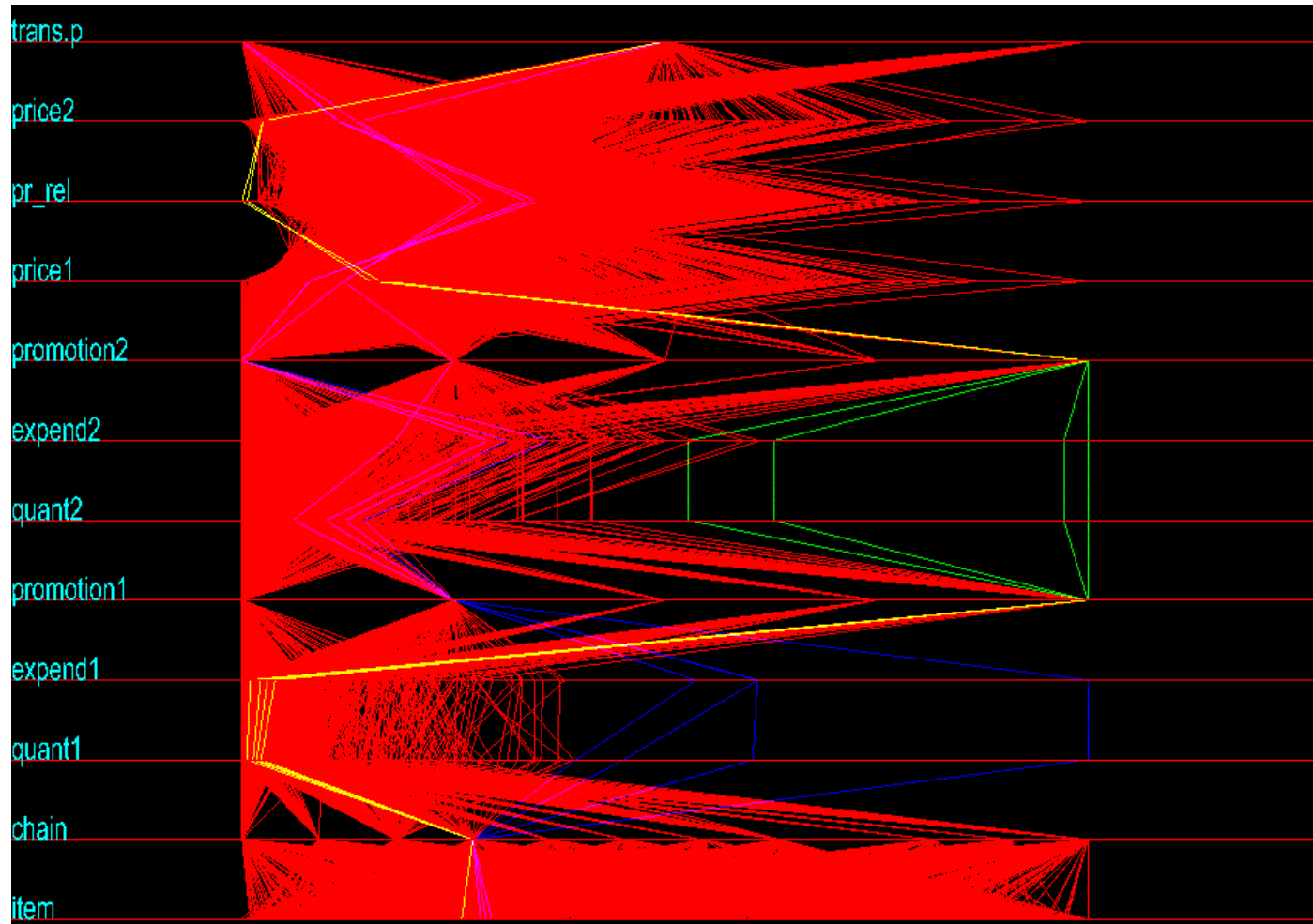
*Reprinted with
permission from
"Visualizing Cereal
World,"
Ed Wegman,
George Mason
University
(ewegman@gmu.
edu)*



Promotion of Cereals Creates Outliers

Scanner Data for Breakfast Cereals

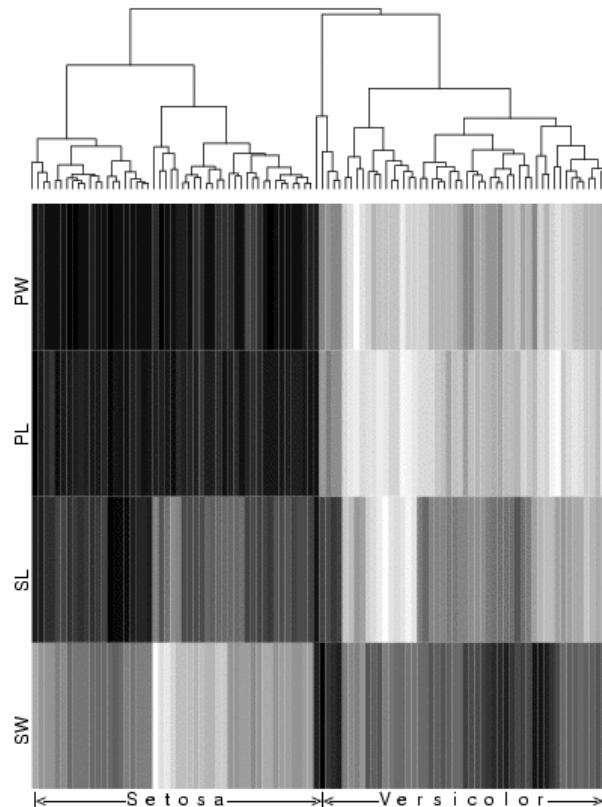
*Reprinted with
permission from
"Visualizing Cereal
World,"
Ed Wegman,
George Mason
University
(ewegman@gmu.
edu)*



Outliers belong to same chain

Data Color Histogram or Data Image

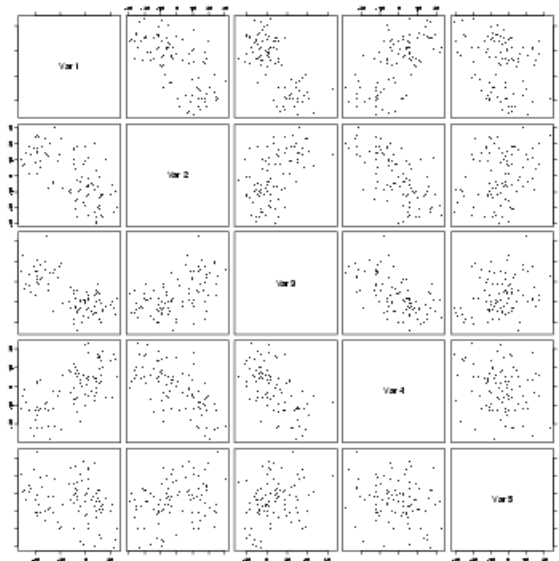
Orderings of the data matrix were first discussed in Bertin. Wegman in 1990 coined the term "data color histogram." Mike Minnotte and Webster West subsequently termed the phrase "data image" in 1998.



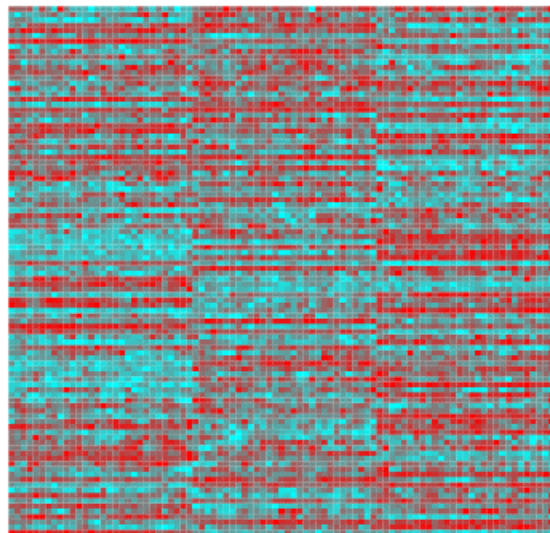
DIMACS Working Group on
Adverse Event/Disease Reportir
Surveillance, and Analysis II

Data Image Reveals Obfuscated Cluster Structure

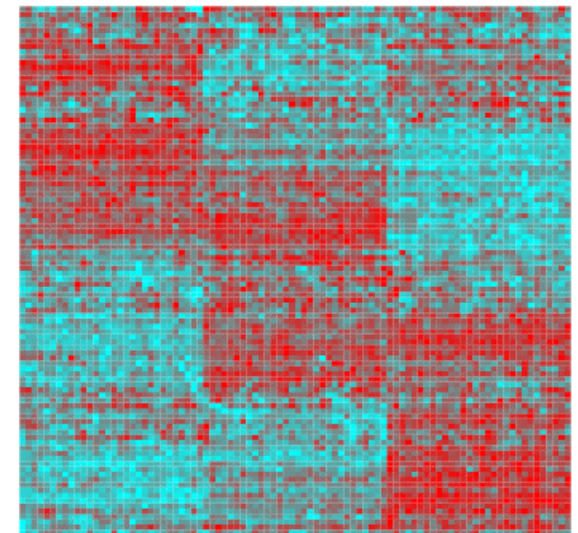
Subset of the pairs plot



Sorted on Observations

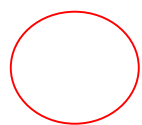
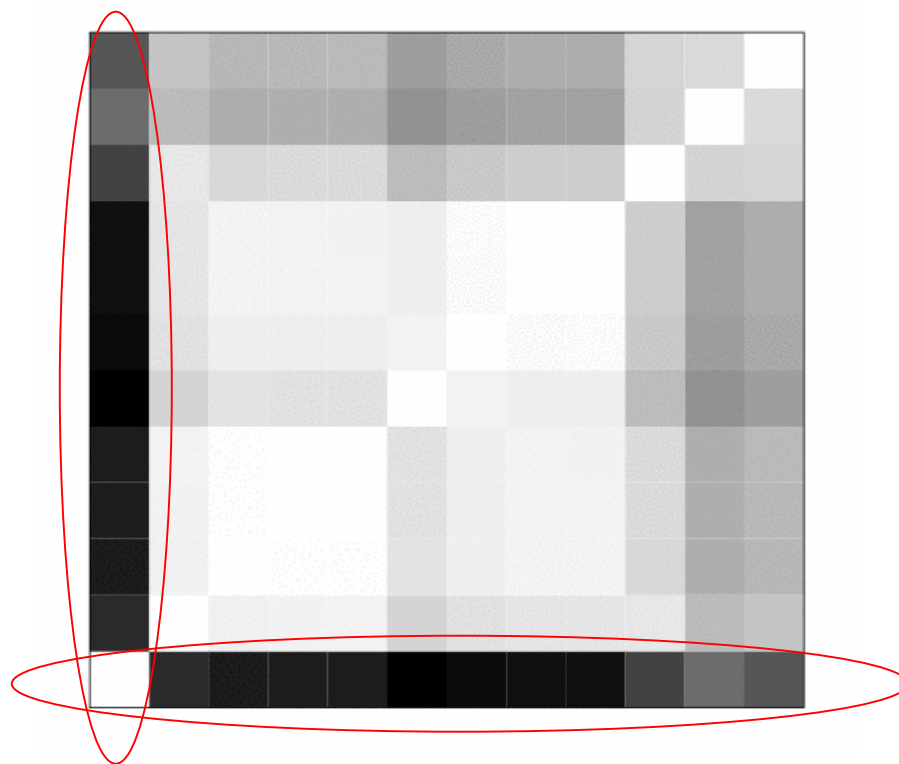
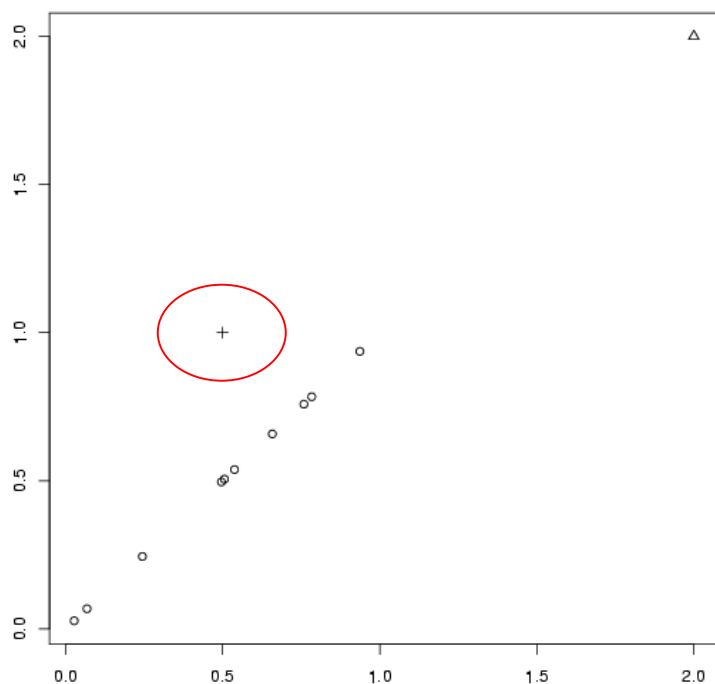


Sorted on Observations and Features



90 observations in R^{100} drawn from a standard normal distribution
The first and second 30 rows were shifted by 20 in their first and second dimensions respectively. This data matrix was then multiplied by a 100 x 100 matrix of Gaussian noise.

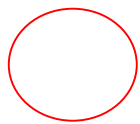
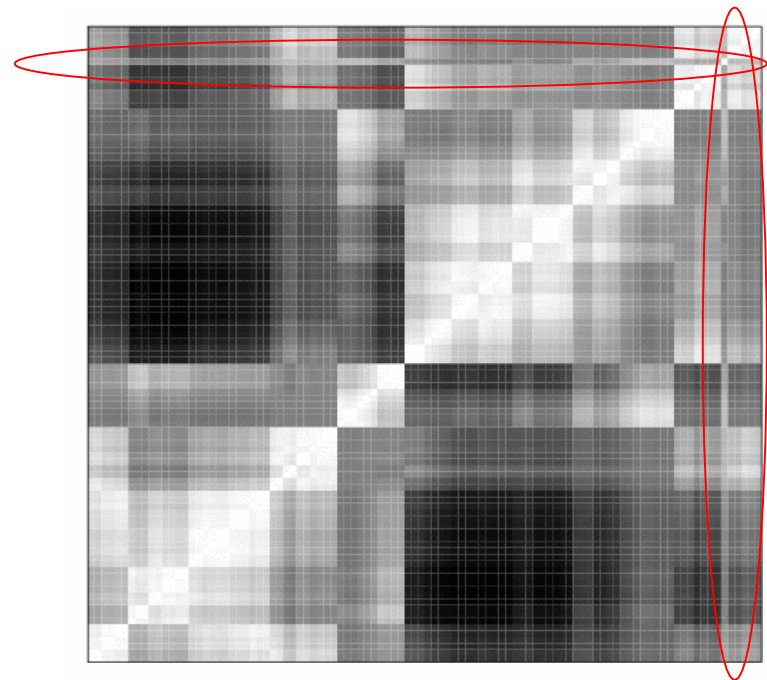
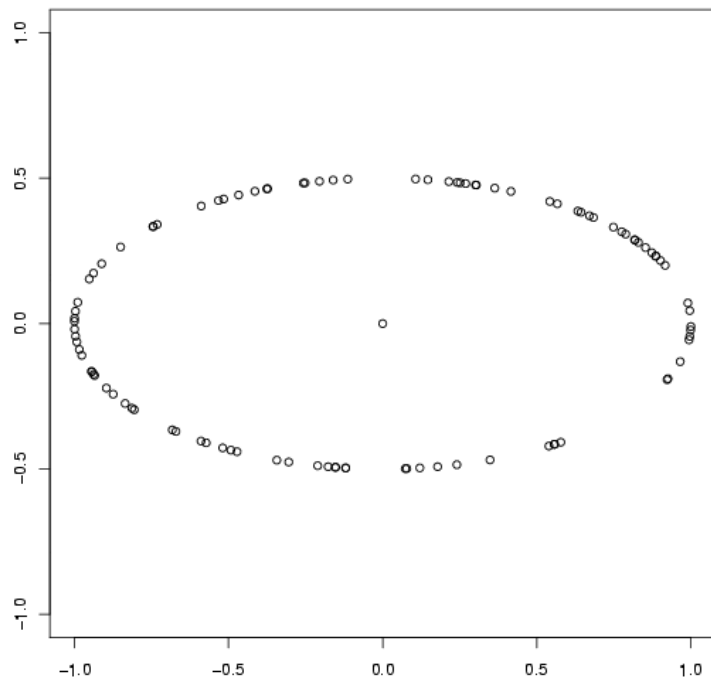
Sorted Data Image of the Interpoint Distance Matrix



Outlier

DIMACS Working Group on
Adverse Event/Disease Reporting,
Surveillance, and Analysis II

Sorted Data Image of the Interpoint Distance Matrix Reveals "Subtle" Outliers



Outlier

DIMACS Working Group on
Adverse Event/Disease Reporting,
Surveillance, and Analysis II

Distance Measures on Manifolds

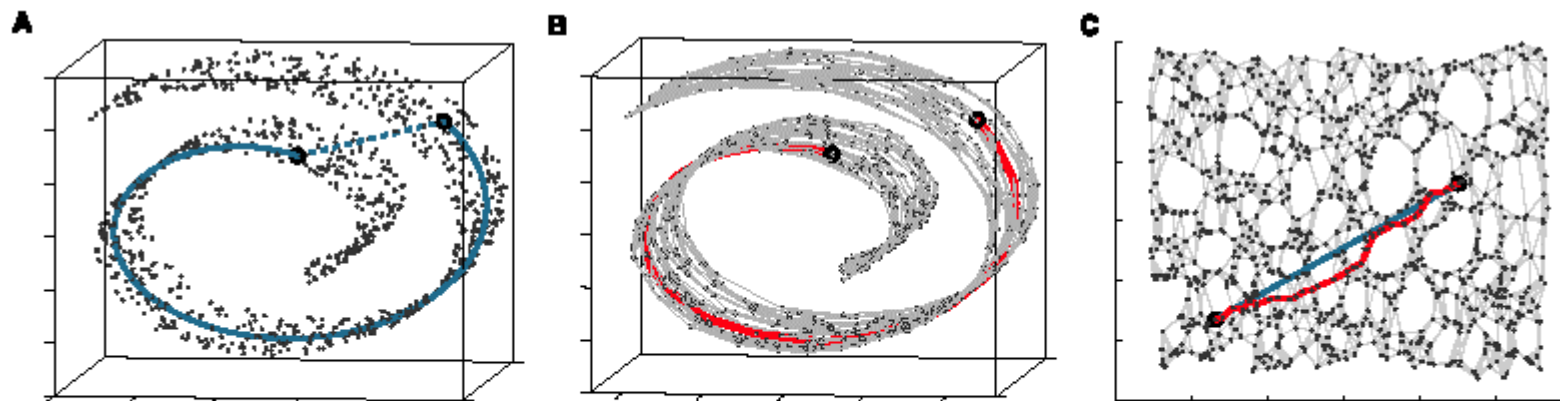


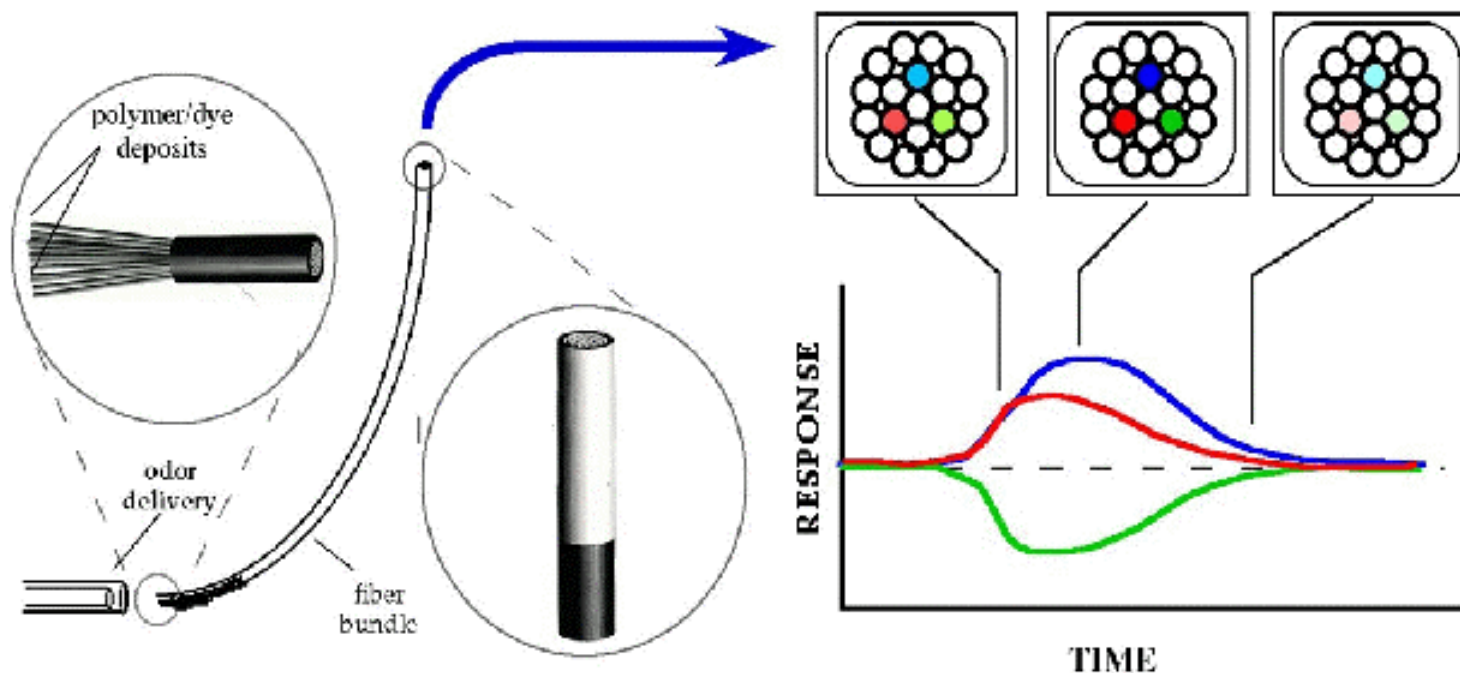
Fig. 3. The "Swiss roll" data set, illustrating how Isomap exploits geodesic paths for nonlinear dimensionality reduction. **(A)** For two arbitrary points (circled) on a nonlinear manifold, their Euclidean distance in the high-dimensional input space (length of dashed line) may not accurately reflect their intrinsic similarity, as measured by geodesic distance along the low-dimensional manifold (length of solid curve). **(B)** The neighborhood graph G constructed in step one of Isomap (with $K = 7$ and $N =$

1000 data points) allows an approximation (red segments) to the true geodesic path to be computed efficiently in step two, as the shortest path in G . **(C)** The two-dimensional embedding recovered by Isomap in step three, which best preserves the shortest path distances in the neighborhood graph (overlaid). Straight lines in the embedding (blue) now represent simpler and cleaner approximations to the true geodesic paths than do the corresponding graph paths (red).

A Global Geometric Framework for Nonlinear Dimensionality Reduction, Joshua B. Tenenbaum, Vin de Silva, John C. Langford

DIMACS Working Group on
Adverse Event/Disease Reporting,
Surveillance, and Analysis II

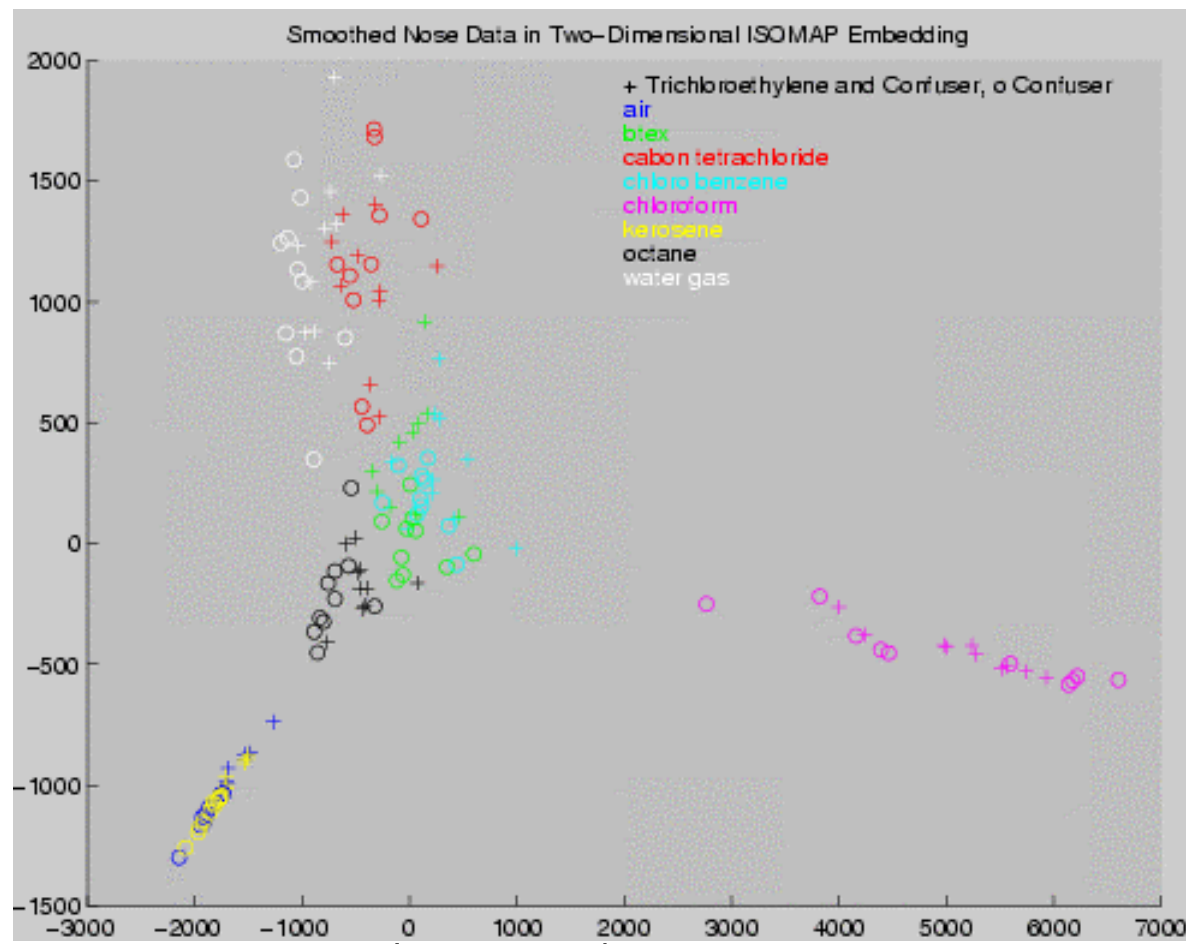
Artificial Nose Data Set



19 fibers x 2 wavelengths x 60 samples/sniff = R²²⁸⁰

DIMACS Working Group on
Adverse Event/Disease Reporting,
Surveillance, and Analysis II

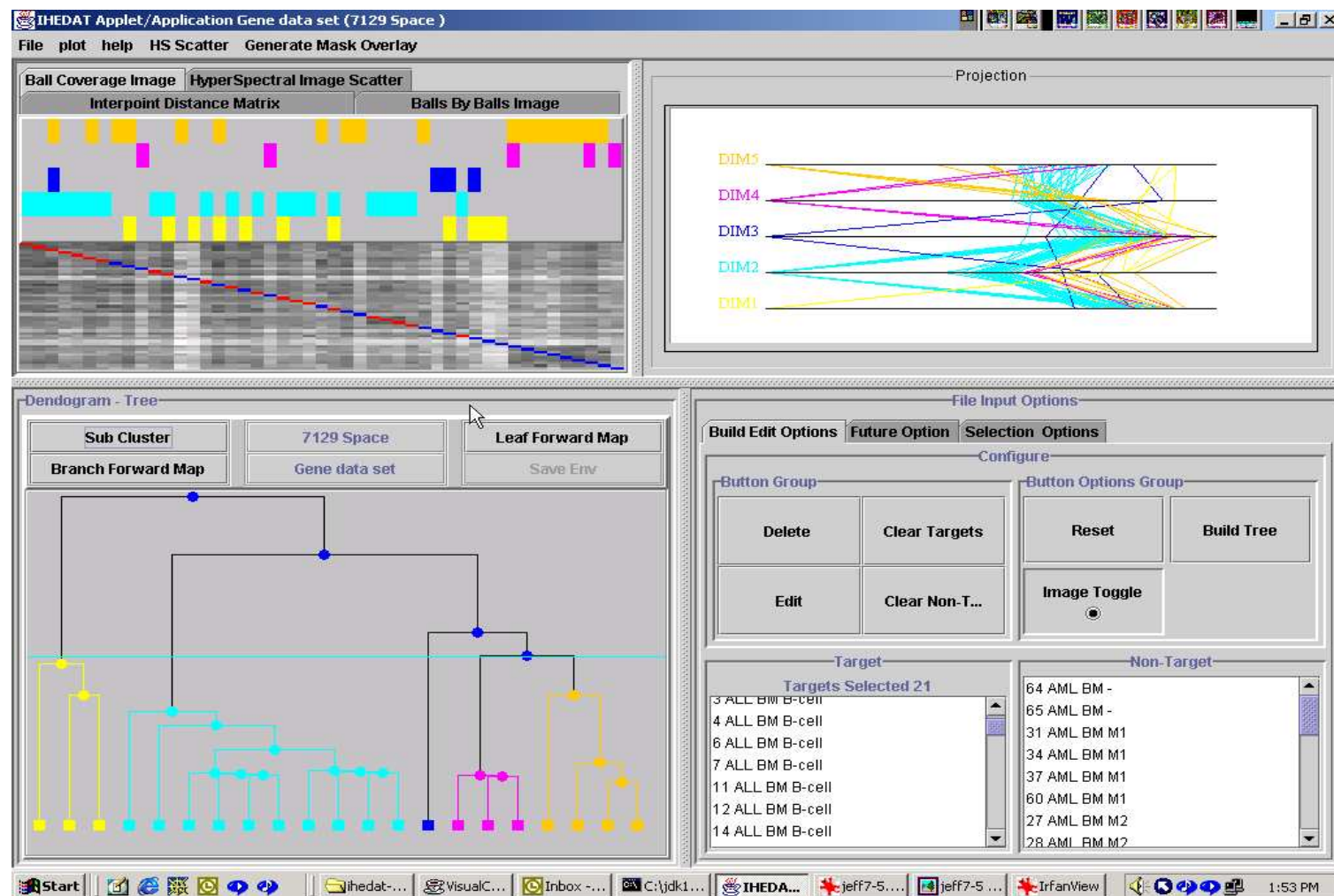
Artificial Nose ISOMAP Results



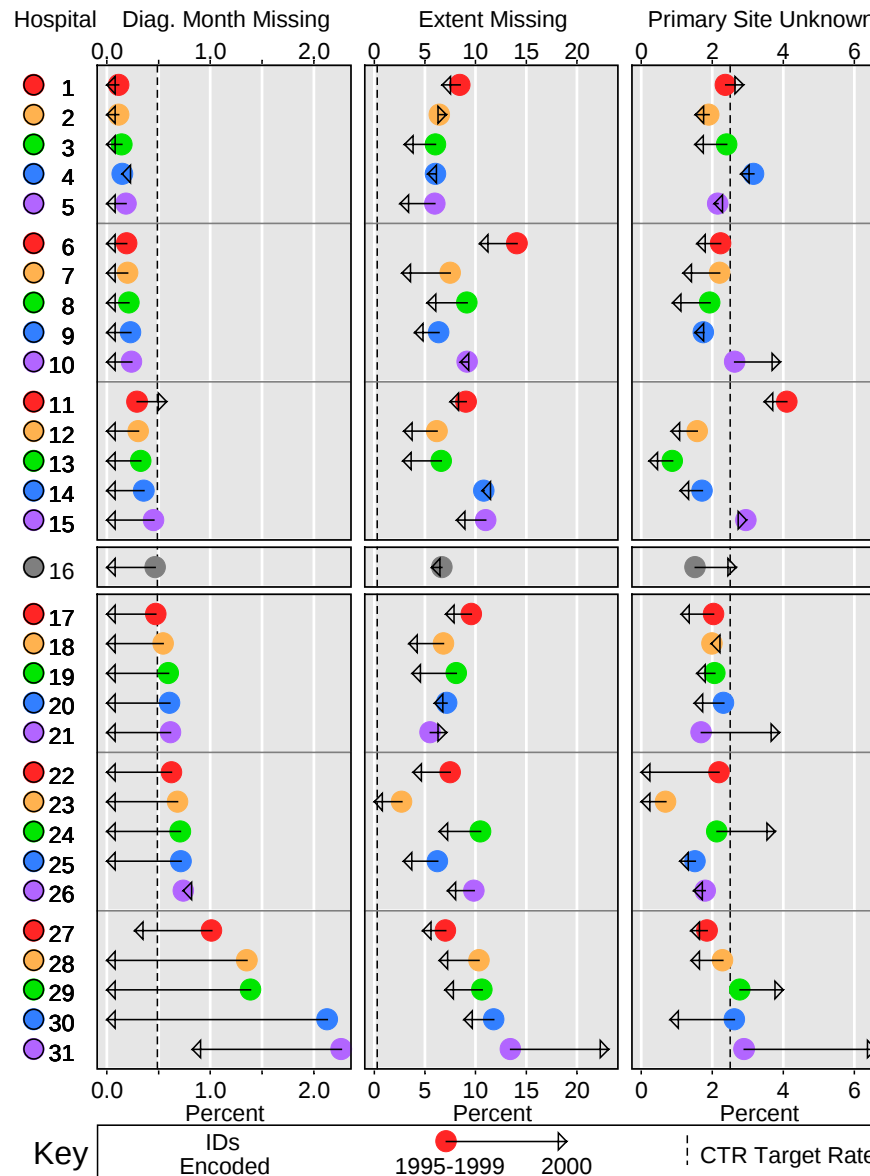
Adverse Event/Disease Reporting,
Surveillance, and Analysis II

Interactive Hyperspectral Exploratory Data Analysis Tool

*Integrated
visualization
environments can
lead to new
insights in the
hands of skilled
users.*



Connecticut Hospitals: Diagnosed Cancers Data Quality Indicators: 1995 - 1999 To 2000



*Reprinted with
permission from
"Quality Graphics
for Federal
Statistical
Summaries,"*

*Dan Carr, George
Mason University
(dcarr@gmu.edu)*

Left Column Data

Area: US - state level

Data Group: Cancer Statistics

Cancer: Lung & Bronchus

Statistic: Mortality Rate

Race: White

Sex: Males

Age: All Ages

Right Column Data (optional)

Data Group: Risk Factors / Screen

Cancer:

Statistic: Current Smokers

Race: All Races

Sex: Males

Age: 18+

Draw Clear

Overview

Options ?

State



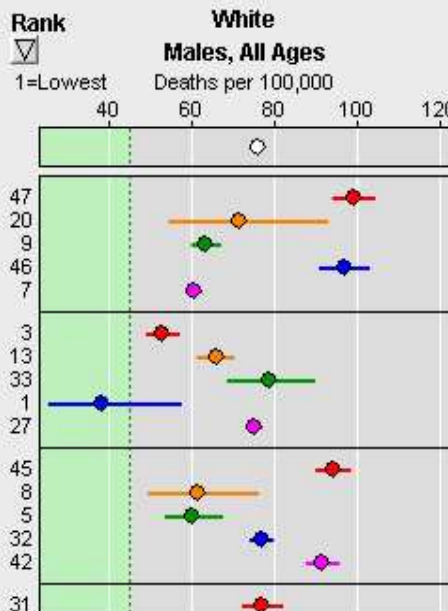
☐ United States

- ☐ Alabama
- ☐ Alaska
- ☐ Arizona
- ☐ Arkansas
- ☐ California
- ☐ Colorado
- ☐ Connecticut
- ☐ Delaware
- ☐ District of Columbia
- ☐ Florida
- ☐ Georgia
- ☐ Hawaii
- ☐ Idaho
- ☐ Illinois
- ☐ Indiana
- ☐ Iowa

Source

Lung & Bronchus Cancer Year 2000

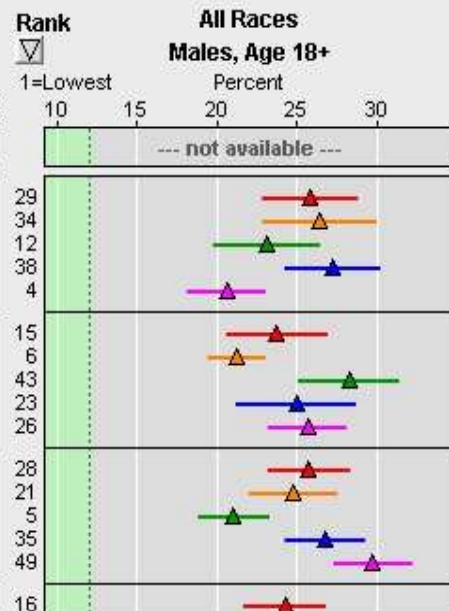
Latest Annual Death Rate



Key

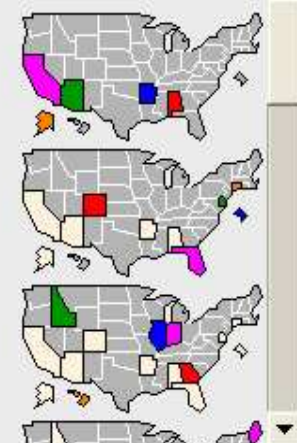
Value and 95% Confidence Interval (CI)
Median value for sorted column

Year 2001 Current Smokers



Micromaps

for sorted column



Above current map
Below current map

Left Column Data

Area: US - state level

Data Group: Cancer Statistics

Cancer: Lung & Bronchus

Statistic: Mortality Rate

Race: White

Sex: Males

Age: All Ages

Right Column Data (optional)

Data Group: Risk Factors / Screen

Cancer:

Statistic: Current Smokers

Race: All Races

Sex: Males

Age: 18+

Draw Clear

Overview

Options



State

☐ United States

☐ Kentucky
☐ Mississippi
☐ West Virginia
☐ Tennessee
☐ Alabama

☐ Arkansas
☐ Georgia
☐ Louisiana
☐ North Carolina
☐ Indiana

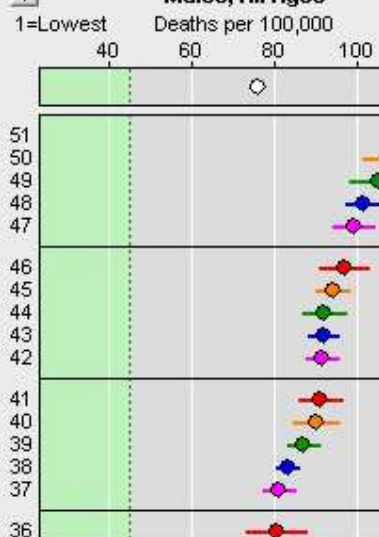
☐ Oklahoma
☐ South Carolina
☐ Missouri
☐ Ohio
☐ Virginia

☐ Maine

Lung & Bronchus Cancer Year 2000

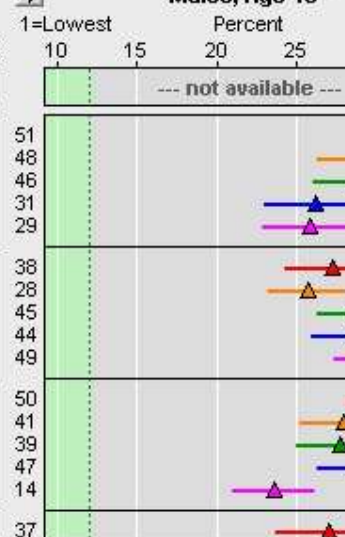
Latest Annual Death Rate

Rank
1=Lowest
White
Males, All Ages
Deaths per 100,000



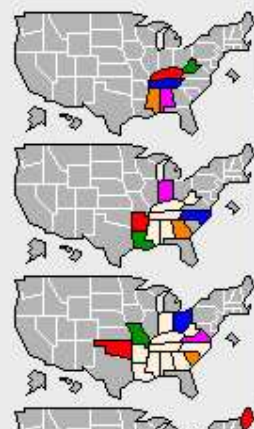
Year 2001 Current Smokers

Rank
1=Lowest
All Races
Males, Age 18+
Percent



Micromaps

for sorted column



Key

Value and 95% Confidence Interval (CI)
Median value for sorted column

Healthy People 2010 U.S. Target

Above current map
Below current map

State Cancer Profiles

Dynamic views of cancer statistics for prioritizing
cancer control efforts in the nation, states, and counties[Profiles Home](#) > Latest Rates, Percents, and Counts

Left Column Data

Area: Iowa Counties

Data Group: Cancer Statistics

Cancer: Lung & Bronchus

Statistic: Mortality Rate

Race: White

Sex: Males

Age: All Ages

Right Column Data (optional)

Data Group: Risk Factors / Scree

Cancer:

Statistic: Current Smokers

Race: All Races

Sex: Males

Age: 18+

Draw

Clear

Overview

Options



Counties (99)

☐ United States☐ Iowa

☐ Davis
☐ Jones
☐ Pottawattamie
☐ Marshall
☐ Mahaska

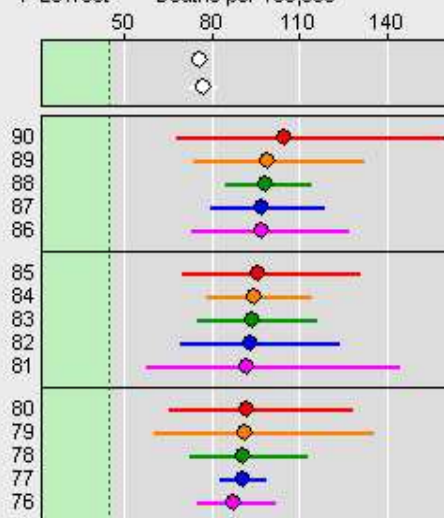
☐ Hamilton
☐ Cerro Gordo
☐ Lee
☐ Jackson
☐ Worth

☐ Harrison
☐ Keokuk
☐ Jasper
☐ Polk
☐ Woodbury

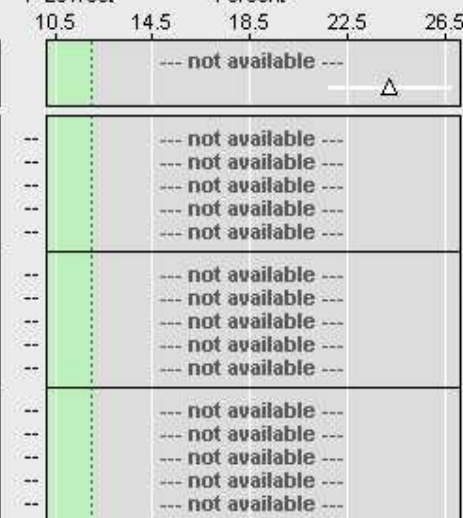
Lung & Bronchus Cancer
Year 1996-2000

Latest Annual Death Rate

Rank
 1=Lowest
 Deaths per 100,000

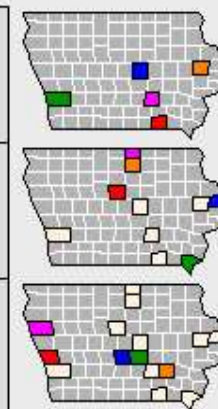


Rank
 1=Lowest
 Percent



Micromaps

for sorted column



Key

Value and 95% Confidence Interval (CI)
 Median value for sorted column

Healthy People 2010 U.S. Target
 See data table for source information

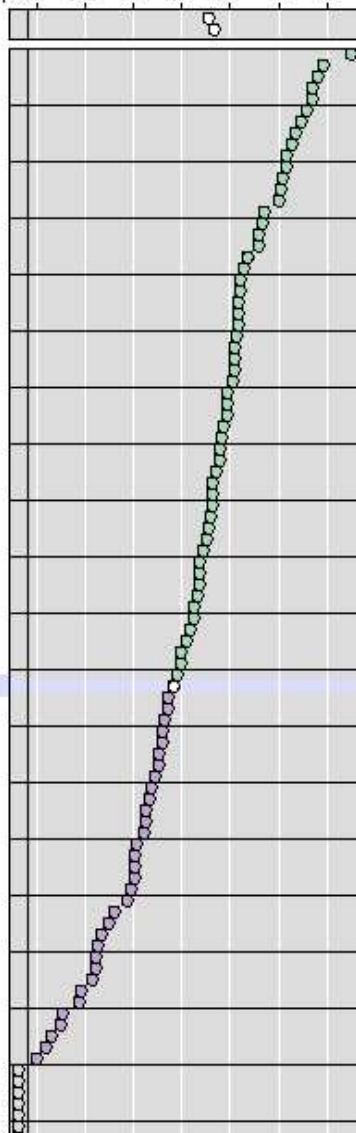
Above current map
 Below current map

Which Column: Left

Zoom in

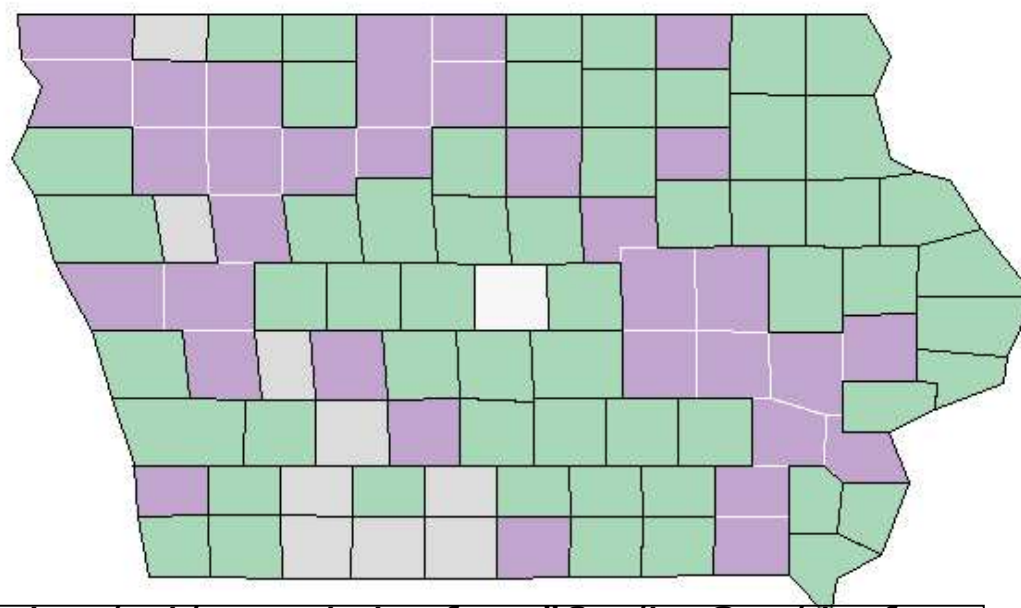
Change colors

Deaths per 100,000
NA / Sup 40 50 60 70 80 90 100



□ Story

Lung & Bronchus Latest Annual Death Rate By County For All Ages, White, Males



Reprinted with permission from "Quality Graphics for Federal Statistical Summaries,"

Dan Carr, George Mason University (dcarr@gmu.edu)

Story County:
1996-2000 Mortality Rate = 68.4, 95% CI = (54.7, 84.8)
Rank = 34

Key: □ Selected county ■ County above selection ■ County below selection

Conditioned Choropleth Map Program

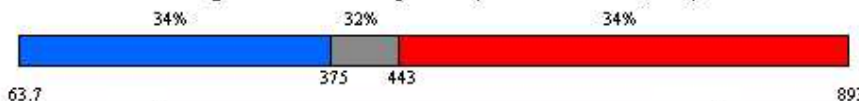
File Variable Edit View Zoom (Play List) (Help)

White Males Ages 65-74

1988-1992

Regions: Health Service Areas

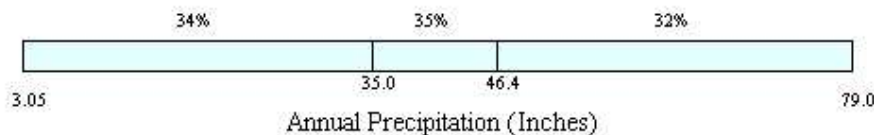
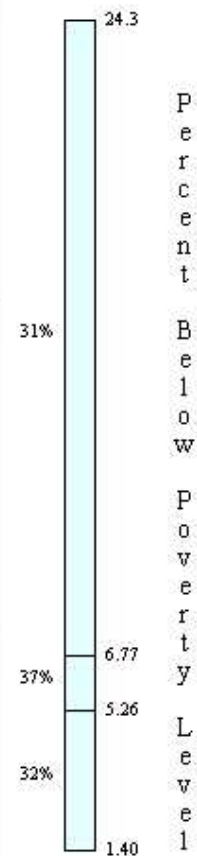
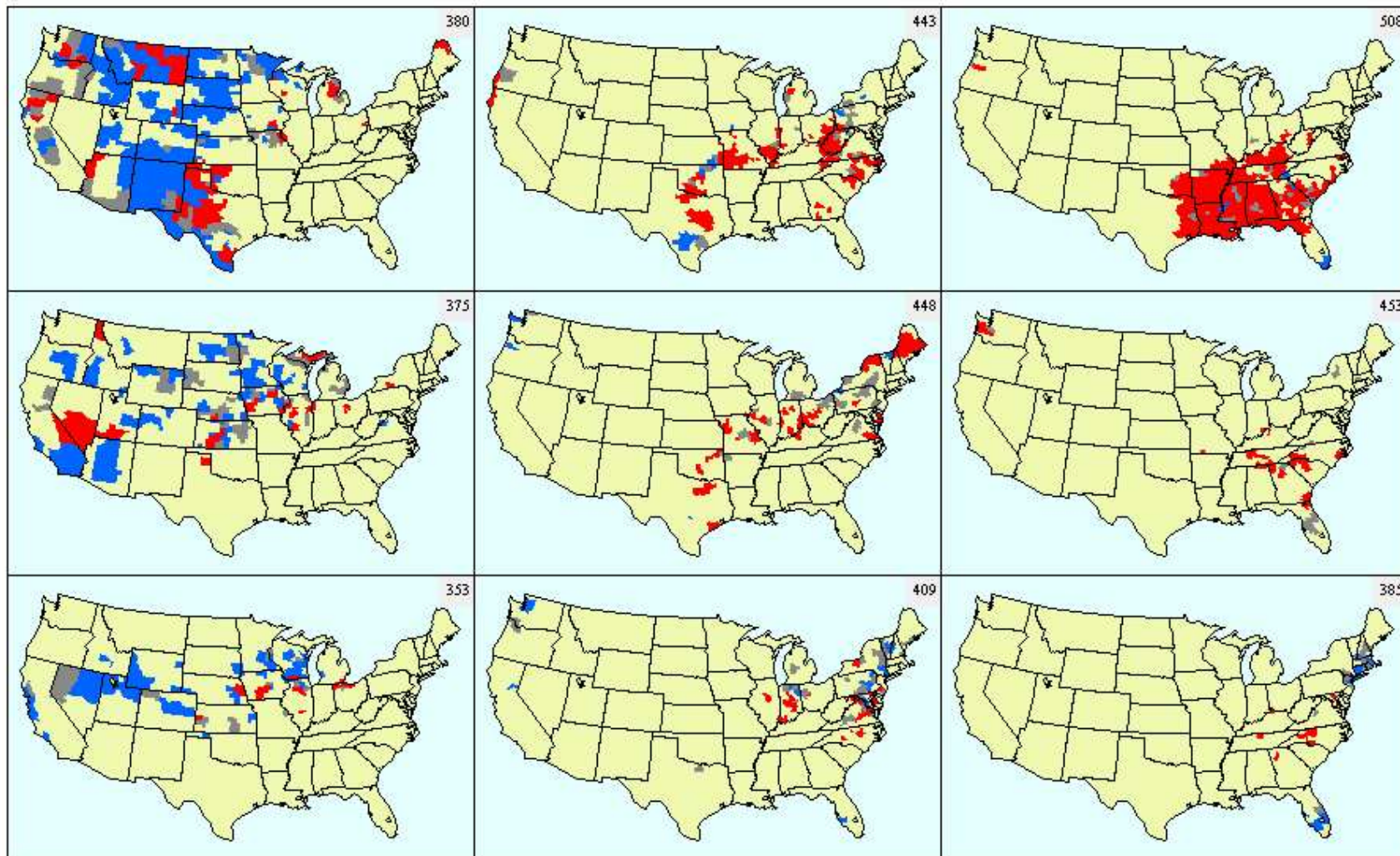
Lung Cancer Mortality Rate (Deaths Per 100,000)



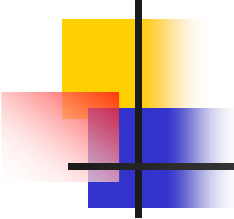
Weight: Population

Statist. Med. 2000;19:2521-2538

www.galaxy.gmu.edu/~dcarr/ccmaps

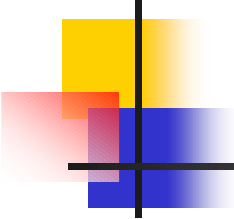


Reprinted with permission from "Quality Graphics for Federal Statistical Summaries," Dan Carr, George Mason University (dcarr@gmu.edu)



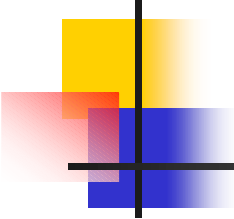
Recommendations

- **Build visualization into all of the software systems from the ground up**
- **Seek ways to exploit the human visual system as an aid in pattern identification/hypothesis generation**
- **Look to exploit multi-technique synthesis where appropriate**
- **Don't relegate oneself to simple Euclidean geometry let the inherent geometric structure in the data come to the forefront**
 - **Incorporate relevant barriers into distance calculations**
- **Provide the ability for the analyst to annotate and log their analysis**
- **Additional funding is needed to develop more user friendly gui development programs**
- **Use statistical summaries to facilitate hotspot identification**
- **Algorithms to prioritize the presentation of data views to the user in order to facilitate subtle pattern discovery**



References - I

- Jacques Bertin, *Semiologie Graphique*, ISBN 2 -1732-1277-4, 1967.
- Dan Carr, <http://statecancerprofiles.cancer.gov/micromaps>, 2003.
- David J. Marchette and Jeffrey L. Solka, "Using data images for outlier detection," *Computational Statistics and Data Analysis*, pp. 541-552, 2003.
- Michael C. Minnotte and R. Webster West, "The data image a tool for exploring high-dimensional datasets," *1998 Proceedings of the ASA Section on Statistical Graphics*, 1998.
- Carey E. Priebe, Jeffrey L. Solka, David J. Marchette, and B. Ted Clark, "Class cover catch digraphs for latent class discovery in gene expression monitoring by DNA microarrays," *Computational Statistics and Data Analysis*, 43, pp. 621-632, 2003.



References - II

- Jeffrey L. Solka, Edward J. Wegman, and David J. Marchette, "Data mining strategies for the detection of chemical warfare agents," in *Statistical Data Mining and Knowledge Discovery*, (Hamparsum Bozdogan, ed.), 57-92, Boca Raton: Chapman and Hall/CRC, 2003.
- Jeffrey L. Solka, Barton T. Clark, and Jeffrey L. Solka, "A visualization framework for the analysis of hyperspectral data," *International Journal of Image and Graphics*, Vol. 2, No. 1, pp. 145-161, 2002.
- Joshua B. Tenenbaum, Vin de Silva, John C. Langford A Global Geometric Framework for Nonlinear Dimensionality Reduction, Science 2001?
- Edward J. Wegman, "Hyperdimensional data analysis using parallel coordinates," *Journal of the American Statistical Association*, 85:664-675, 1990.
- Edward J. Wegman and Alan Dorfman, "Visualizing cereal world," *Computational Statistics and Data Analysis*, 43, pp. 633-649, 2003.

DIMACS Working Group on
Adverse Event/Disease Reporting,
Surveillance, and Analysis II