Introduction

Databases and Visualization

Database Support
for Visualization Systems

**Visualization Support for
Database Systems**

**Data
Visualization**

Query
Visualization

Schema
Visualization

**Exploratory
Analysis**

Confirmatory
Analysis

Presentation
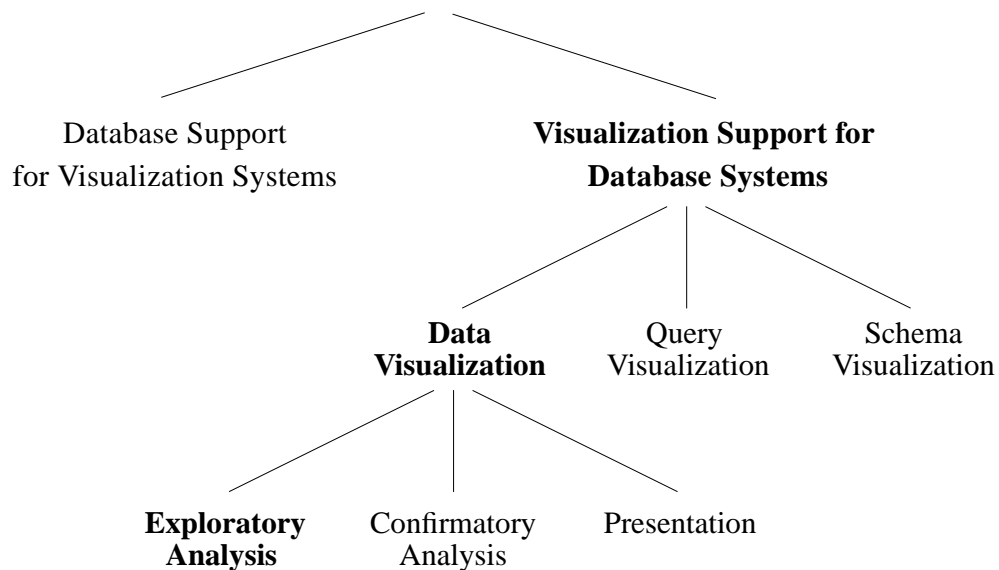
Introduction

Database Support of Visualization Systems:
Specific Requirements

❑ Specific Data Types (extensibility important)
(e.g., 2D- & 3D-grids, arbitrary graphs, ...)

❑ Specific Operations (extensibility important)

⇨ data selection (e.g., sampling, subsetting, ...)

⇨ data manipulation (e.g., filter, interpolation, ...)

⇨ data representation (e.g., attribute mapping, color mapping, ...)

⇨ viewing (e.g., zoom, rotation, ...)

❑ Specific Transaction Types (read-only)
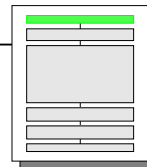
❑ Specific Performance Requirements (real-time)

Introduction
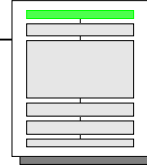
Objectives of
Visualization Support of Database Systems

❑ Visualization of the Database Schema
(for database design)

❑ Visualization of Queries
(for query specification)

❑ **Visualization of the Data**
(for database exploration)

Introduction

Visual Query Interfaces

❑ Tabular Query Interfaces
(e.g., QBE [Zlo 77], Example-based Graphical Query Interfaces [OW 93])

❑ Graphical Support of Textual Query Languages
(e.g., GRADI [KL 93])

❑ Schema-oriented Query Interfaces (e.g., SNAP [BH 86])

❑ Slider-based Query Interfaces (e.g., IVEE [AW 95], VisDB [KK 95])

Problem: specification of complex queries is difficult

⇨ example of a query interface for complex queries:
InfoCrystal [Spo 93]

Introduction
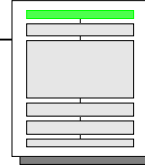
Goals of Data Visualization

❑ **Explorative Analysis**
- starting point: data without hypotheses about the data
- process: interactive, usually undirected search for structures, trends, etc.
- result: visualization of the data, which provides hypotheses about the data

❑ Confirmative Analysis
- starting point: hypotheses about the data
- process: goal-oriented examination of the hypotheses
- result: visualization of the data, which allows the confirmation
    or rejection of the hypotheses

❑ Presentation
- starting point: facts to be presented are fixed a priori
- process: choice of an appropriate presentation technique
- result: high-quality visualization of the data presenting the facts

Introduction

Historical Overview of Exploratory Data Visualization Techniques (cf. [WB 95])

❑ pioneering work of Tufte [Tuf 83, Tuf 90] and Bertin [Ber 81] focuses on
⇨ visualization of data with inherent 2D-/3D-semantics
⇨ general rules for layout, color composition, attribute mapping, etc.

❑ development of visualization techniques for different types
of data with an underlying physical model
⇨ geographic data, CAD data, flow data, image data, voxel data, etc.

❑ development of visualization techniques for arbitrary multidimensional
data (without an underlying physical model)
⇨ applicable to databases and other information resources

# Introduction
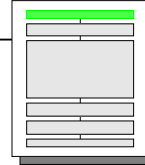
Classification of Data Visualization Techniques

❏ Geometric Techniques: Scatterplots, Projection Pursuit, Prosection Views, Hyperslice, *Parallel Coordinates*, ...

❏ Icon-based Techniques: Chernoff Faces, *Stick Figures*, Shape-Coding, Grouping Technique...

❏ Pixel-oriented Techniques: *Recursive Pattern Technique*, Spiral- & Axes-Techniques, ...

❏ Hierarchical Techniques: Dimensional Stacking, Hierarchical Plotting, Worlds-within-Worlds, *Treemap*, Cone Trees, ...

❏ Graph-Based Techniques: SeeNet, *Hygraphs*, Fisheye Views, 3D-Graphs (Narcissus), ...

❏ 3D-Techniques: *Perspective Wall*, Hyperbox, Landscapes, Cone Trees, 3D-Graphs (Narcissus), ...

❏ Dynamic Techniques: Dyn. Projections, *Dyn. Linking & Brushing*, Dyn. Environment, *Dyn. Zooming*, Dyn. Detail on Demand, Dyn. Data-to-Visualization Mapping, ...

❏ Hybrid Techniques: arbitrary combinations from above


Data Preprocessing Techniques
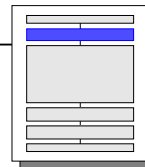
❏ **Techniques for Dimension Reduction**

(Set of d-dim Data Items -> Set of k-dim. Data Items; k << d)

• Principal Component Analysis [DE 82]

Determines a minimal set of principal components (linear combinations of the original dimensions) which explain the main variations of the data.

• Factor Analysis [Har 67]

Determines a set of unobservable common factors which explain the main variations of the data. The original dimensions are linear combinations of the common factors.

• Multidimensional Scaling [SRN 72]

Uses the similarity (or dissimilarity) matrix of the data as defining coordinate axes in multidimensional space. The Euclidean distance in that space is a measure of the similarity of the data items.

• Fastmap [FL 95]

Fastmap also operates on a given similarity matrix and iteratively reduces the number of dimensions while preserving the distances as much as possible.

Data Preprocessing Techniques
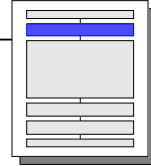
❑ **Subsetting Techniques**

(Set of Data Items -> Subset of Data Items)

- Sampling (determines a representative subset of the database)
- Querying (determines a certain, usually a-priori fixed subset of the database)

❑ **Segmentation Techniques**

(Set of Data Items -> Set of (Set of Data Items))

- Segmentation based upon attribute values or attribute ranges

❑ **Aggregation Techniques**

(Set of Data Items -> Set of Aggregate Values)

- Aggregation (sum, count, min, max, ...) based upon
  - attribute values
  - topological properties, etc.
- Visualizations of Aggregations:
  - Histograms
  - Pie Charts, Bar Charts, Line Graphs, etc.

Geometric Techniques
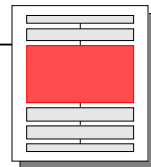
Basic Idea: Visualization of geometric transformations and projections of the data.

Overview

❑ **Scatterplot-Matrices [And 72, Cle 93]**

❑ **Projection Pursuit Techniques [Hub 85]**

(⇨ techniques for finding meaningful projections of multidimensional data)

❑ **Prosection Views [FB 94, STDS 95]**

❑ **Hyperslice [WL 93]**

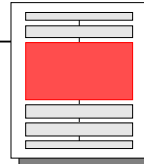❑ **Parallel Coordinates [Ins 85, ID 90]**

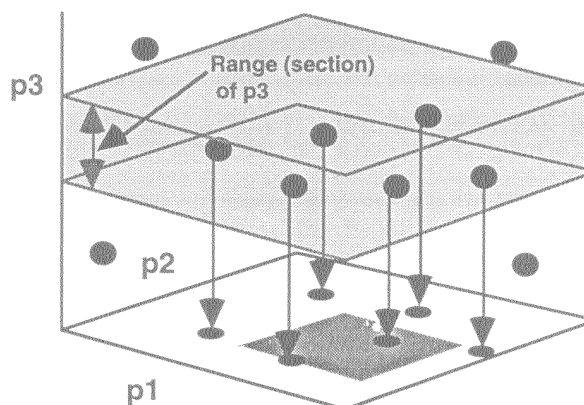# Geometric Techniques
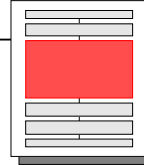
Scatterplot-Matrices [Cle 93]
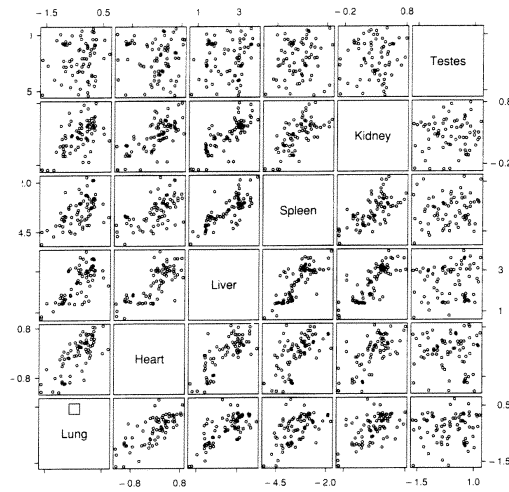
⇨ matrix of scatterplots (x-y-diagrams) of the k-dim. data
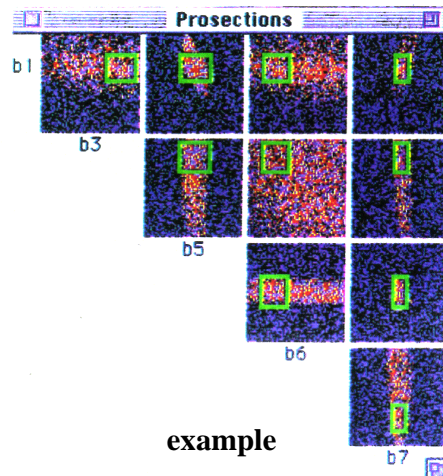[total of ($k^2/2$ - k) scatterplots]

Geometric Techniques

Prosection Views [FB 94, STDS 95]
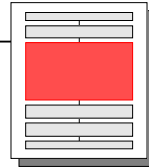

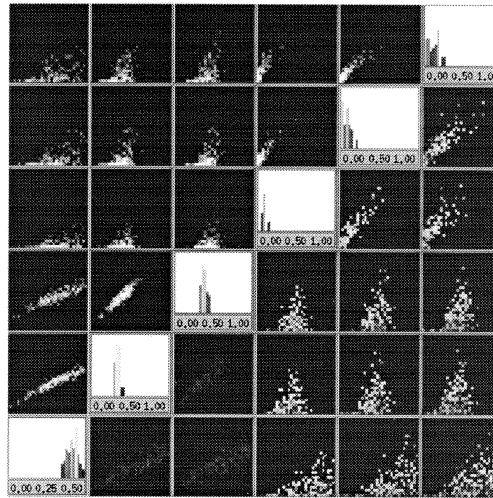**schematic representation**   **example**

⇨ matrix of all orthogonal projections of a certain multidimensional range
(combination of selections and projections)
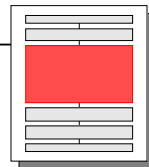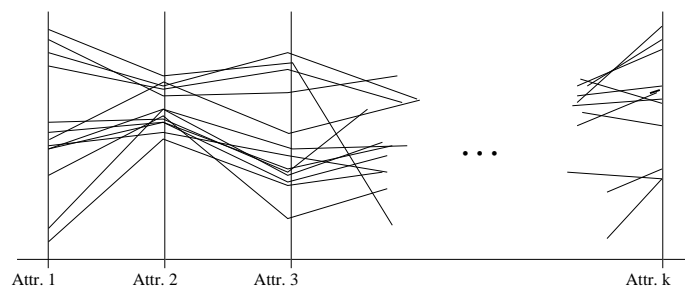
Geometric Techniques

Hyperslice [WL 93]

⇨ matrix of k^2 slices through the k-dim. data (the slices are determined interactively)
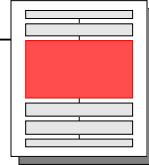
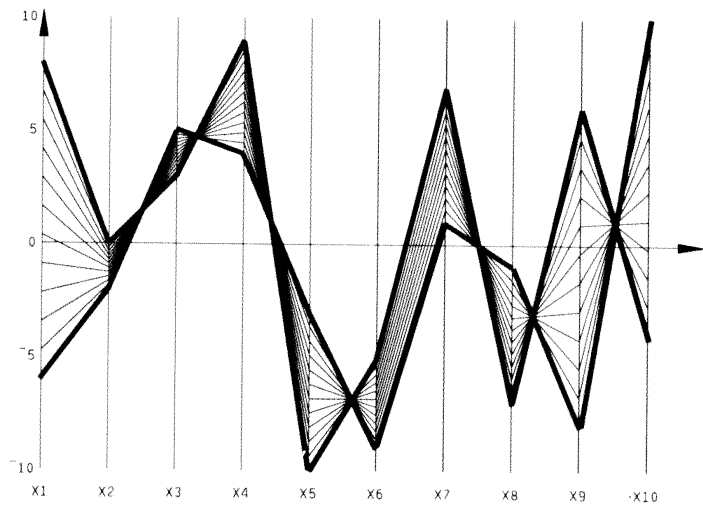# Geometric Techniques

Parallel Coordinates [Ins 85, ID 90]

⇨ n equidistant axes which are parallel to one of the screen axes and correspond to the attributes

⇨ the axes are scaled to the [minimum, maximum] - range of the corresponding attribute

⇨ every data item corresponds to a polygonal line which intersects each of the axes at the point which corresponds to the value for the attribute
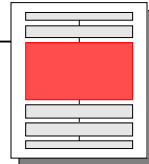
Geometric Techniques
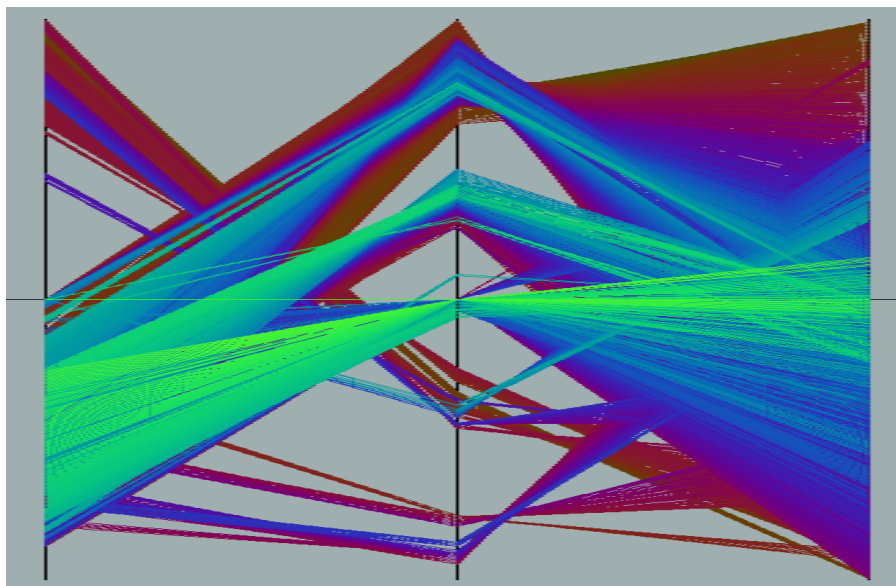
Parallel Coordinates (cont'd)

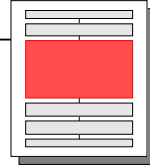points on a line in 10-dim. space

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Geometric Techniques

Parallel Coordinates (cont'd)
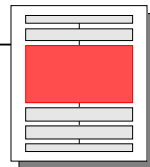

example of 3-dim. data

Icon-based Techniques

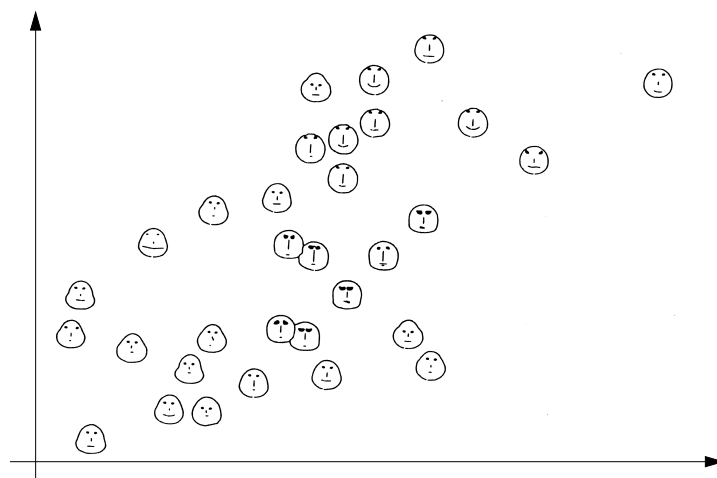**Basic Idea**: Visualization of the data values as features of icons.

Overview

❑ **Chernoff-Faces [Che 73, Tuf 83]**

❑ **Stick Figures [Pic 70, PG 88]**

❑ **Shape Coding [Bed 90]**

❑ **Grouping Technique [KK 94]**
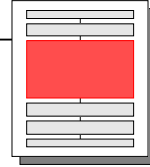

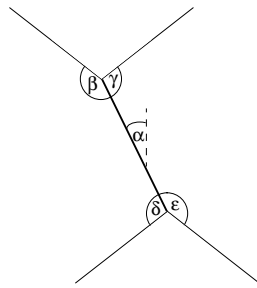
Icon-based Techniques

Chernoff-Faces [Che 73, Tuf 83]

⇨ visualization of the multidim. data using the properties of a face icon
(shape of nose, mouth, eyes, and the shape of the face itself)

Stick Figures [Pic 70, PG 88]

⇨ visualization of the multidimensional data using stick figure icons

⇨ two attributes of the data are mapped to the display axes and the remaining attributes are mapped to the angle and/or length of the limbs

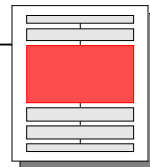⇨ texture patterns in the visualization show data characteristics

Stick Figure Icon
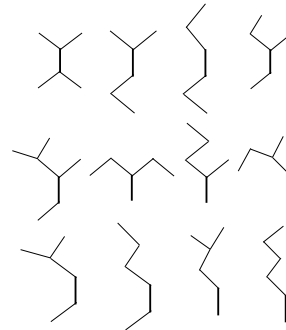
A Family of Stick Figures

Stick Figures (cont'd)
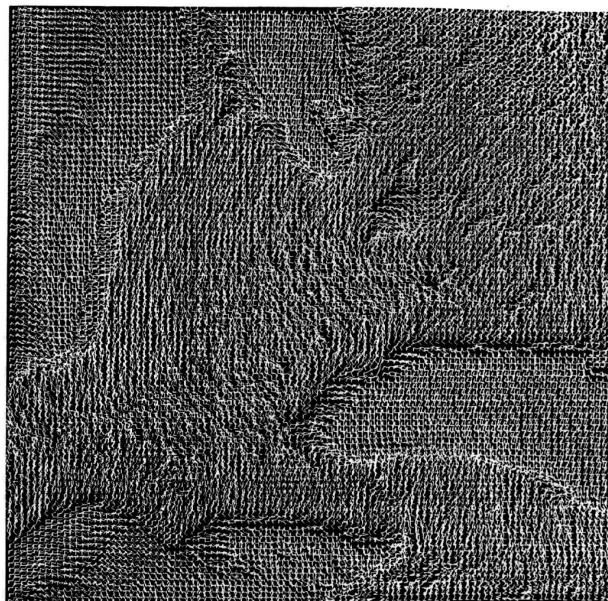
5-dim. image data from the great lake region

Icon-based Techniques
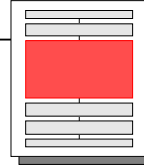
Stick Figures (cont'd)
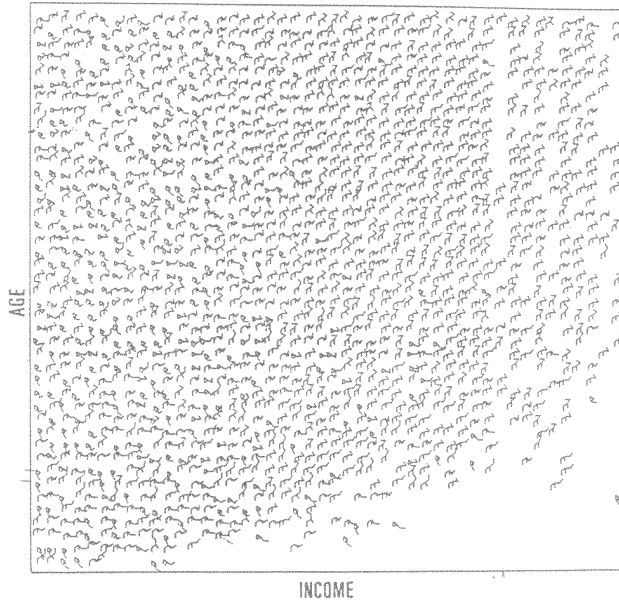
census data showing
age, income, sex,
education, etc.

Icon-based Techniques
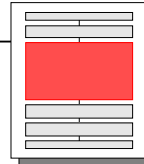
Shape Coding [Bed 90]

⇨ the data are visualized using small arrays of fields

⇨ each field represents one attribute value

⇨ arrangement of attribute fields (e.g., 12-dimensional data):
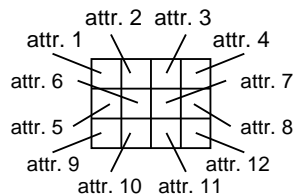
⇨ arrays are arranged line-by-line according to a given sorting
(e.g., the time attribute for time-series data)
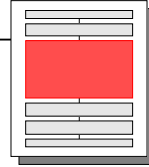
Shape Coding (cont'd)
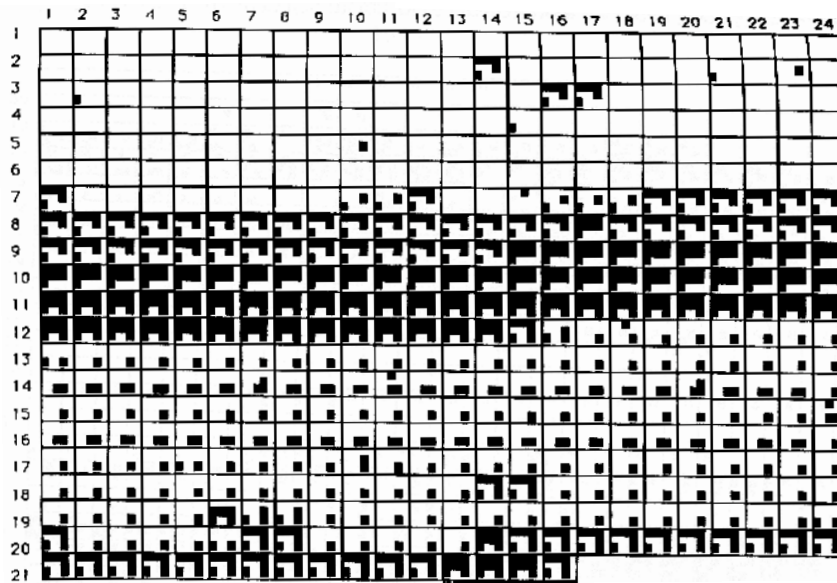
time series of
NASA earth
observation data
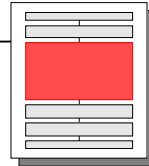
Grouping Technique [KK 94]

⇨ visualization of the data using color icons

⇨ color icons are array of color fields representing the attribute values

⇨ arrangement is query-dependent (e.g., spiral)

schematic
representation
of 6-dim. data

attr. 1   attr. 2   attr. 3

attr. 4   attr. 5   attr. 6
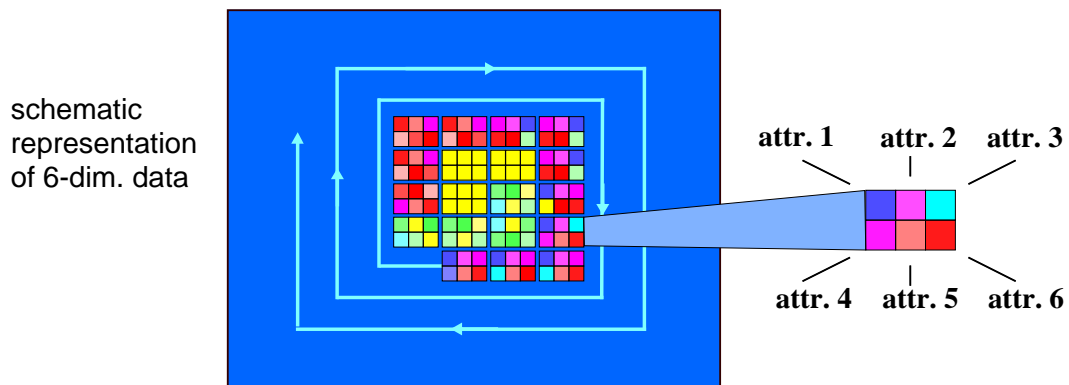
Icon-based Techniques
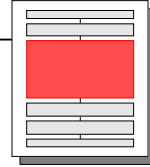
Grouping Technique (cont'd)
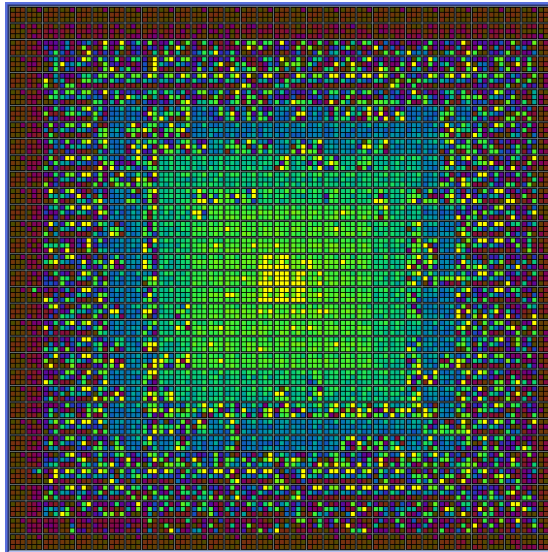
random data containing
several clusters

Pixel-oriented Techniques
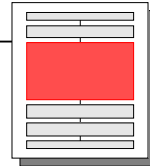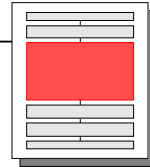
Basic Idea

- each attribute value is represented by one colored pixel
  (⇨ the value ranges of the attributes are mapped to a fixed colormap)

- the attribute values for each attribute are presented in separate subwindows

**visualization of
six-dim. data**

attribute 1   attribute 2   attribute 3

**attribute values
of a data item**

attribute 4   attribute 5   attribute 6
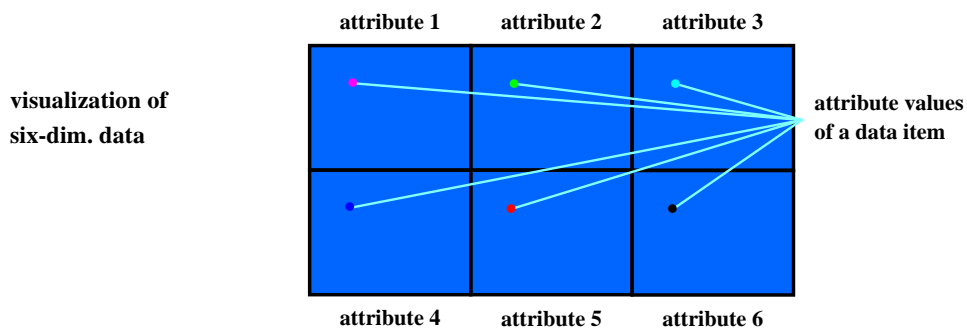
Pixel-oriented Techniques

Overview

Simple Techniques — line-by-line
column-by-column ...

Query-Independent Techniques — Space-Filling Curves — Peano-Hilbert
Morton ...

Recursive Pattern Technique

Query-Dependent Techniques — Spiral Technique
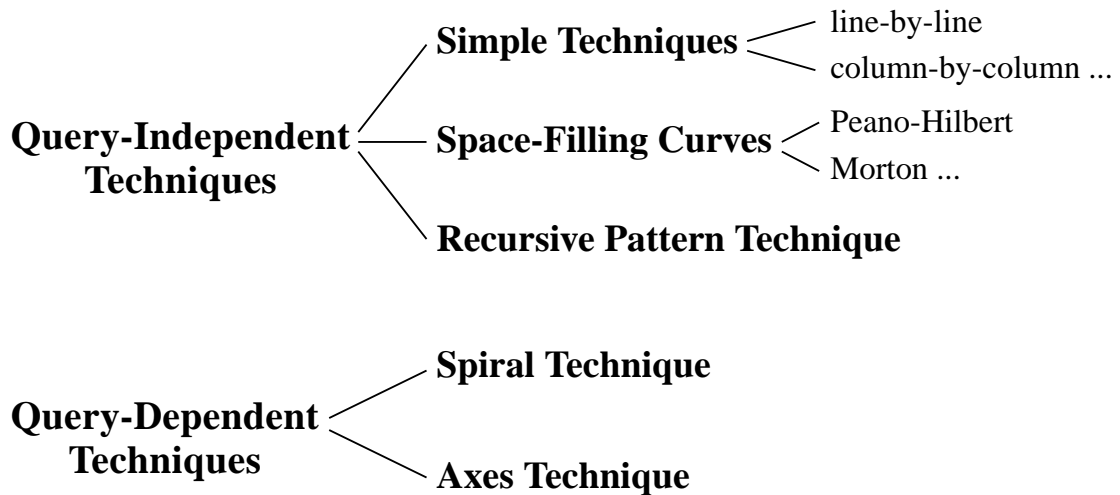
Axes Technique

Pixel-oriented Techniques
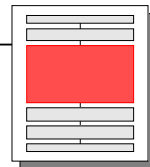
Query-Independent Techniques:
Simple Arrangement (Line-by-Line)
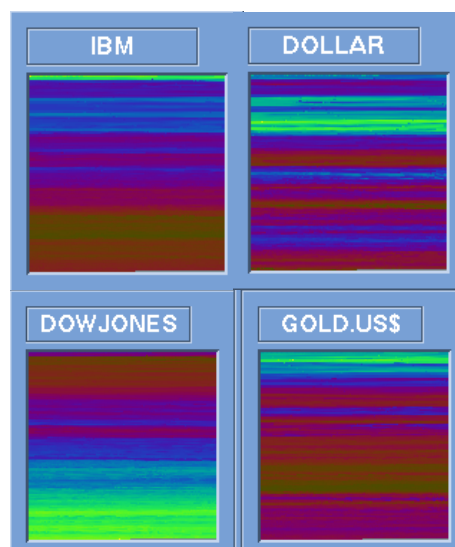
time series of
financial data

Pixel-oriented Techniques
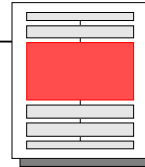
Query-Independent Techniques:
Space-Filling Curve Arrangements

IBM	DOLLAR	IBM	DOLLAR

DOW JONES	GOLD.US$	DOW JONES	GOLD.US$
Peano-Hilbert		**Morton (Z-Curve)**	
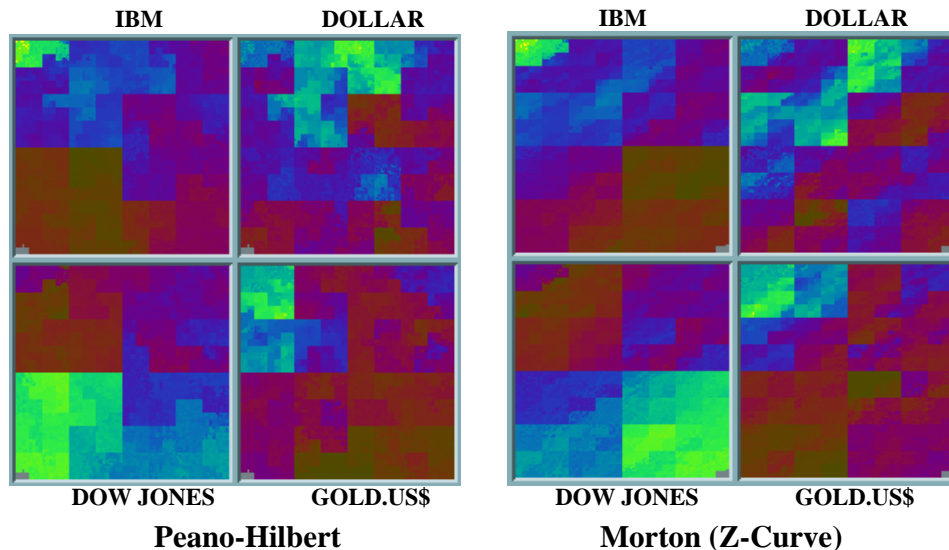
Pixel-oriented Techniques
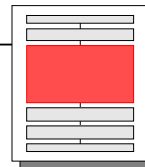
Query-Independent Techniques:
Recursive Pattern Technique

- recursive generalization of an iterated line- and column-based arrangement

- the user may specify the height h_i and width w_i for each recursion level

- on recursion level i, w_i patterns of recursion level *(i-1)* are drawn in left-right direction and this is repeated h_i times in top-down direction

  => the pattern on recursion level *i* consists of $w_i \times h_i$ patterns of recursion level *(i-1)*

Pixel-oriented Techniques

Recursive Pattern: Example of a Structured Arrangement

① ② columns 3 - 11 ⑫

row 1

row 2

rows 3 -7

row 8
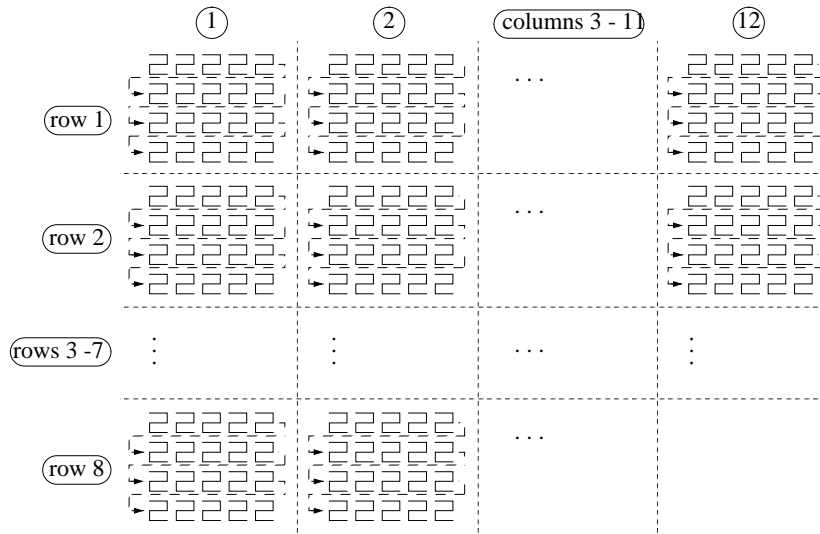
$(w_1, h_1) = (3, 3), (w_2, h_2) = (5,1), (w_3, h_3) = (1, 4), (w_4, h_4) = (12, 1)$ and $(w_5, h_5) = (1, 8)$

Pixel-oriented Techniques

Recursive Pattern: Example of Financial Data

IBM   **DOLLAR**

time series of financial data

DOW JONES   **GOLD (US$)**

Recursive Pattern: FAZ-Index (Jan. '74 - Apr. '95)
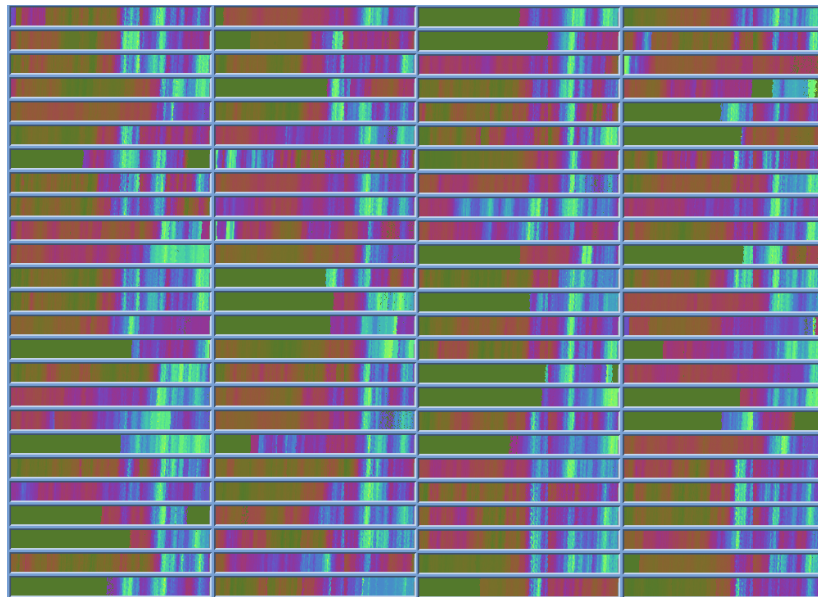
time series of
stock data
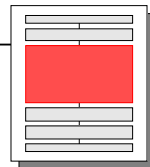
Query-Dependent Techniques:
Basic Idea

- data items $(a_1, a_2, ..., a_m)$ & query $(q_1, q_2, ..., q_m)$

$$\Rightarrow \quad \text{distances } (d_1, d_2, ... d_m)$$

- extend distances by overall distance (d_{m+1})

- determine data items with lowest overall distances

- map distances to color (for each attribute)

- visualize each distance value d_i by one colored pixel

Pixel-oriented Techniques
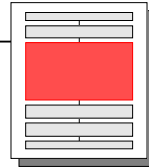
Query-Dependent Techniques

Attribute j

pos

neg

neg   pos

Attribute i
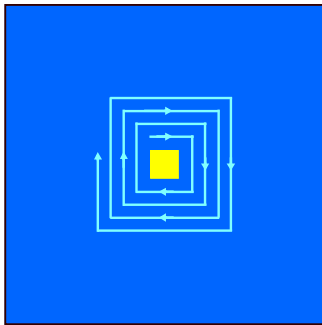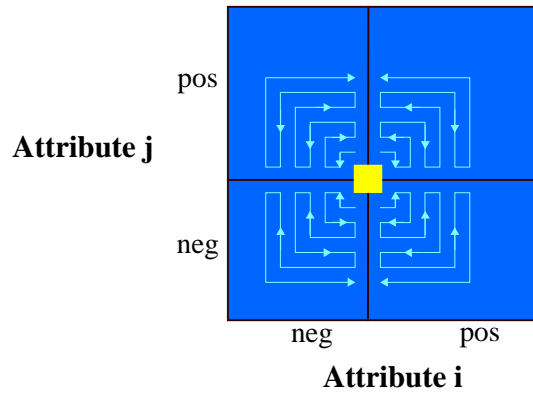
Spiral Technique   **Axes Technique**

Pixel-oriented Techniques
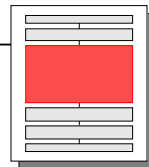
Spiral and Axes Techniques
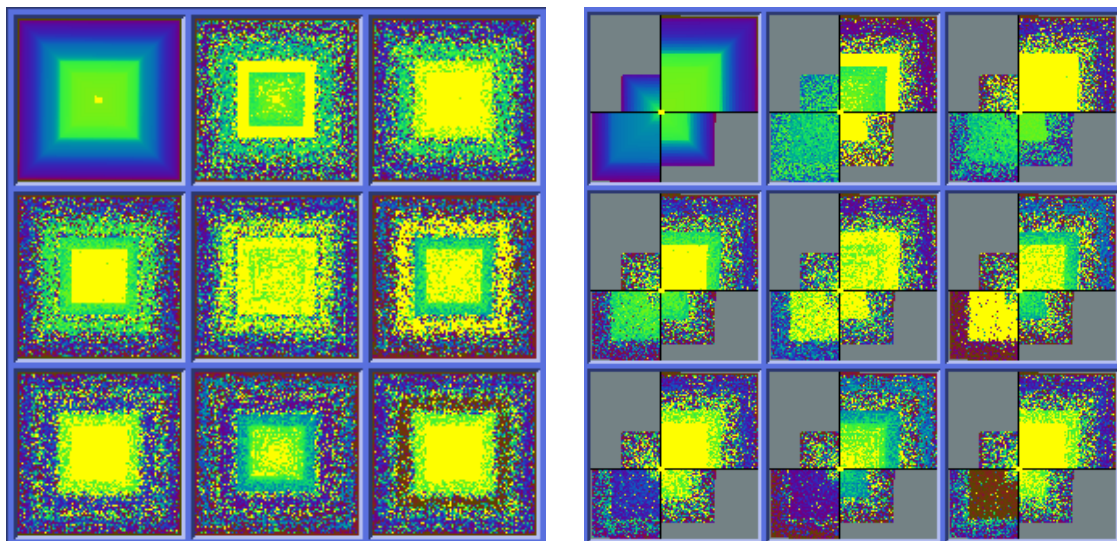
random data containing several clusters

Pixel-oriented Techniques
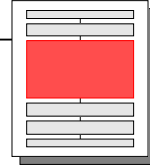
Spiral, Axes, and Grouping Techniques
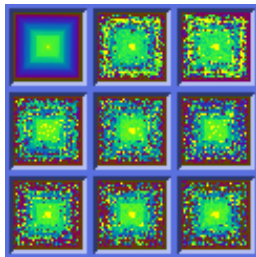
Spiral Technique
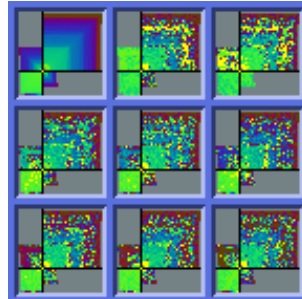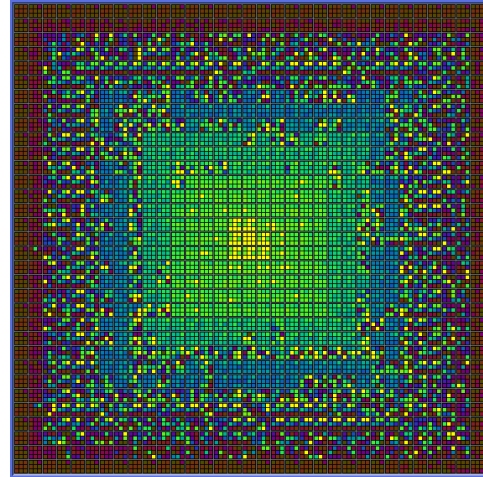
Axes Technique

Grouping Technique

Hierarchical Techniques
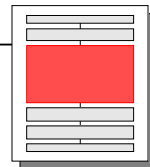
Basic Idea: Visualization of the data using a hierarchical partitioning
into two- or three-dimensional subspaces.

Overview

❑ **Dimensional Stacking [LWW 90]**

❑ **Hierarchical Plotting [MGTS 90]**

 (⇨ hierarchical drawing of an n-dimensional grid)

❑ **Worlds-within-Worlds [FB 90]**

❑ **Treemap [Shn 92, Joh 93]**

❑ **Cone Trees [RMC 91]**

 (⇨ 3D-Techniques)

Dimensional Stacking [LWW 90]
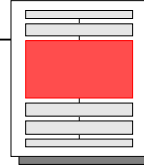
attribute 2

attribute 1

attribute 4
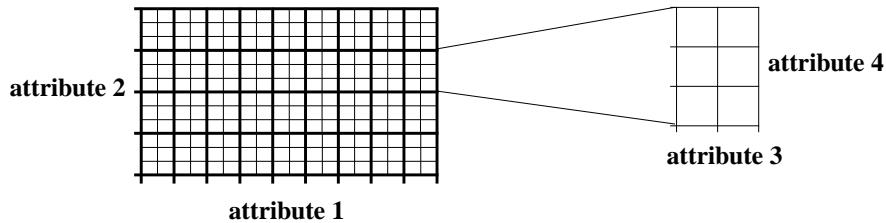
attribute 3

⇨ partitioning of the n-dimensional attribute space in 2-dimensional subspaces which are 'stacked' into each other

⇨ partitioning of the attribute value ranges into classes

⇨ the important attributes should be used on the outer levels

⇨ adequate especially for data with ordinal attributes of low cardinality
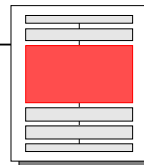
Worlds-within-Worlds [FB 90]
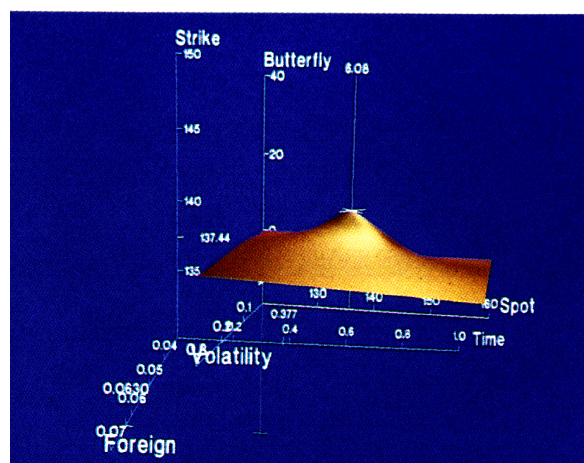
visualization of a six-dim. function

⇨ partitioning of the n-dim. space into 3-dim. subspaces (e.g., a six-dim. object is displayed by having a new coordinate system for the last three dimensions sit inside the coordinate system for the first three)

Hierarchical Techniques
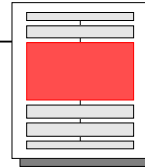
Treemap [Shn 92, Joh 93]

⇨ screen-filling method which uses a hierarchical partitioning of the screen into regions depending on the attribute values

⇨ the x- and y-dimension of the screen are partitioned alternately according to the attribute values (the attribute value ranges have to be partitioned into classes)

⇨ the attributes used for the partitioning and their ordering are user-defined (the most important attributes should be used first)

⇨ the color of the final regions may correspond to an additional attribute

⇨ suitable to get an overview over large amounts of hierarchical data (e.g., file system) and for data with multiple ordinal attributes (e.g., census data)

Hierarchical Techniques
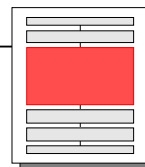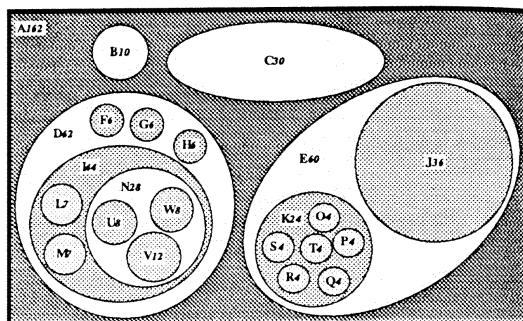
Treemap (cont'd)
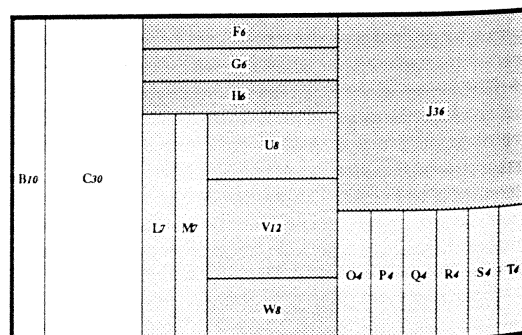
Venn Diagram   **Tree-Map**

Hierarchical Techniques

Treemap (cont'd)
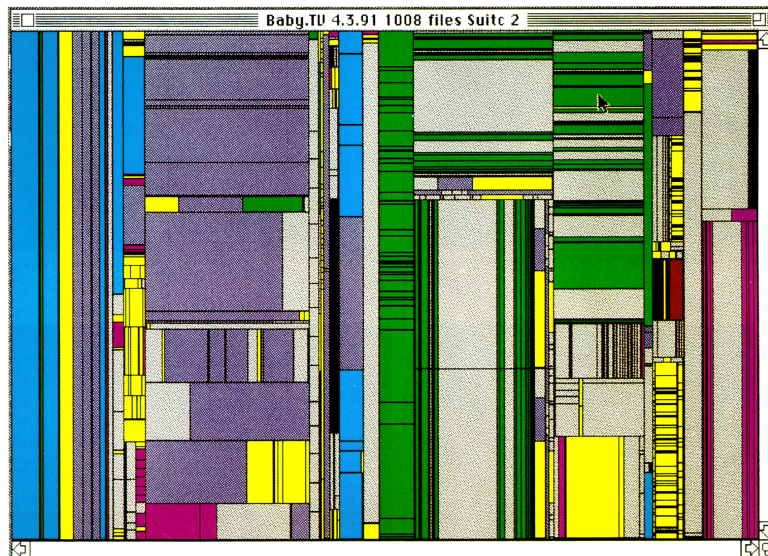
treemap of a
file system
containing about
1000 files

Graph-based Techniques
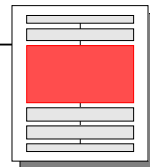
Basic Idea: Visualization of large graphs using techniques
to avoid clutter.

Overview

❑ **SeeNet [EW 93, BEW 95]**

❑ **Hygraphs [CM 93, Con 94]**

❑ **Fisheye Views [Fur 86, SB 94]**

❑ **3D-Graphs (Narcissus [HDWB 95])**
(⇨ 3D-Techniques)

Graph-based Techniques
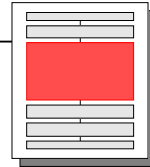
SeeNet [EW 93, BEW 95]

⇨ visualization of hierarchical networks with weighted links

⇨ special features of SeeNet:
- semantic node placement
  (minimizing the distance of nodes with high-weighted links)
- attributes are mapped to size and color of nodes and links
- interactivity for   - changing the mappings
  - expanding or collapsing nodes within the hierarchy
  - getting additional information, etc.

⇨ mappings in the example:
- size of nodes: number of e-mail messages of a person
- color of nodes: function of staff members
- size of links: number of e-mail messages of the link
- color of links: blue for few through green and yellow to red for many messages

---

Graph-based Techniques
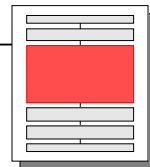
SeeNet (cont'd)
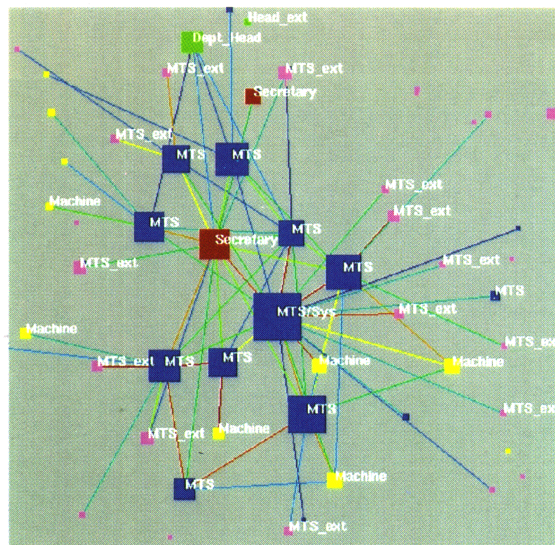
visualization of all
e-mail connections
in a department
over a period of time
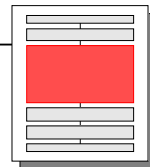
Graph-based Techniques
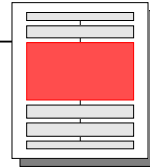
Hygraphs [CM 93, Con 94]

⇨ Hy$^+$ is an integrated query and visualization system

⇨ Hy$^+$ visualizes structural data as hygraphs

⇨ hygraphs are a generalization of several diagrammatic notations
(e.g., labelled graphs are a special case of hygraphs)

⇨ Hy$^+$ queries are described as expressions of the
GraphLog query language

⇨ Hy$^+$ queries may be used to define new relationships but
may also be used to reduce the data to be visualized

⇨ data may be presented at different levels of detail
(e.g., the user may interactively collapse nodes to avoid clutter)

Graph-based Techniques
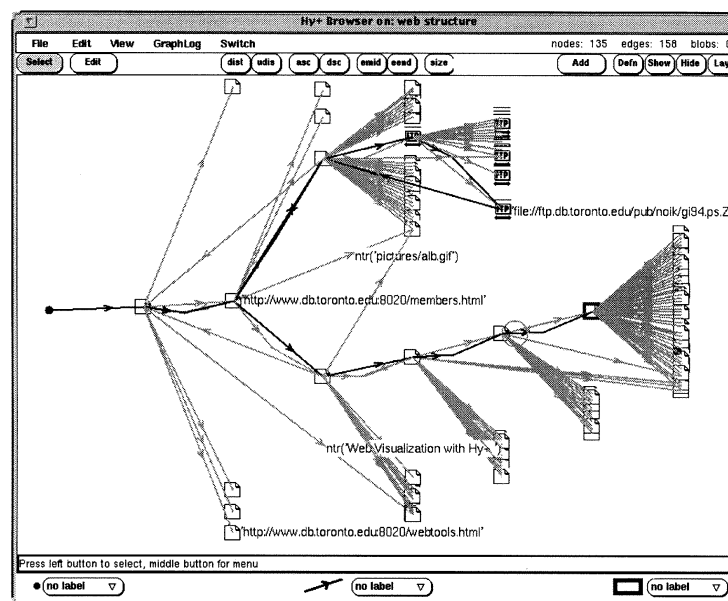
Hygraphs (cont'd)

visualization of a
web browsing session

Graph-based Techniques
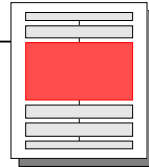
Hygraphs (cont'd)
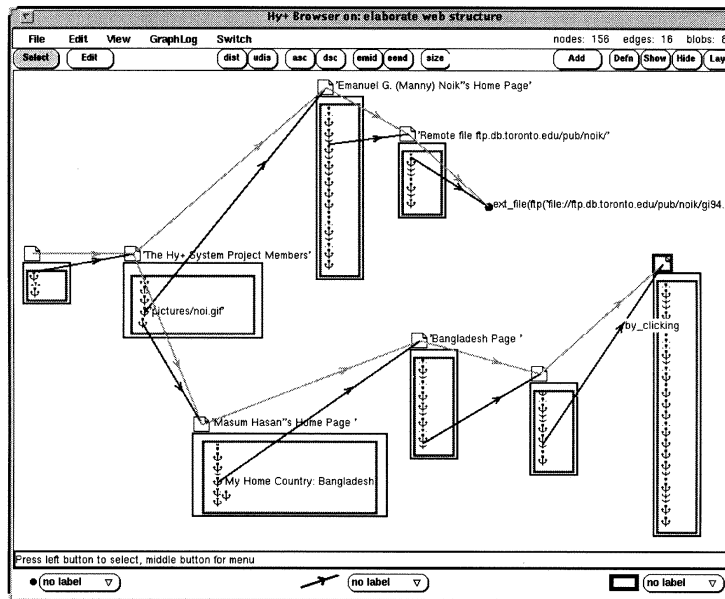
web browsing session
with collapsed nodes
(called blobs in Hy$^+$)

Graph-based Techniques

Fisheye View [Fur 86, SB 94]

original graph

fisheye view of the graph
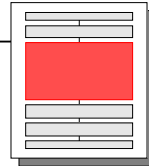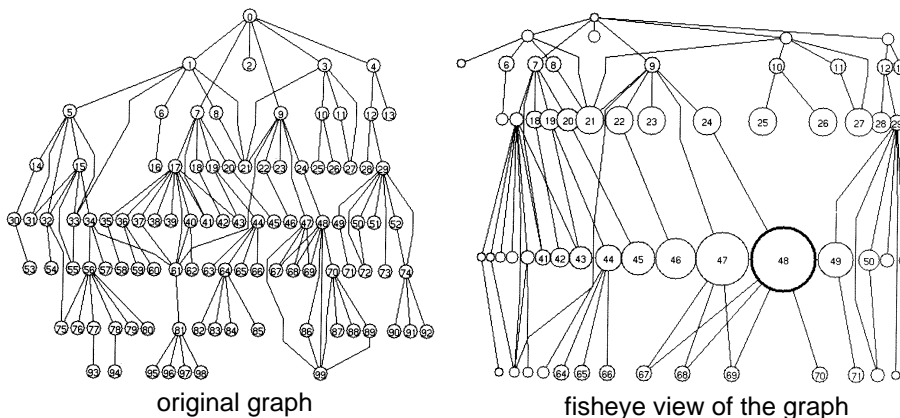
⇨ graph visualization using a fisheye perspective

⇨ shows an area of interest quite large and with detail and
the other areas successively smaller and in less detail

3D-Techniques
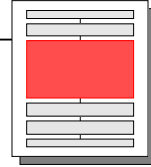
Basic Idea: Visualization of large data sets using 3D graphics.

Overview

❑ **Perspective Wall [MRC 91]**

❑ **Hyperbox [AC 91]**

❑ **Landscapes [Wri 95]**

❑ **Cone Trees [RMC 91]**

❑ **3D-Graphs (Narcissus [HDWB 95])**
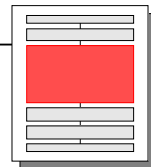
3D-Techniques

Perspective Wall [MRC 91]

⇨ presentation of the data on a perspective wall

⇨ the data outside the focal area are perspectively
    reduced in size

⇨ the perspective wall is similar to the fisheye view
    (⇨ Graph-based Techniques); it is a variant of the
    bifocal lens display [SA 82] which horizontally com-
    presses the sides of the workspace by direct scaling

3D-Techniques
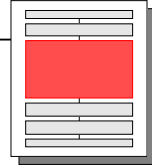
Perspective Wall (cont'd)
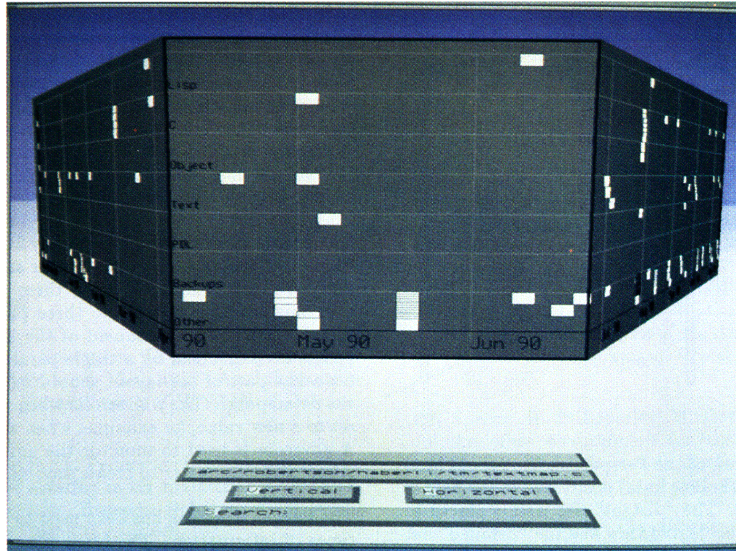
a number of documents arranged on a perspective wall

3D-Techniques
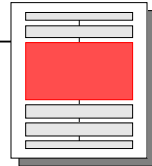
Hyperbox [AC 91]
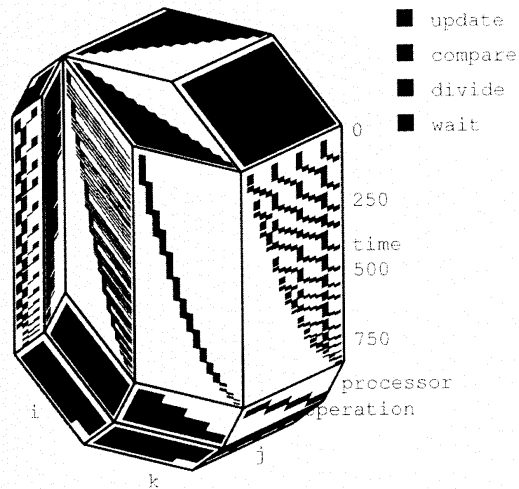
parallel processor performance data visualized as a hyperbox

⇨ mapping of scatterplots onto a hyperbox

3D-Techniques
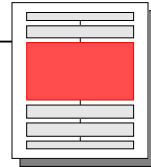
Landscapes [Wri 95]
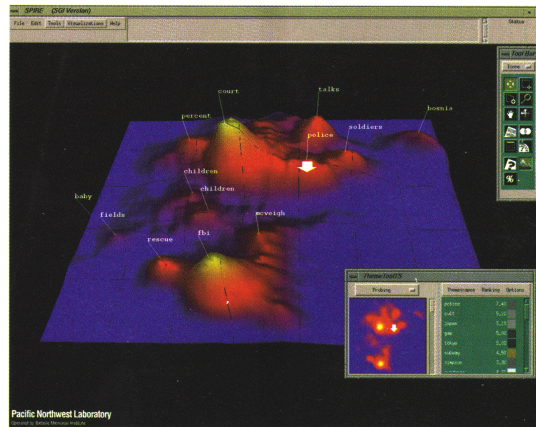
news articles
visualized as
a landscape

⇨ visualization of the data as landscapes

⇨ the data needs to be transformed into a spatial representation
which preserves the characteristics of the data

3D-Techniques
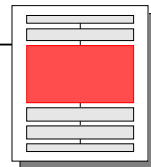
Cone Trees [RMC 91]
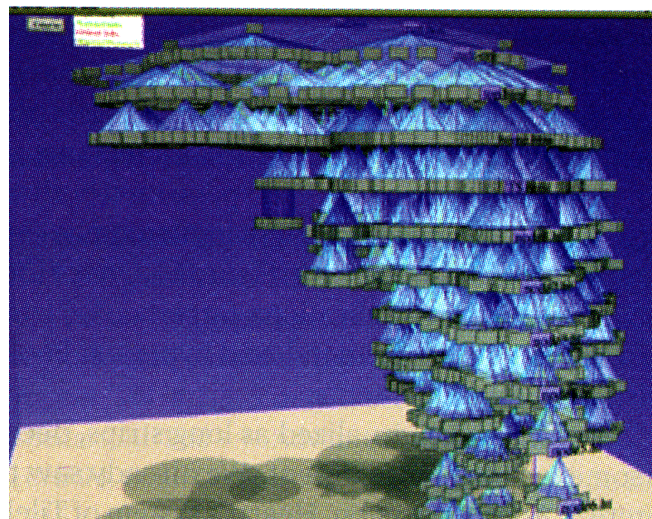
20.000 web pages
visualized as a
cone tree

⇨ animated 3D visualizations of hierarchical data

3D-Techniques
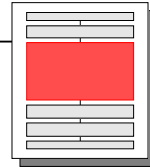
3D-Graphs (Narcissus [HDWB 95])
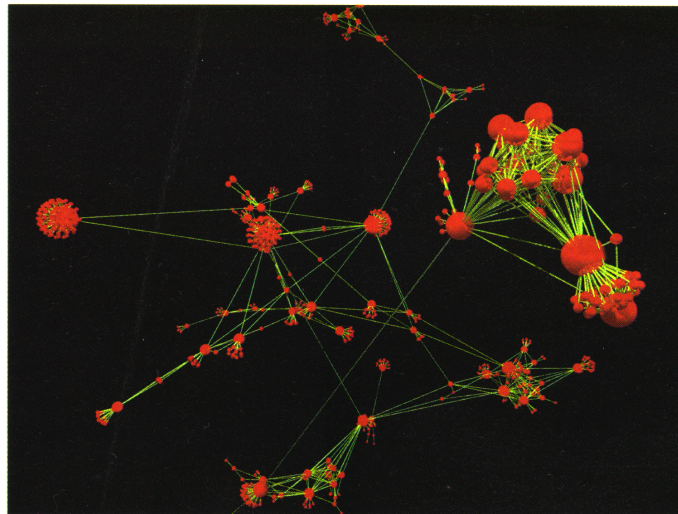
visualization of
a large number
of web pages

⇨ visualization of complex highly interconnected data (e.g., graphs such as the web)

Dynamic Techniques
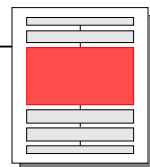
Basic Idea: Dynamic interaction with the visualization of the data for a more effective exploration of the data.

Overview

❑ **Dynamic Projections**

❑ **Dynamic Linking & Brushing**

❑ **Dynamic Environment**

❑ **Dynamic Zooming**

❑ **Dynamic Detail on Demand**

❑ **Dynamic Data-to-Visualization Mapping**

Dynamic Techniques
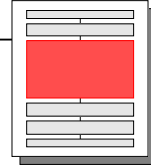
❑ **Dynamic Projections**

⇨ interactive or automatic variation of the projections

⇨ visualization of the remaining parameters in 2D or 3D

⇨ automatic variation results in an animation of the data

⇨ examples: • GrandTour [Asi 85]
- S Plus [BCW 88]
- XGobi [SCB 92, BCS 96]

Dynamic Techniques
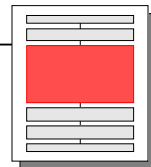
❑ **Dynamic Linking & Brushing**

⇨ prerequisite: multiple visualizations of the same data
(e.g., visualizations of different projections)

⇨ interactive changes made in one visualization are
automatically reflected in the other visualizations

⇨ examples: • S Plus [BCW 88]
- XGobi [SCB 92, BCS 96]
- DataDesk [Vel 92, WUT 95]

Dynamic Techniques
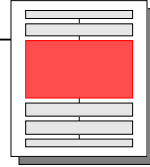
❑ **Dynamic Zooming**

⇨ visualization of large amounts of data in reduced form
  to provide an overview of the data

⇨ variable zooming of the data with automatic changes
  of the visualization modes to present more details

⇨ examples: • PAD++ [Bed 94]
  • IVEE [AW 95]

Dynamic Techniques
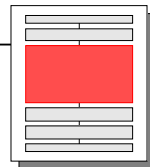
❑ **Dynamic Detail on Demand**

⇨ the possibility to interactively obtain more details
  of the visualized data

⇨ details are, for example, the attribute values corresponding
  to an icon or additional attribute values of a data item

⇨ examples: • Table Lens [RC 94, PR 96]
  • VisDB [KK 94, KK 95]
  • IVEE [AW 95]

Dynamic Techniques
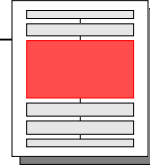
❑ **Dynamic Environment**

⇨ interactive movements in the visualization space
which causes the visualization to change accordingly

⇨ combination of a partial dynamic zooming with a
detail on demand strategy for the focal point

⇨ examples: • Fisheye Views [Fur 86, SB 94]
• Perspective Wall [MRC 91]
• Dyn. Landscapes [LRP 95]

Dynamic Techniques

❑ **Dynamic Data-to-Visualization Mapping**

⇨ dynamic mapping of the data attributes
to the parameters of the visualization

⇨ parameters of the visualization are
• x-, y-, and z-axes
• color and size of icons, links, etc.

⇨ examples: • AutoVisual [BF 93]
• IVEE [AW 95]
• SeeNet [EW 93]
• S Plus [BCW 88]
• XGobi [SCB 92, BCS 96]

Hybrid Techniques
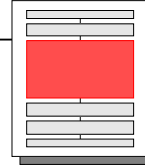
Basic Idea: Integrated use of multiple techniques in one or multiple windows to enhance the expressiveness of the visualizations.

⇨ linking diverse visualization techniques may provide additional information

⇨ virtually all visualization techniques are combined with dynamics & interactivity

Examples: IVEE [AW 95] uses *Starfield Displays* [AS 94] which are scatterplots of icons with dynamic zooming and mapping
(combination of geometric, icon-based, and dynamic techniques)

XmDv [War 94] allows to dynamically link and brush scatterplot matrices, star icons, parallel coordinates, and dimensional stacking
(combination of geometric, icon-based, hierarchical and dynamic techniques)

Comparison of the Techniques

Criteria for Comparison

comparison of the described data visualization techniques based on their suitability for certain

⇨ data characteristics
(e.g., no. of variates, no. of data items, categorical data, ...)

⇨ task characteristics
(e.g., clustering, multi variate hot spots, ...)

⇨ visualization characteristics
(e.g., visual overlap, learning curve, ...)

Note: The following comparison table expresses my personal opinion obtained from reading the literature and experimenting with several of the described techniques. Many of the ratings are arguable and largely depend on the considered data, the exploration task, experience of the user, etc. In addition, implementations of the techniques in real systems usually avoid the drawbacks of a single technique by combining it with other techniques, which is also not reflected in the ratings.

Comparison of the Techniques
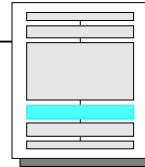
Comparison: A First Attempt

		cluster-ing	multi-variate hot spot	no. of variates	no. of data items	cate-gorical data	visual overlap	learning curve
Geometric Techniques	Scatterplot Matrices	++	++	+	+	-	o	++
	Prosection Views	++	++	+	+	-	o	+
	Hyperslice	+	+	+	+	-	o	o
	Parallel Coordinates	o	++	++	-	o	--	o
Icon-based Techniques	Stick Figure	o	o	+	-	-	-	o
	Shape Coding	o	-	++	+	-	+	-
	Grouping	o	-	++	+	-	+	-
Pixel-oriented Techniques	Query-Independent	+	+	++	++	-	++	+
	Query-Dependent	+	+	++	++	-	++	-
Hierarchical Techniques	Dimensional Stacking	+	+	o	o	++	o	o
	Worlds-within-Worlds	o	o	o	+	o	o	o
	Treemap	+	o	+	o	++	+	o
Graph-based Techniques	SeeNet	+	+	o	+	+	+	+
	Hygraphs	+	+	-	+	+	+	+
	Fisheye Views	+	+	-	+	+	++	+
3D-Techniques	Perspective Wall	+	+	o	+	+	++	+
	Hyperbox	+	+	+	+	o	o	o
	Landscapes	+	+	-	o	o	+	+
	Cone Trees	+	+	o	+	o	+	+
	3D-Graphs (Narcissus)	+	+	-	+	o	+	+

Database Visualization Systems
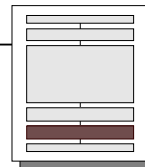
Overview

❑ **Statistics-oriented Systems**

❑ **Visualization-oriented Systems**

❑ **Database-oriented Systems**

❑ **Special Purpose Systems**

Database Visualization Systems
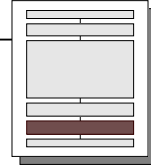
❑ **Statistics-oriented Systems**

⇨ S Plus [BCW 88]

(-> generic system for statistical analysis and visualization)

⇨ XGobi [SCB 92, BCS 96]

(-> extensible lisp-based system for statistical analysis and visualization)

⇨ Data Desk [Vel 92, WUT 95]

(-> commercial system for statistical analysis and visualization;
features: dyn. linking & brushing of scatterplots and histograms)

⇨ Diamond (SPSS)

(-> commercial system for statistical analysis and visualization;
features: dyn. linking & brushing of scatterplots, parallel coordinates, etc.)

⇨ DataSpace [ADLP 95]

(-> 3D-arrangement of a large number of arbitrary visualizations)

Database Visualization Systems
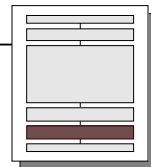
❑ **Visualization-oriented Systems**

⇨ ExVis [GPW 89]

(-> features: stick figure and other icon-based techniques)

⇨ Parallel Visual Explorer (IBM)

(-> features: parallel coordinate technique with query-based coloring, etc.)

⇨ XmDv [War 94, MW 95]

(-> features: scatterplot matrices, star icons, parallel coordinates,
dimensional stacking, dynamic linking and brushing)

⇨ Influence & Attribute Explorer [STDS 95, SDTS 95]

(-> features: scatterplot and prosection matrices, histograms,
dynamic linking and brushing)

⇨ Information Visualizer (Xerox)

(-> features: diverse information visualization techniques including
perspective wall, table lens, cone trees)

Database Visualization Systems
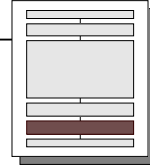
❑ **Database-oriented Systems**

⇨ Hy$^+$ [CM 93]
(-> features: query and visualizations of hygraphs)

⇨ TreeViz [Joh 93]
(-> features: treemap technique)

⇨ Table Lens [RC 94, PR 96]
(-> features: graphical display of tabular data with detail-on-demand capabilities)

⇨ VisDB [KK 94, KK 95]
(-> system for interactive slider-based exploration of very large databases
features: stick figure, parallel coordinate, and pixel-oriented techniques)

⇨ IVEE [AW 95a,b]
(-> commercial system for database exploration;
features: generic interactive slider-based visualization environment)

Database Visualization Systems
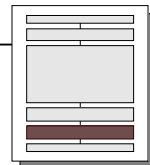
❑ **Special Purpose Systems**

⇨ Margritte [Vas 94]
(-> graph visualization for CAD and Software Engineering databases)

⇨ Narcissus [HDWB 95]
(-> system for visualizing complex highly interconnected data such as the web)

⇨ WebBook and WebForager [CRY 96]
(-> systems for accessing and exploring the web)

⇨ SeeNet [EW 93] (-> system for visualizing networks with weighted links)

⇨ SeeSoft [ESS 92] (-> system for visualizing structured text such as software)

⇨ Information Retrieval interfaces such as VIBE [Ols 93]
(a listing of Information Retrieval Interfaces can be found under
"http://www-cui.darmstadt.gmd.de/visit/Activities/Viri/visual.html")

⇨ ...

Summary and Conclusions
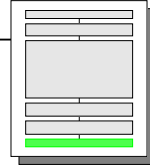
Summary

- there are a number of recently developed visualization techniques which are applicable to database exploration

- there are different techniques for different types of data (relational tables, hierarchies, graphs, etc.)

- many of the techniques also applicable to traditional relational databases

- there are a number of research prototypes and commercial systems which are available

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Summary and Conclusions
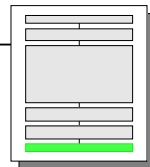
Research Issues

- new applications of visualization technology in the database area such as
  Visual Database Exploration for Knowledge Discovery

- development of integrated database visualization and exploration systems

- evaluation and comparison of visualization techniques for database exploration (-> possibilities for improvement)

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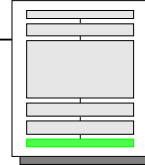
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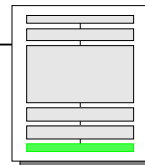
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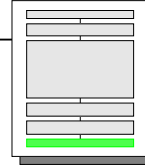
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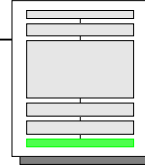
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