

# Vorlesung Advanced Topics in HCI (Mensch-Maschine-Interaktion 2)

Ludwig-Maximilians-Universität München  
LFE Medieninformatik  
Andreas Butz & Albrecht Schmidt  
SS2005

<http://www.medien.informatik.uni-muenchen.de/>

## Chapter 2: Information Visualization

### Table of Content

- Information & representation
- What is information visualization
- Perception basics
- Standard techniques
- Principles and Taxonomy
- Options for visualization & Examples

*“Graphical excellence is that which gives to the viewer the greatest number of ideas in the shortest time with the least ink in the smallest space.”*

-- Edward R. Tufte

## Representation

- What is a good Representation?
  - Capture and present the essential
  - Deliberately hide irrelevant parts
  - Appropriate for the recipient and his/her abilities
  - To understand and interpret by the recipient
  - Appropriate for the task
  
- “Solving a problem simply means representing it so as to make the solution transparent” (Simon, 1981)
  
- Allow people to look at the presentation and draw the “right” conclusions!

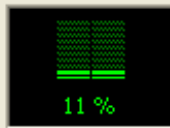
# Representations

## Physikalischer Speicher (KB)

Insgesamt	514544
Verfügbar	177396
Systemcache	204792

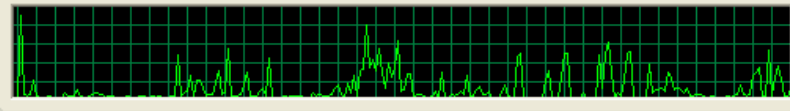
- Figures / numbers

## CPU-Auslastung



- Numbers in bar graph

## Verlauf der CPU-Auslastung



- Plot with history

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

5

# How to read representations

- Read the plain facts
- Compare representations (visual calculations)
- Identify patterns
- Make interpretations
  
- Can be enhanced by active diagrams
  - Allow interactive manipulation

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

6

# External aids for thinking

*The power of the unaided mind is highly overrated. Without external aids, memory, thought, and reasoning are all constrained. But human intelligence is highly flexible and adaptive, superb at inventing procedures and objects that overcome its own limits. The real powers come from devising external aids that enhance cognitive abilities. How have we increased memory, thought, and reasoning? By the inventions of external aids: It is things that make us smart. (Norman, 1993)*

- External cognition
  - Internal and external representation and processing weave together in thought
- External cognitive aids can enhance cognition
- An important class of external cognitive aids that make us smart are graphical inventions
  - Charts for navigation
  - Diagrams

# Use of visual representations

- Pictures and diagrams are used to communicate existing ideas and thoughts
- Graphical representations can help in developing and formulating ideas and thoughts
- Using visual representations “to think”

## Information – to visualize

- What is “Information”?
  - Entities, concepts, things, items that may not have a direct physical correspondence
  - Information is often abstract
  
- Large sets of data and information
  - Great amount of data
  - Information is generated in many processes
  
- visualize: to form a mental image or vision of ...
- visualize: to imagine or remember as if actually seeing.  
(American Heritage dictionary, Concise Oxford dictionary)

## What is Information Visualization

- The use of computer-supported, interactive visual representations of data to amplify cognition. (Card, Mackinlay, Shneiderman '98)
  
- “Transformation of the symbolic into the geometric”  
(McCormick et al., 1987)
  
- “... augmenting ... natural intelligence in the best possible way, ... finding the artificial memory that best supports our natural means of perception.” (Bertin, 1983)
  
- “The depiction of information using spatial or graphical representations, to facilitate comparison, pattern recognition, change detection, and other cognitive skills that make use of the visual system.” (Hearst, 2003, CHI-Tutorial)

# Information Visualization

- The basic idea
  - Finding for information items an appropriate and meaningful mapping into a 2-D or 3-D physical space.
  - Creating a visual representation that helps to understand data and is useful for analysis and decision-making
- Visual representation are helpful
  - External cognition
  - frame of reference
  - “temp storage” for thinking
- “The purpose of visualization is insight, not pictures”
  - Insight – understanding, discovery, decision making, explanation

## Definition by Shneiderman



- Compact graphical presentation and
  - user interface for
  - manipulating large numbers of items (102 - 106),
  - possibly extracted from far larger datasets.
- Enables users to make
  - discoveries,
  - decisions, or
  - explanations
- about
  - patterns (trend, cluster, gap, outlier...),
  - groups of items, or
  - individual items.

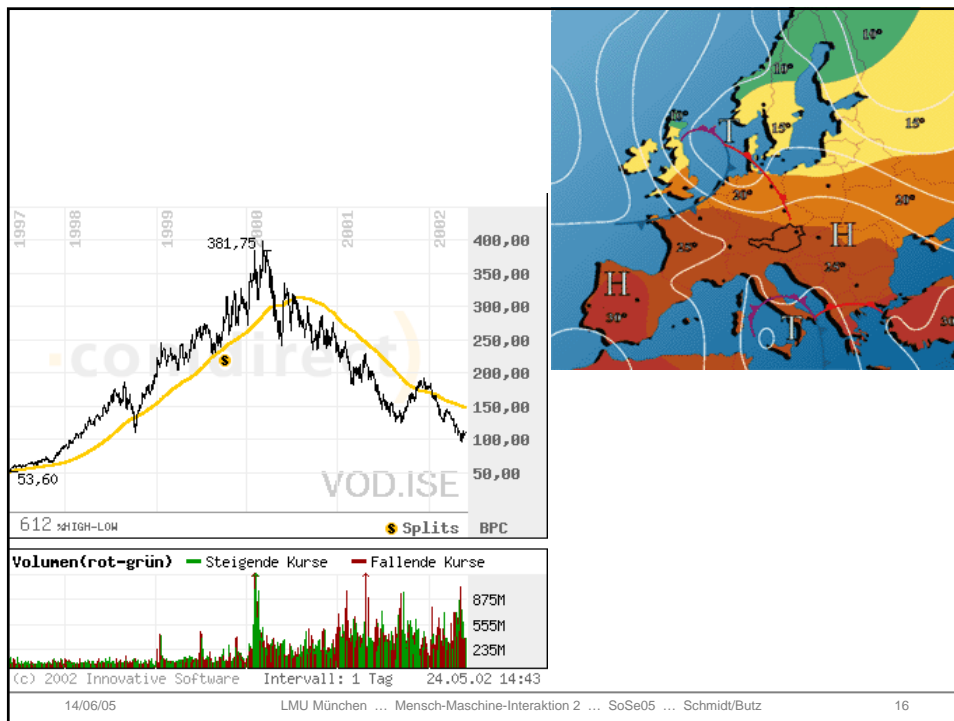
## InfoVis is applicable to:

- Text, documents, text archives
- Databases
- Statistics
- Financial data, business data
- Geographic data
- Network information, internet information
- Software
- ...

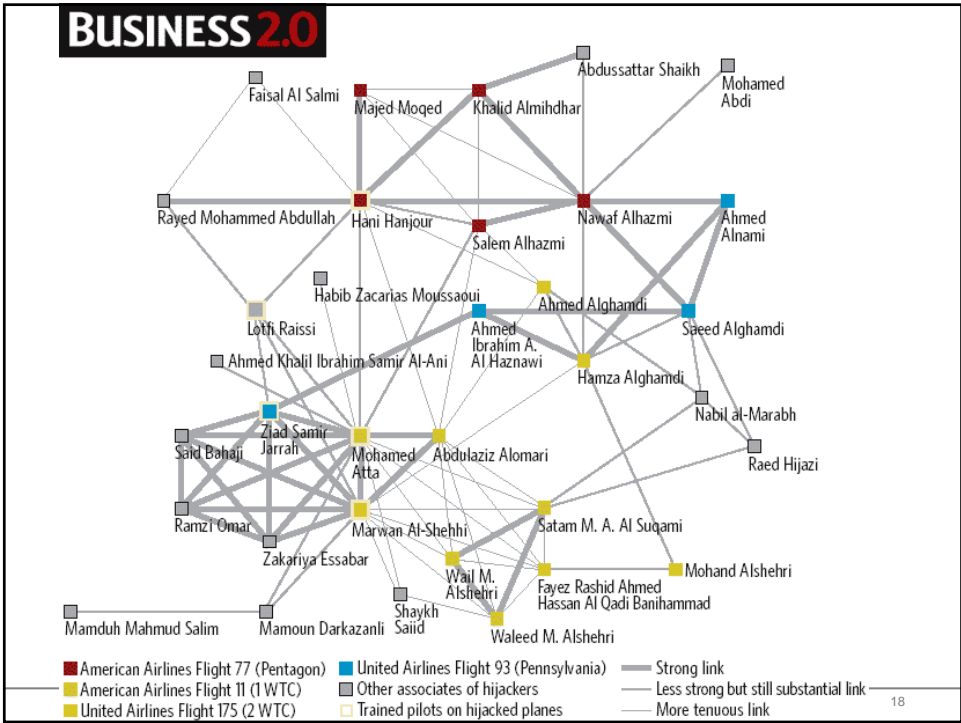
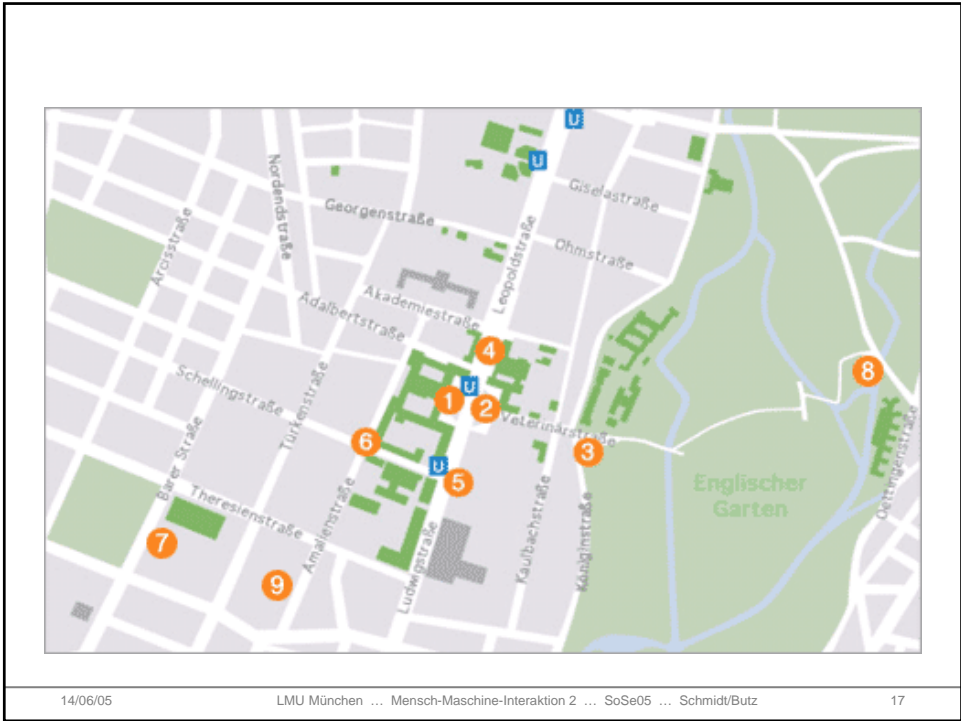
## What tasks are supported by Information Visualization?

- Search
  - Finding a specific information in a data set
- Browse
  - survey, inspect, look for interesting information
- Analysis
  - Comparison-Difference, find outliers and extremes, spot patterns
- Many more...
  - Categorize, Associate
  - Locate, Rank
  - Identify, Reveal
  - Monitor, Maintain awareness

# Examples







**vizster**  
 visualizing online social networks  
 jeffrey heer + danah boyd

**publications**  
 research paper  
 video demonstration  
 early design report

**photo gallery**  
 egocentric  
 community  
 linkage  
 search  
 x-ray 1  
 x-ray 2

**Vizster**  
 [jeffrey heer + danah boyd]

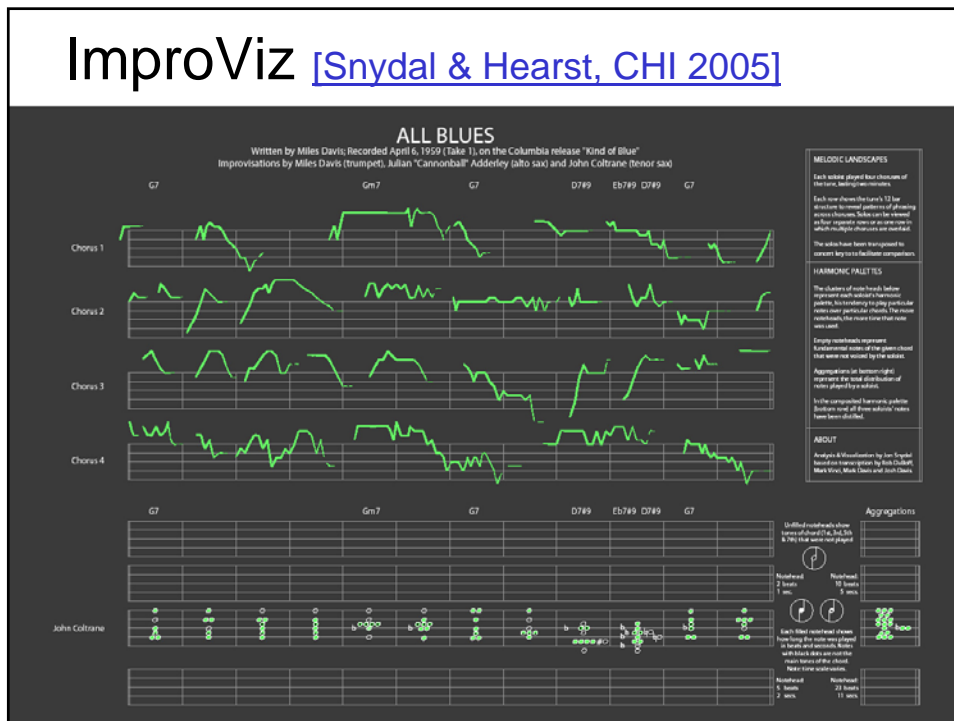
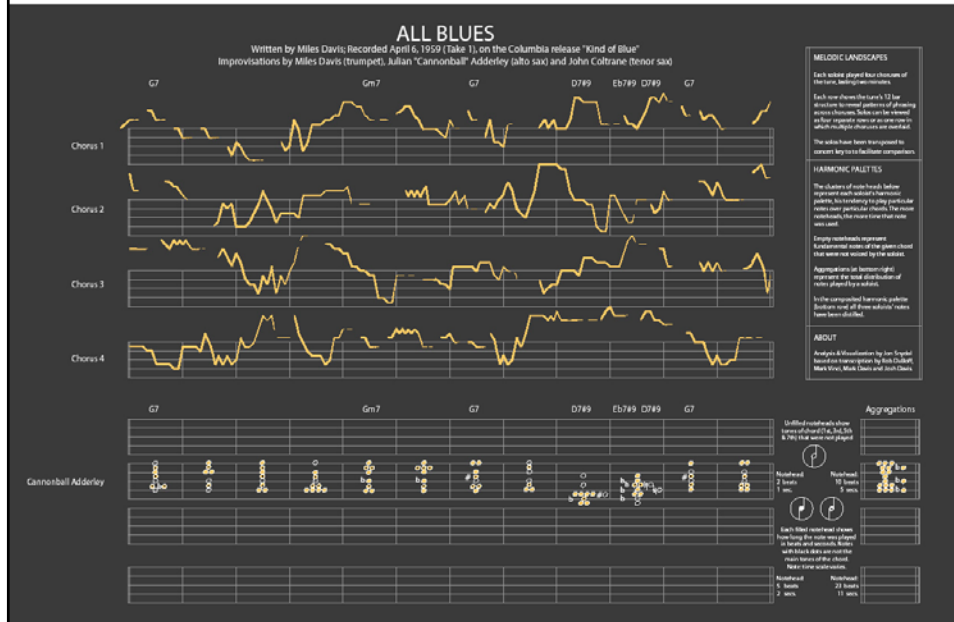
[Video](#)

[http://www.peets.com/selector\\_coffee/coffee\\_selector.asp](http://www.peets.com/selector_coffee/coffee_selector.asp)

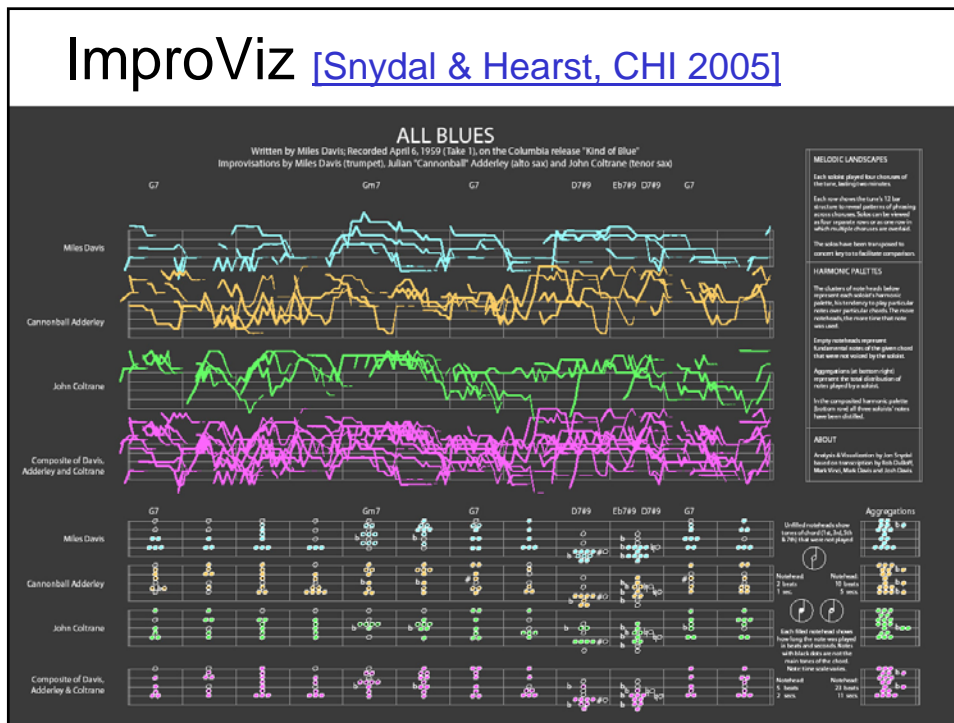
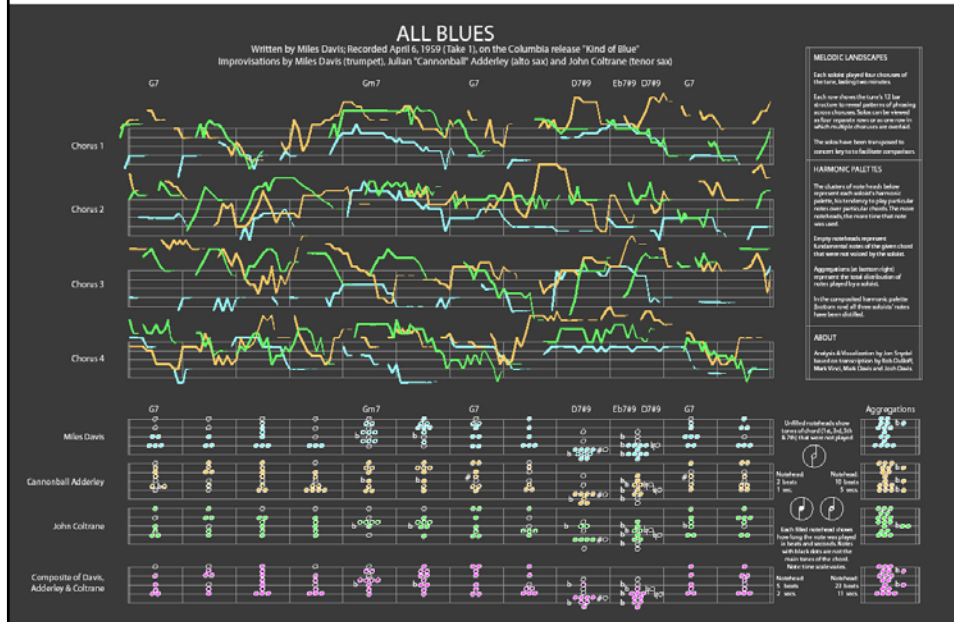
14/06/05      LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz      20



# ImprovViz [\[Snydal & Hearst, CHI 2005\]](#)



# ImprovViz [Snydal & Hearst, CHI 2005]



## Goal of Information Visualization

- Use human perceptual capabilities to gain insights into large data sets that are difficult to extract using standard query languages
- Exploratory Visualization, look for
  - Structure
  - Patterns
  - Trends
  - Anomalies
  - relationships
- Provide a qualitative overview of large, complex data sets
- Help to find regions of interest and appropriate parameters for more focused quantitative analysis

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

27

## Knowledge crystallization



(Storey, 2004)

- Knowledge crystallization involves getting insight about data relative to some task
- Steps required in a Knowledge Crystallization task:
  - Information foraging/browsing (from repositories, people...)
  - Search for/build a schema (representation) –need to know what to include/omit
  - Instantiate schema with data
  - Problem solve to trade-off features
  - May have to search for a new schema..
  - Package the patterns found in some output product (i.e. a concise briefing of results)
- A visualization tool has to support or automate some of these steps, it is a cognitive aid during our process of schematization
- So we need data, a task and a schema

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

28

# Example – Air fare (1)

**Boston, MA, United States (BOS-Logan Intl.) to Victoria, BC (YYJ-Victoria)**

**Change your search**

Departure airport: BOS (Boston)  
 Destination airport: YYJ (Victoria)  
 Departing: (DDMMYY) 6/11/2001  
 Morning  
 Returning: (DDMMYY) 13/11/2001  
 Evening  
 Airline: All Airlines  
 Nonstop flights only  
[Start again](#) [Go](#)

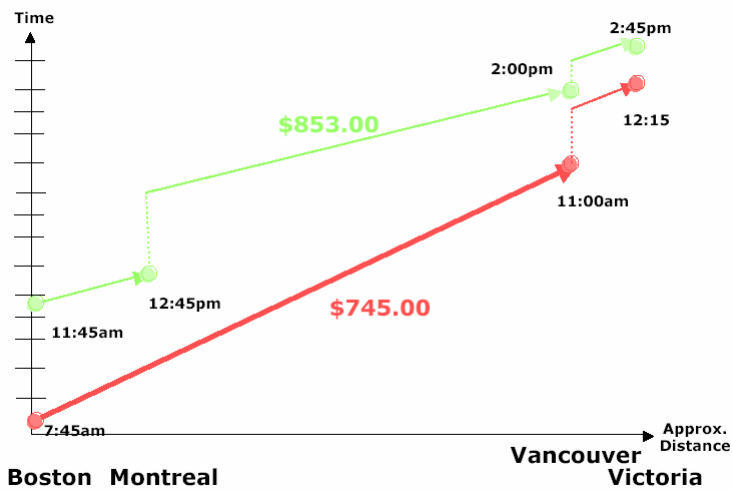
**To build your own trip, first pick any departing flight**

Sort by:  lowest price  shortest flights  departure time  arrival time

from C\$753	from C\$753	from C\$753	from C\$753	from C\$753	from C\$753	from C\$753	from C\$753	from C\$753	from C\$753
7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM	7:45 AM Depart Boston (BOS) Arrive Victoria (YYJ) 12:35 PM
Tue 6-Nov 7hr 50min	Tue 6-Nov 8hr 50min	Tue 6-Nov 8hr 50min	Tue 6-Nov 9hr 20min	Tue 6-Nov 9hr 50min	Tue 6-Nov 9hr 50min	Tue 6-Nov 9hr 50min	Tue 6-Nov 9hr 50min	Tue 6-Nov 9hr 50min	Tue 6-Nov 10hr 20min
Air Canada 763 / 8357 Connect in Vancouver (YVR)	Air Canada 763 / 1613 Connect in Vancouver (YVR)	Air Canada 763 / 1613 Connect in Vancouver (YVR)	Air Canada 763 / 1613 Connect in Montreal (YUL), Vancouver (YVR)	Air Canada 307 / 129 / 1635 Connect in Montreal (YUL), Vancouver (YVR)	Air Canada 801 / 133 / 1613 Connect in Toronto (YYZ), Vancouver (YVR)	Air Canada 306 / 3363 Connect in Toronto (YYZ)	Air Canada 763 / 1652 Connect in Vancouver (YVR)	Air Canada 763 / 1619 Connect in Vancouver (YVR)	Air Canada 763 / 1619 Connect in Vancouver (YVR)

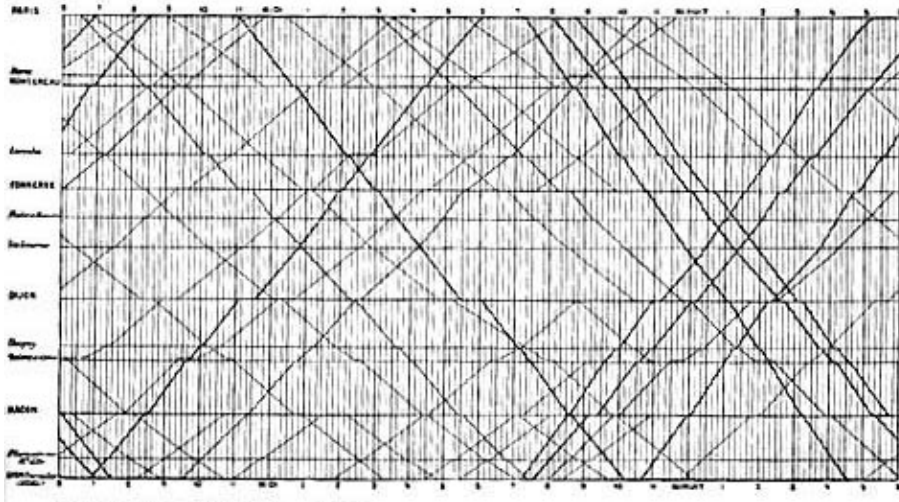
(Storey, 2004)

# Example – Air fare (2)



(Storey, 2004)

# 1885 French train schedule by E.J. Marey



# Knowledge crystallization (2)

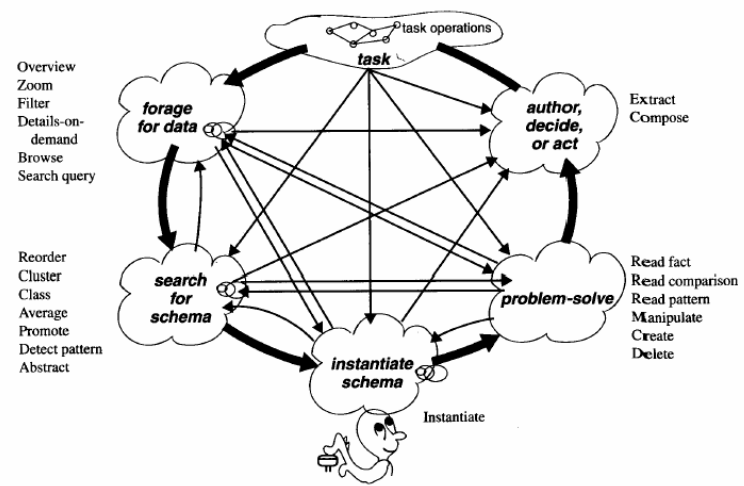


FIGURE 1.15 Knowledge crystallization. (Storey, 2004)



# Mapping Problem

- A lot of information does not imply any obvious spatial mapping!
- Basic Question:  
How to map non-spatial abstractions into effective visual representation?
- Approach:  
Use interactive techniques and visual representations to augment or amplify the user's cognition

# How Information Visualization can Amplify Cognition

Different ways in which visualizations *could* help amplify cognition:

1. By increasing memory and processing resources available
  - Parallel perceptual processing
  - Offload work from cognitive to perceptual system
2. By reducing the amount of time to search
  - High data density
  - Greater access speed
3. Enhancing the detections of patterns and enabling perceptual inference operations
  - Abstraction and Aggregation
4. Aid perceptual monitoring
  - Color or motion coding to create pop out effect
5. By encoding information in an Interactive Medium

# Information Visualization

## Basic Key Principles

- Abstraction
- Overview → Zoom+Filter → Details-on-demand
- Direct Manipulation
- Dynamic Queries
- Immediate Feedback
- Linked Displays
- Linking + Brushing
- Provide Focus + Context
- Animate Transitions and Change of Focus
- Output is Input
- Increase Information Density

# Human Perception & Visual Properties

## Human Perception & Visual Properties

- Preattentive Processing
- Accuracy of Interpretation of Visual Properties
- Illusions and the Relation to Graphical Integrity

[All Preattentive Processing figures from Healey 97](http://www.csc.ncsu.edu/faculty/healey/PP/PP.html)  
<http://www.csc.ncsu.edu/faculty/healey/PP/PP.html>

## User's Expectations from the physical world

- Well-Defined Surfaces  
**Objects have mostly smooth surfaces**
- Temporal Persistence  
**Objects don't randomly appear/vanish**
- Light travels in Straight Lines  
**reflects off surfaces in certain ways**
- Law of Gravity



Marti Hearst, 2003

# Perception & Representation

- Sensory Representations
  - Tap into Perceptual Power of Brain Without Learning
- Sensory Representations are effective
  - because well matched to early stages of neural processing

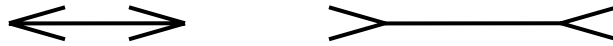
# Visual Illusions

- People don't perceive length, area, angle, brightness the way they "should".
- Some illusions have been reclassified as systematic perceptual errors
  - e.g., brightness contrasts (grey square on white background vs. on black background)
  - partly due to increase in our understanding of the relevant parts of the visual system
- Nevertheless, the visual system does some really unexpected things.

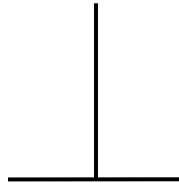
Hearst, 2003

## Illusions of Linear Extent

- Mueller-Lyon (off by 25-30%)



- Horizontal-Vertical



Hearst, 2003

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

41

## Preattentive Processing

- A limited set of visual properties are processed preattentively
  - (without need for focusing attention).
- This is important for design of visualizations
  - what can be perceived immediately
  - what properties are good discriminators
  - what can mislead viewers

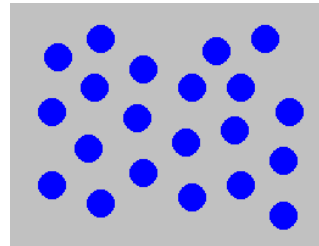
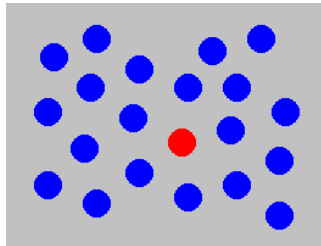
Hearst, 2003

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

42

## Example: Color Selection



Viewer can rapidly and accurately determine whether the target (red circle) is present or absent. Difference detected in color.

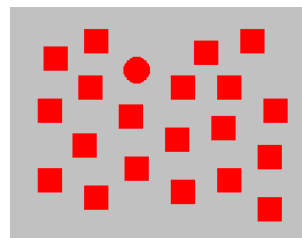
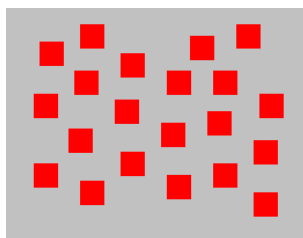
Hearst, 2003

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

43

## Example: Shape Selection



Viewer can rapidly and accurately determine whether the target (red circle) is present or absent. Difference detected in form (curvature)

Hearst, 2003

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

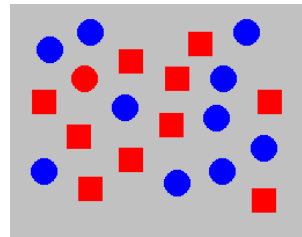
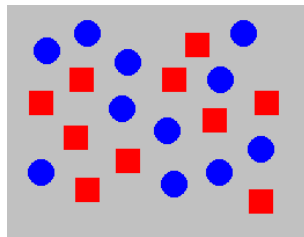
44

## Pre-attentive Processing

- < 200 - 250ms qualifies as pre-attentive
  - eye movements take at least 200ms
  - yet certain processing can be done very quickly, implying low-level processing in parallel
- If a decision takes a fixed amount of time regardless of the number of distractors, it is considered to be preattentive.

Hearst, 2003

## Example: Conjunction of Features



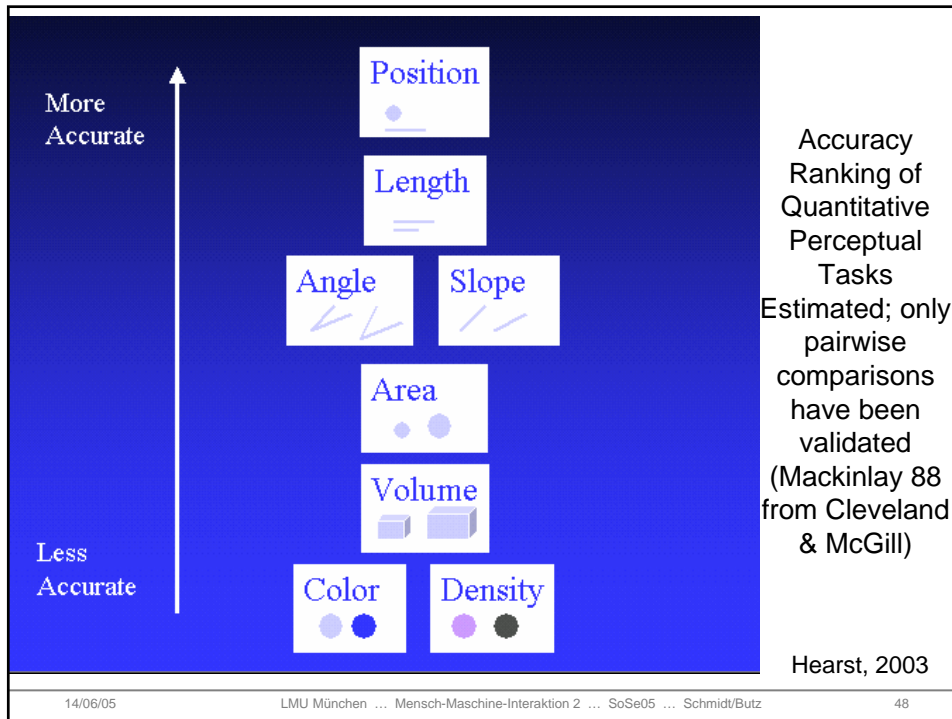
Viewer *cannot* rapidly and accurately determine whether the target (red circle) is present or absent when target has two or more features, each of which are present in the distractors. Viewer must search sequentially.

[All Preattentive Processing figures from Healey 97](http://www.csc.ncsu.edu/faculty/healey/PP/PP.html)  
<http://www.csc.ncsu.edu/faculty/healey/PP/PP.html>

# Preattentive Visual Properties (Healey 97)

length	Triesman & Gormican [1988]
width	Julesz [1985]
size	Triesman & Gelade [1980]
curvature	Triesman & Gormican [1988]
number	Julesz [1985]; Trick & Pylyshyn [1994]
terminators	Julesz & Bergen [1983]
intersection	Julesz & Bergen [1983]
closure	Enns [1986]; Triesman & Souther [1985]
colour (hue)	Nagy & Sanchez [1990, 1992]; D'Zmura [1991] Kawai et al. [1995]; Bauer et al. [1996]
intensity	Beck et al. [1983]; Triesman & Gormican [1988]
flicker	Julesz [1971]
direction of motion	Nakayama & Silverman [1986]; Driver & McLeod [1992]
binocular lustre	Wolfe & Franzel [1988]
stereoscopic depth	Nakayama & Silverman [1986]
3-D depth cues	Enns [1990]
lighting direction	Enns [1990]

Hearst, 2003





## Preattentive Processing (Pop Out)

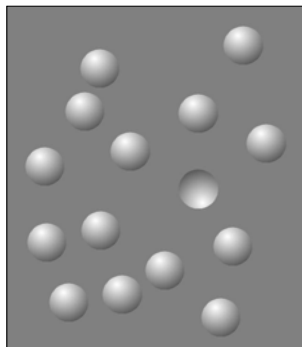
- Time required to find target independent of number of overall number
- Form:
  - line orientation, length, width
  - spatial orientation, added marks, numerosity (4)
- Colour:
  - hue, intensity
- Motion:
  - flicker, direction of motion
- Spatial Position:
  - stereoscopic depth, convex/concave shape, shadows

14/06/05

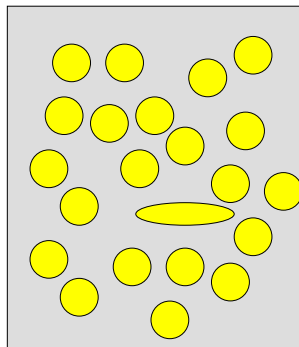
LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

49

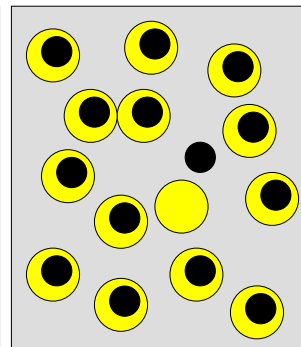
## Examples (pop-out)



Shading



Shape



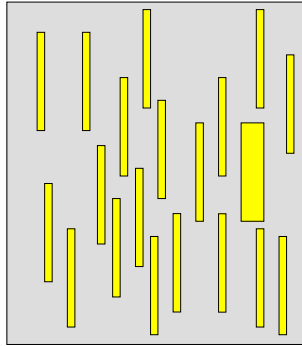
Enclosure

14/06/05

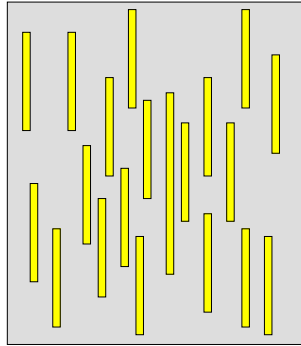
LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

50

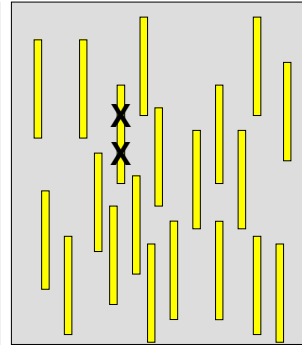
## Examples (pop-out)



width



length



marked

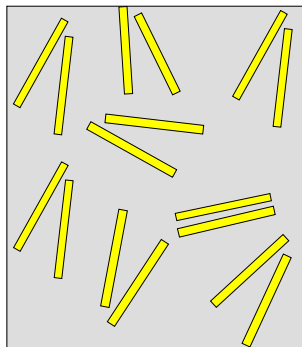
Hiding features  
due to placement

14/06/05

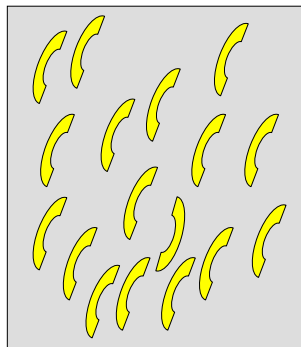
LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

51

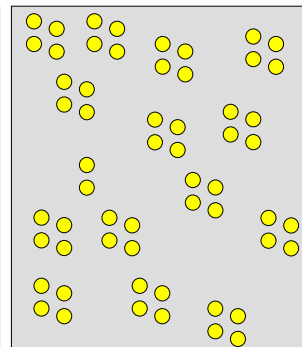
## Examples (pop-out)



angle



curve



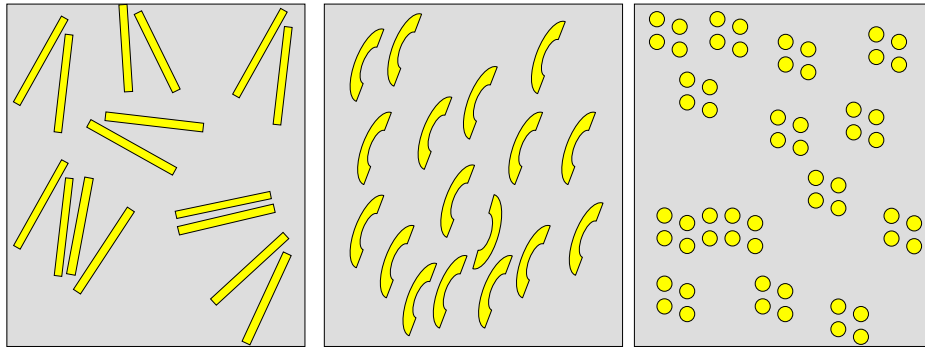
Clusters/count

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

52

## Examples (pop-out)



angle

curve

Clusters/count

Hiding features  
due to placement

Hiding features  
due to placement

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

53

## References

- Marti Hearst
  - <http://bailando.sims.berkeley.edu/infovis.html>
  - <http://bailando.sims.berkeley.edu/talks/chi03-tutorial.ppt>
- Margret-Anne Storey
  - <http://www.csr.uvic.ca/~mstorey/>
  - [http://www.cs.uvic.ca/~mstorey/teaching/infovis/course\\_notes/introduction.pdf](http://www.cs.uvic.ca/~mstorey/teaching/infovis/course_notes/introduction.pdf)
- Ben Shneiderman
  - <http://www.cs.ubc.ca/~tmm/courses/cpsc533c-03-spr/readings/shneiderman96eyes.pdf>

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

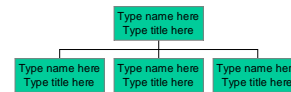
54

# Basic Types of Symbolic Displays (Kosslyn 89)

- Graphs



- Charts



- Maps



- Diagrams



From Hearst, 2003

# Basic Types of Data

- Nominal (qualitative)
  - (no inherent order)
  - city names, types of diseases, ...
- Ordinal (qualitative)
  - (ordered, but not at measurable intervals)
  - first, second, third, ...
  - cold, warm, hot
- Nominal/Interval (quantitative)
  - list of integers or reals

Hearst, 2003

# Data Types - Overview

- **Generic**
  - **entity, relationship,**
  - **Attribute to entity or relationship**
  - **operation**
  
- **Specific**
  - **1-D Linear** Document Lens, SeeSoft, Info Mural, Value Bars
  - **2-D Map** GIS, ArcView, PageMaker, Medical imagery
  - **3-D World** CAD, Medical, Molecules, Architecture
  - **Multi-Dim** Parallel Coordinates, Spotfire, XGobi, Visage, Influence Explorer, TableLens, DEVis
  - **Temporal** Perspective Wall, LifeLines, Lifestreams, Project Managers, DataSpiral
  - **Tree** Cone/Cam/Hyperbolic, TreeBrowser, Treemap
  - **Network** Netmap, netViz, SeeNet, Butterfly, Multi-trees

*Shneiderman, 2003*

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

57

# Interpretations of Visual Properties

Some properties can be discriminated more accurately but don't have intrinsic meaning

(Senay & Ingatious 97, Kosslyn, others)

- **Density (Greyscale)**  
Darker -> More
- **Size / Length / Area**  
Larger -> More
- **Position**  
Leftmost -> first, Topmost -> first
- **Hue**  
??? no intrinsic meaning
- **Slope**  
??? no intrinsic meaning

Hearst, 2003

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

58

## Accuracy Ranking of Quantitative Perceptual Tasks static features

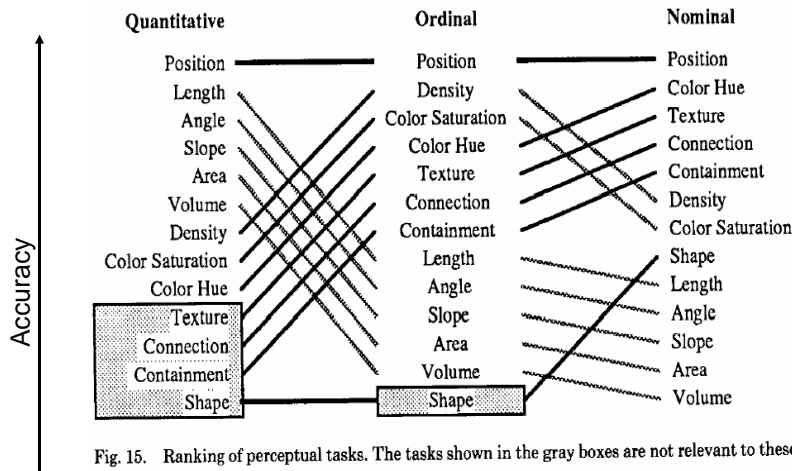


Fig. 15. Ranking of perceptual tasks. The tasks shown in the gray boxes are not relevant to these types of data.

Mackinlay 88

14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

59

## Ranking of Applicability of Properties for Different Data Types (Mackinlay 88, Not Empirically Verified)

QUANTITATIVE	ORDINAL	NOMINAL
Position	Position	Position
Length	Density	Color Hue
Angle	Color Saturation	Texture
Slope	Color Hue	Connection
Area	Texture	Containment
Volume	Connection	Density
Density	Containment	Color Saturation
Color Saturation	Length	Shape
Color Hue	Angle	Length

Hearst, 2003

14/06/05

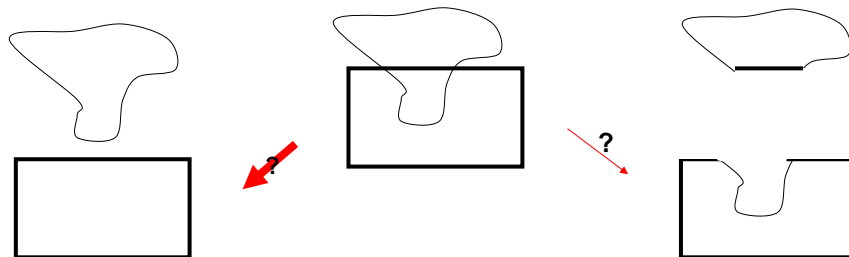
LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

60

# Concepts & Principles

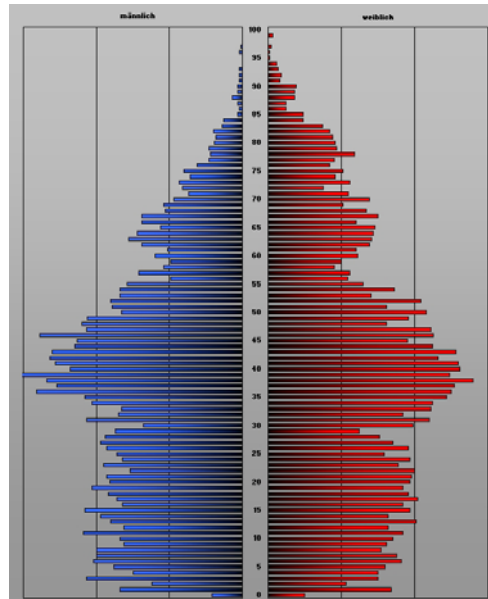
## Continuity

- Experience tells that visual elements are more likely to be continuous
- Implied connection
- connections are used to show relations



# Symmetry

- Symmetrical to emphasizes relationship



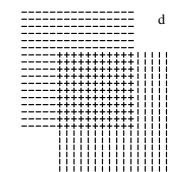
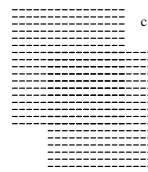
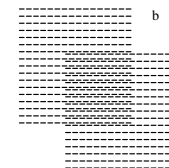
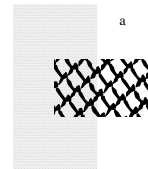
14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

63

# Figure, Background Transparency, Overlap

- What is foreground and what is background?
- Transparency is perceived only when good continuity and color correspondence exists.
- visual interference in overlapping textures



14/06/05

LMU München ... Mensch-Maschine-Interaktion 2 ... SoSe05 ... Schmidt/Butz

64



## Tufte – Principles of Graphical Excellence

- Graphical excellence
  - the well-designed presentation of interesting data – a matter of substance, of statistics, and of design
  - consists of complex ideas communicated with clarity, precision and efficiency
  - is that which gives to the viewer the greatest number of ideas in the shortest time with the least ink in the smallest space
  - **requires telling the truth about the data.**

Hearst, 2003

## Tufte Principle

Maximize the data-ink ratio  
(Avoid “chart junk”)

$$\text{Data-ink ratio} = \frac{\text{data ink}}{\text{total ink used in graphic}}$$

Hearst, 2003

# Tufte's Graphical Integrity

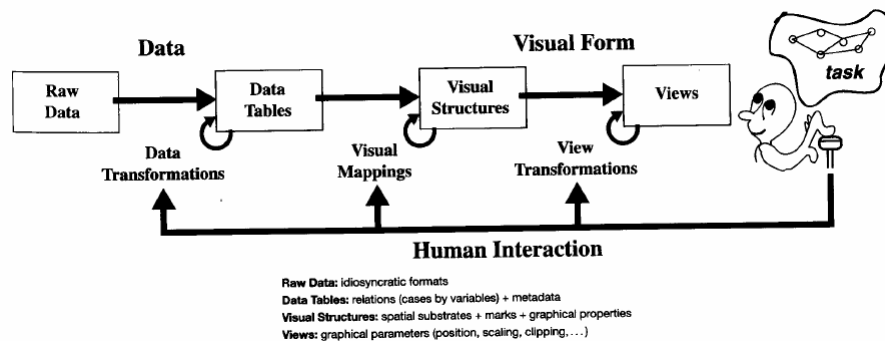
- Some lapses intentional, some not

$$\text{Lie Factor} = \frac{\text{size of effect in graph}}{\text{size of effect in data}}$$

- Misleading uses of area
- Misleading uses of perspective
- Leaving out important context
- Lack of taste and aesthetics

Hearst, 2003

# Visualization Reference Model Human Interaction



- Raw Data → Data Table  
filtering
- Data Table → Visual Structure  
pick mappings
- Visual Structure → Views  
probes, viewpoints, distortions

(Storey, 2004)

# References

- A Review and Taxonomy of Distortion-Oriented Presentation Techniques, Leung & Apperley, 1994
- Barlow et al. "A Comparison of 2-D Visualizations of Hierarchies" INFOVIS'01 <http://www.sims.berkeley.edu/courses/is247/s02/readings/barlow.pdf>
- Martin Wattenberg. Arc Diagrams: Visualizing Structure in Strings IBM Watson Research Center, Technical report 2002-11 <http://domino.research.ibm.com/cambridge/research.nsf/0/e2a83c4986332d4785256ca7006cb621?OpenDocument>
- Thread Arcs <http://www.research.ibm.com/research/threadarcs.html>
- Focus+Context Taken Literally, Robert Kosara, Silvia Miksch, Helwig Hauser, 2000
- Marti Hearst, <http://bailando.sims.berkeley.edu/talks/chi03-tutorial.ppt>
- Storey, [http://www.cs.uvic.ca/~mstorey/teaching/infovis/course\\_notes/introduction.pdf](http://www.cs.uvic.ca/~mstorey/teaching/infovis/course_notes/introduction.pdf)
- Shneiderman, <http://www.cs.ubc.ca/~tmm/courses/cpsc533c-03-spr/readings/shneiderman96eyes.pdf>