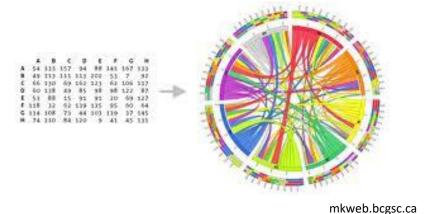
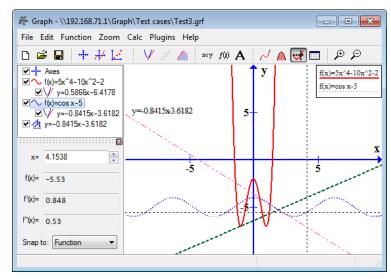
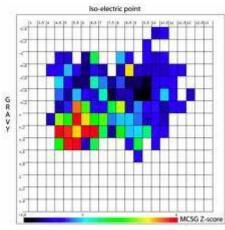
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      <Domain>@AX.LOCAL</Domain>
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         <COSTCENTER>112400</COSTCENTER>
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     - <DETAILS>
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        <DESCRIPTION>Row 2</DESCRIPTION>
       </DETAILS>
      </ROW>
    </ROWS>
   </INFORMATION>
                                axblog4u.wordpress.com
 </INFORMATIONS>
```



2. Input data characteristics



www.padowan.dk



technology.sbkb.org

Process of data visualization

- Generating data
 - Measuring, simulation, modeling
 - Can be lengthy (measuring, simulation) and costly (simulation, modeling)
- Visualization (the rest of the visualization pipeline)
 - Visual mapping, rendering
 - Can be fast or slow, depending on hardware and implementation
- Interaction (user feedback)
 - How the user can interact with the visualization

Passive visualization

- The following three steps are strictly separated
 - Generating data after finishing this phase
 - Off-line visualization
 - Displaying the generated data
 - Result is a video or animation
 - Passive visualization
 - Exploration of the results of the previous phase

Interactive visualization

- Here only the generating data phase is separated
 - Off-line data generation



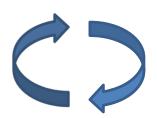
Generated data is available for interactive visualization



- Options: selection, parametrization of visualization
- Currently very popular technique

Interactive steering

- All three steps are connected
 - Generating data on the fly



- Interactive visualization enabling real-time view onto data
- Extended interaction
 - The user can control the simulation process, change design when modeling, etc.
 - Very complicated and costly process

Comparison

complexity, technical requirements

interactive steering

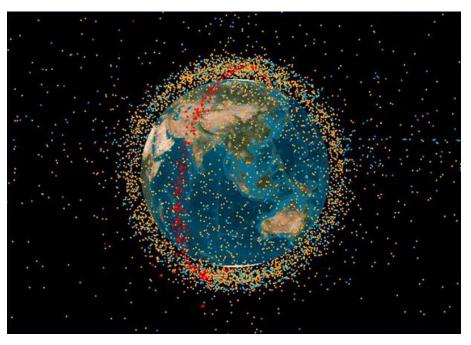
interactive visualization

passive visualization

advantages, possibilities

Data

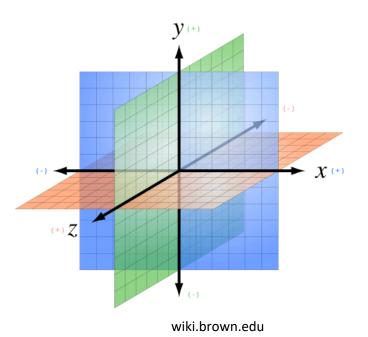
- Central topic of visualization
- Data influences the selection of appropriate visualization technique (along with the user)
- Important questions:
 - Where data "lives"(what is the data space)
 - Type of data
 - Which representation is meaningful

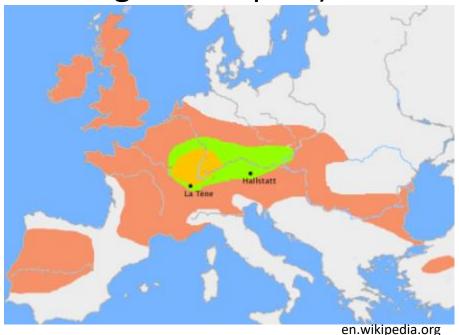


Data space

- Different properties
 - Dimensionality of data space
 - Coordinate system

Region of influence (local or global impact)





Data definition

Raw data x preprocessed data

- Data item (r₁, r₂, ..., r_n)
- Each ri record contains m variables (v₁, v₂, ..., v_m)
- v_i is often denoted as observation

Definition of variables

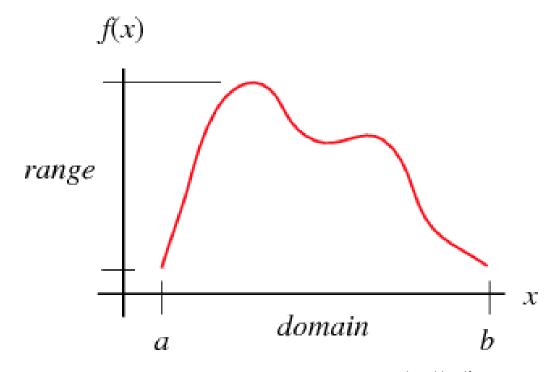
- Independent variable ivi
 - not influenced by any other variable (e.g., time)
- Dependent variable dv;
 - is influenced by one or more independent variables (e.g., temperature)
- Record can be represented as

```
r_i = (iv_1, iv_2, ..., iv_{m_i}, dv_1, dv_2, ..., dv_{m_d})
```

```
where m = m_i + m_d
```

Data generated by function

- Independent variables = domain
- Dependent variables = range



Types of variables

- Physical types
 - Characterized by the input format
 - Characterized by the type of possible operations
 - Example: bool, string, int, float,...
- Abstract types
 - Data description
 - Characterized by methods/attributes
 - Can be hierarchical
 - Example: plants, animals, ...

Data types

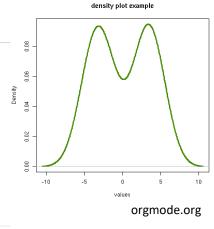
Ordinal

- Binary
- Discrete

Continuous







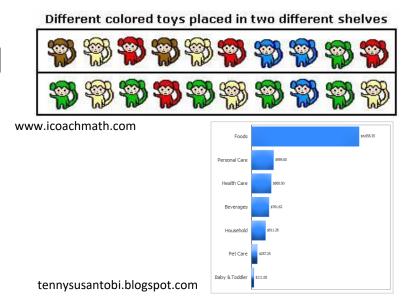
Viewfinder

www.cincomsmalltalk.com

File Edit View Light Misc

Nominal

- Categorical
- Sorted
- Random



Scale

- 3 basic attributes:
 - Ordering relation on data
 - Distance metric
 - Existence of absolute zero ABSOLUTE ZERO
 - Fixing the minimal value of variable

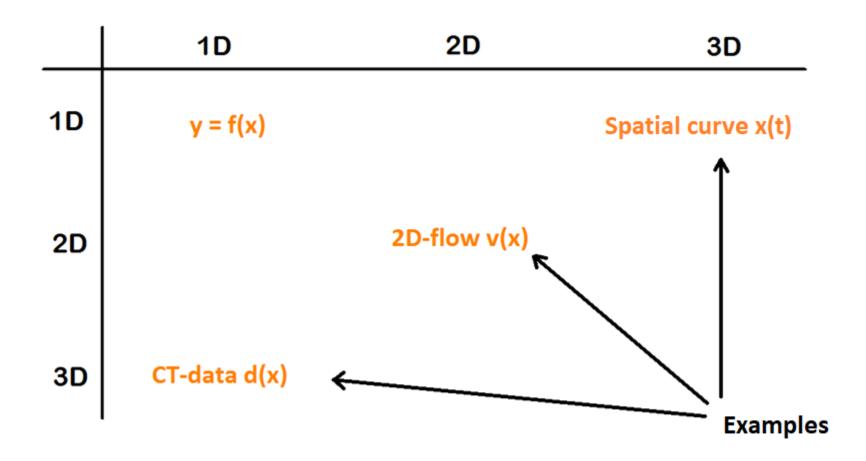


The only thing cooler is not being a scientist.

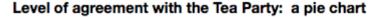
Data representation

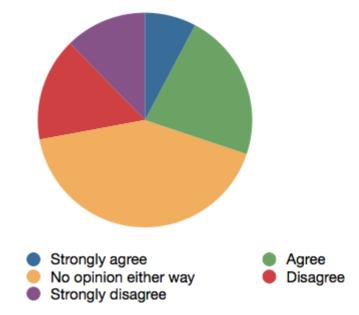
- Depends on:
 - The presence of spatial domain
 - If it is not inherently in data, which domain to choose?
 - How the dimensions are used?
 - Data characteristics
 - Available visualization space (2D/3D)
 - Which part of data is in focus?
 - In which parts we can use more abstracted representation?

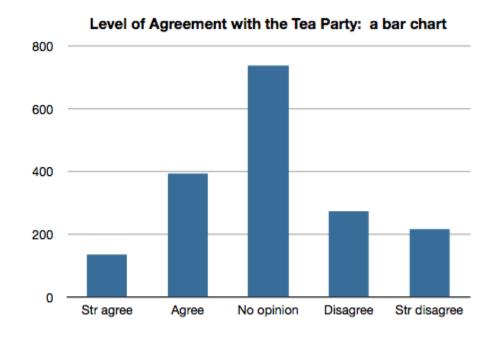
Data space vs. data properties



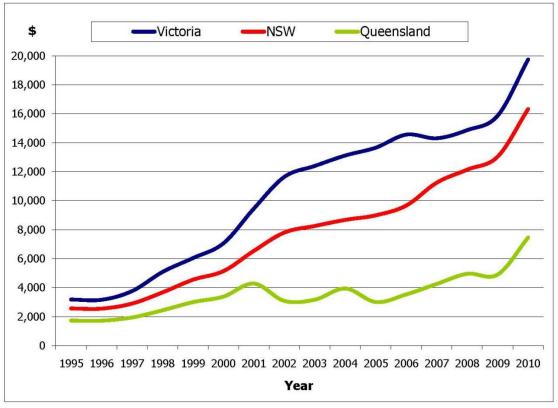
 Discrete data – set of values, visualization using bar charts, pie charts, ...



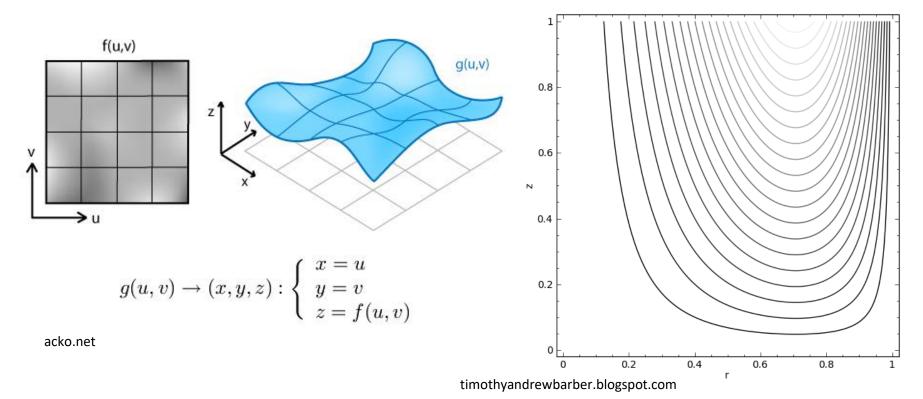




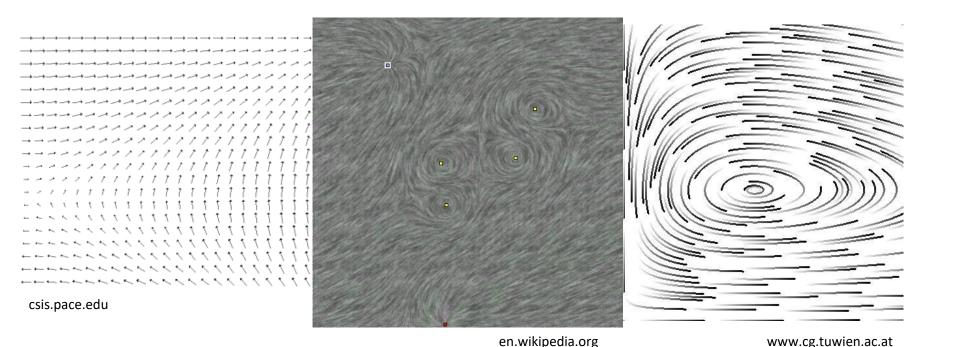
 Continuous data – function, visualization using graphs



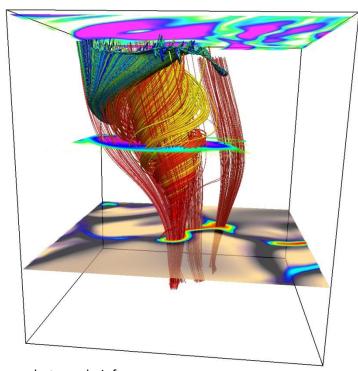
- 2D real numbers
 - Function of two variables, visualization using 2D height maps, contours in 2D, ...

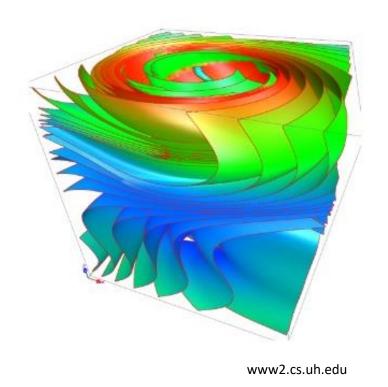


 2D vector fields, visualization using hedgehog plots, LIC (line integral convolution), streamlets, ...



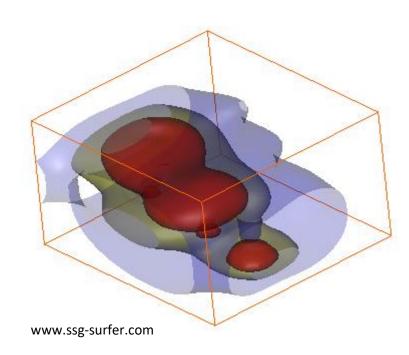
- Spatial data + time
 - 3D flow, visualization using streamlines, streamsurfaces





Spatial data

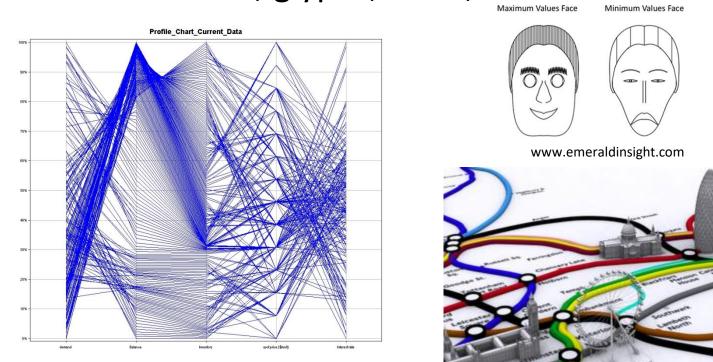
3D density, visualization using isosurfaces, volume rendering





- Multidimensional data
 - Set of n dimensions, visualization using parallel coordinates, glyphs, icons, ...

datamining.typepad.com



www.cs.umd.edu

Structure inside and between records

- Data sets consist of:
 - Syntax data representation (so called data model)
 - Semantics relationships within one record or between records (so called conceptual model)
- Types of structures:
 - Scalars, vectors, tensors
 - Geometry and grids
 - Other forms

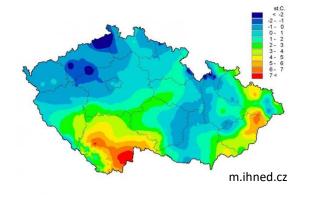


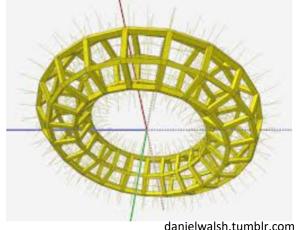
Scalars, vectors, and tensors

- Scalar = individual number in record
 - e.g., age
- Vector = composition of several variables to one record
 - e.g., point in 2D space, RGB
- Tensor = defined by its order and space dimension. Represented by field or matrix.
 - e.g., transformation matrix in 3D

Geometry and grids

Geometry is represented using coordinates of records



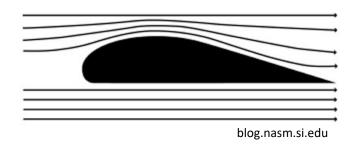


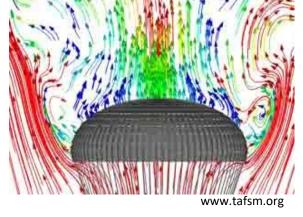
 Grid – geometry can be derived from the starting position, orientation, and step size in horizontal and vertical direction

Non-uniform geometry

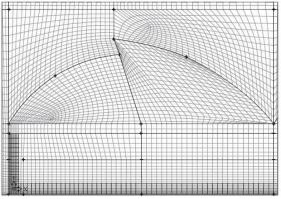
We need to store coordinates of all records –

they cannot be derived





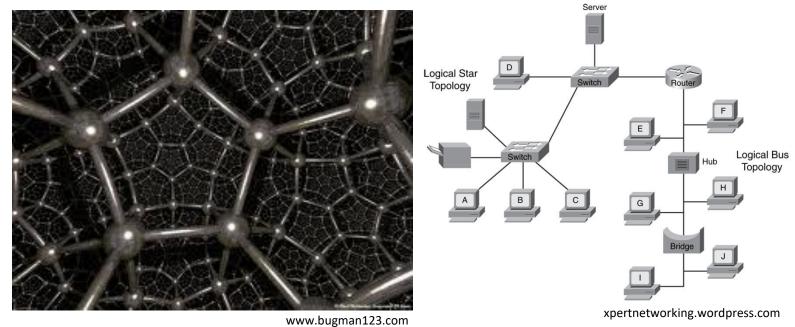
Non-uniform grid



www.scielo.org.mx

Other types of structures

- Another important structure is topology
- It defines so called connectivity



Important in resampling and interpolation

Time

 Enormous range of values (picosecs vs. millenia)



megworden.com

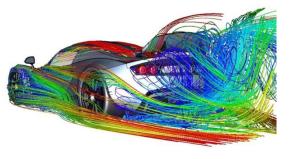
- Expressed absolutely or relatively
- Data sets containing time can have regular distribution (regular sampling) or irregular one (e.g., transaction processing – according to the time stamp of its execution)

Other examples of structured data

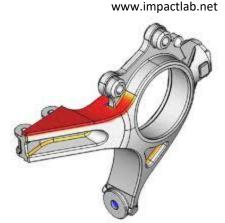
- Magnetic Resonance Imaging (MRI)
- Computational fluid dynamics (CFD)
- Financing
- CAD systems
- Counting people

Social networks

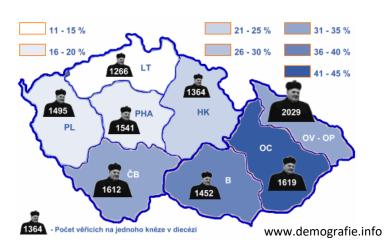




lotusenthusiast.net

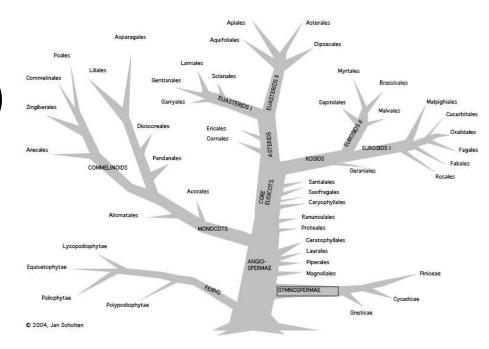


www.mjmdesigns.co.uk



Taxonomy – 7 types of data

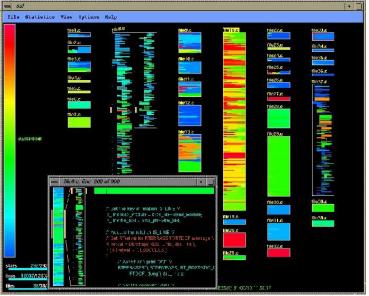
- 1D (linear sets and sequences)
- 2D (maps)
- 3D (objects, shapes)
- nD (relations)
- Trees (hierarchy)
- Networks (graphs)
- Temporal data

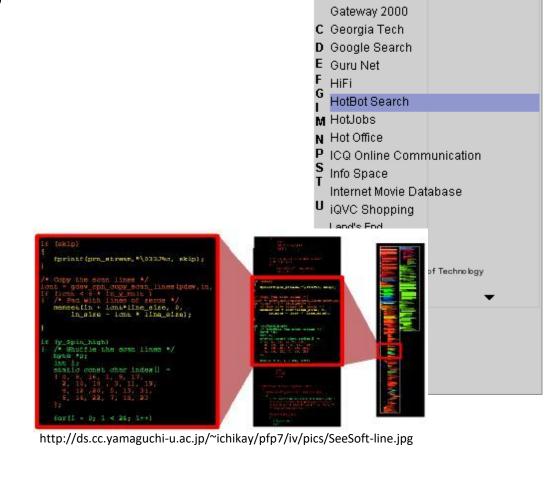


www.ianscholten.com

Linear data

- Long lists of items
 - Menu items
 - Source code
- Fisheye displays





http://www.cs.umd.edu/hcil/fisheyemenu/

Fisheye

Free Merchant Business

Free Shop Furniture

Garden

2D data - maps

- GIS (Geographical Information Systems)
 - Maps (e.g., Google Earth)

http://www.wimp.com/unusualplaces/

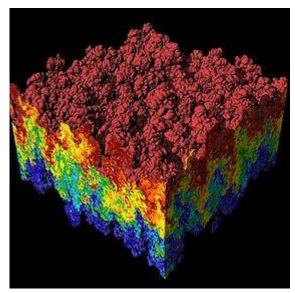
- Spatial queries
- Spatial data analysis



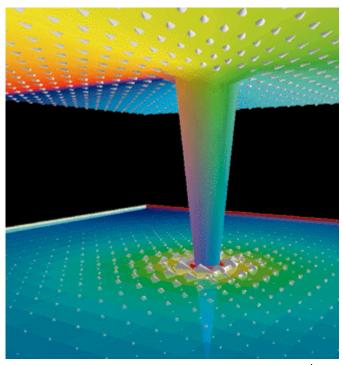
www.bowdoin.edu

3D data

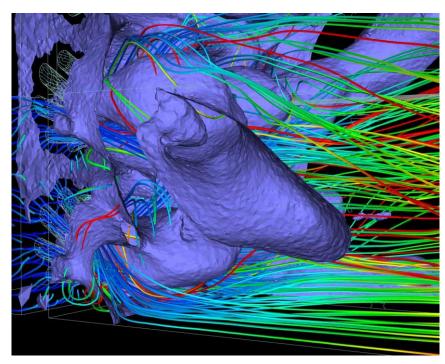
- Different types of 3D data vis
- Scientific visualization



en.wikipedia.org



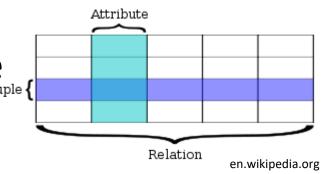
www.ornl.gov



gvis.grc.nasa.gov

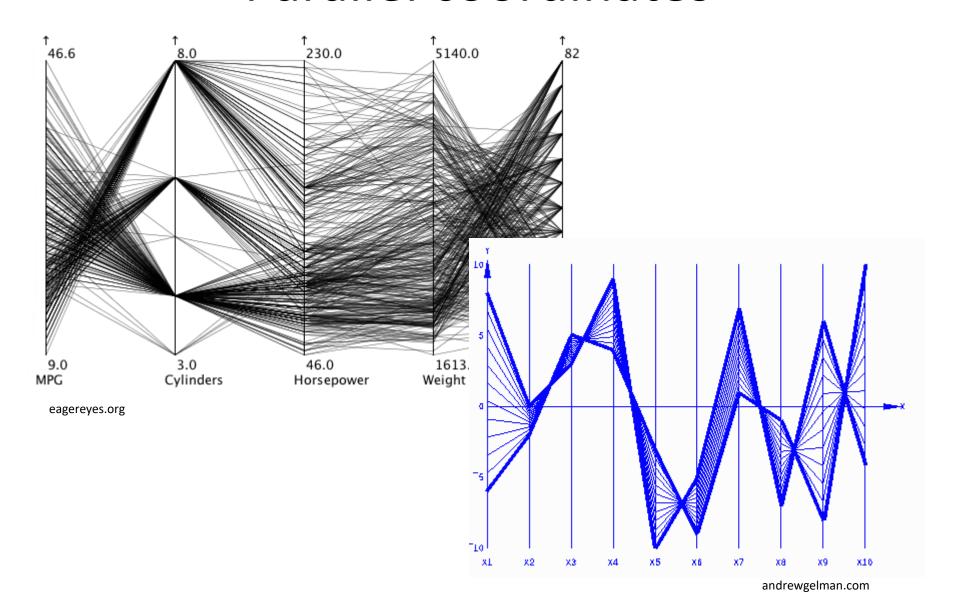
Multidimensional data

Records in relational database



- Two solutions:
 - Drawing all possible pairs of variables in 2D graph
 - Simple but unusable for general overview of the data
 - "Parallel coordinates"
 - Method for displaying multidimensional data (Alfred Inselberg)

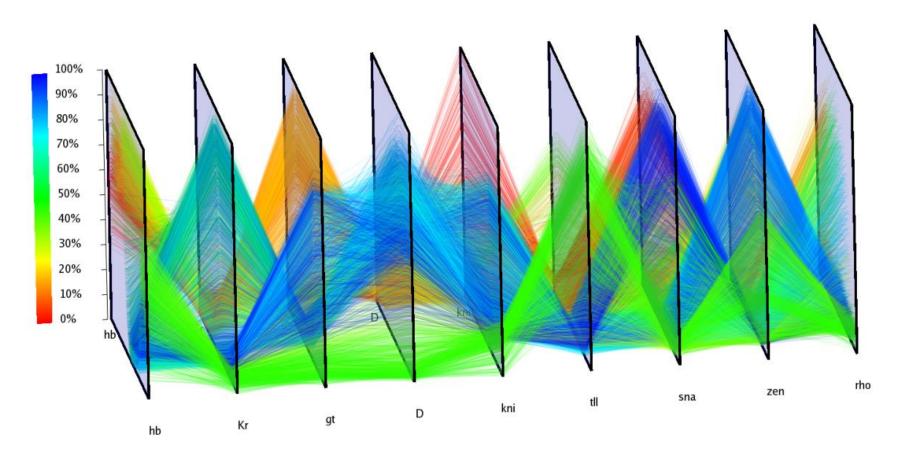
Parallel coordinates



Parallel coordinates

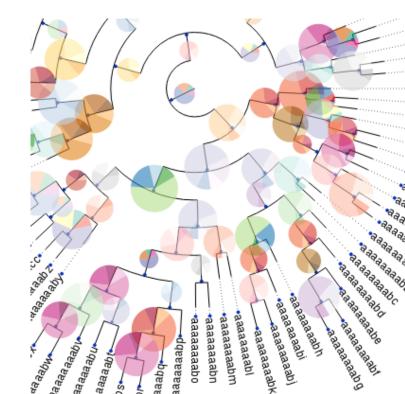


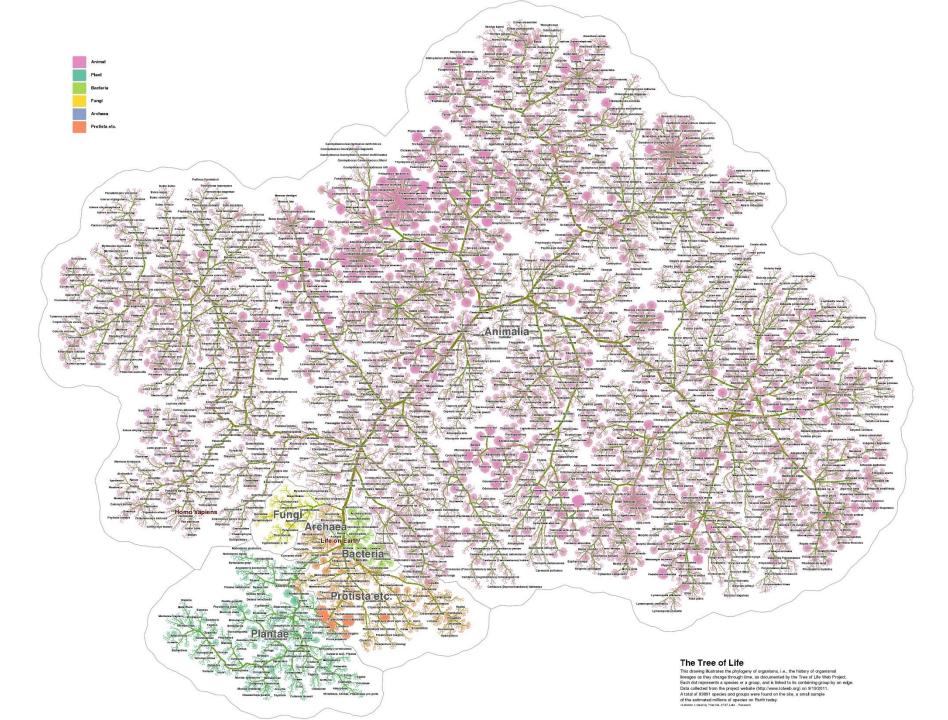
whyevolutionistrue.wordpress.com

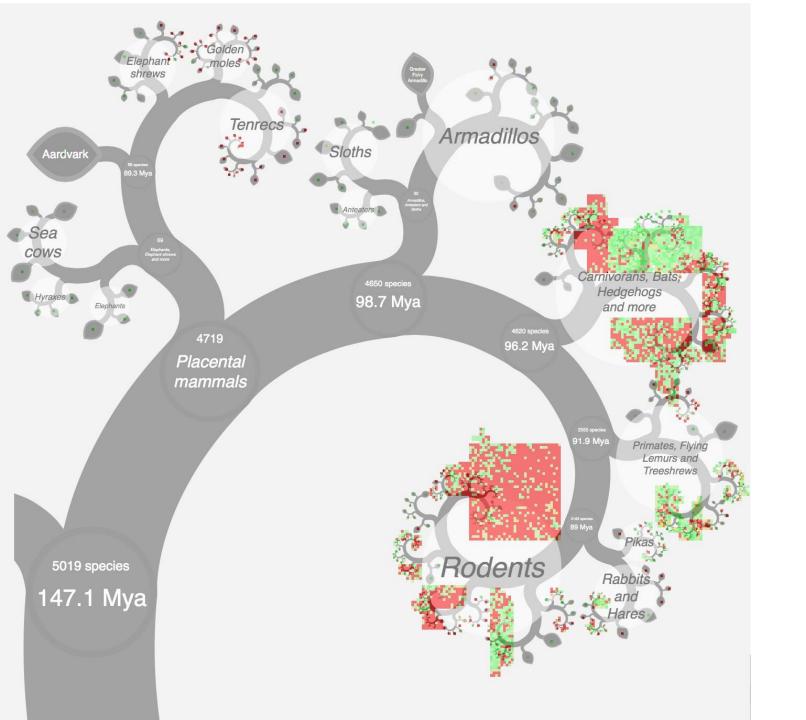


Trees

- Displays not only data itself but also their structure
 - e.g., genetic trees, file systems
- Number of items
 increases significantly
 in lower levels of tree



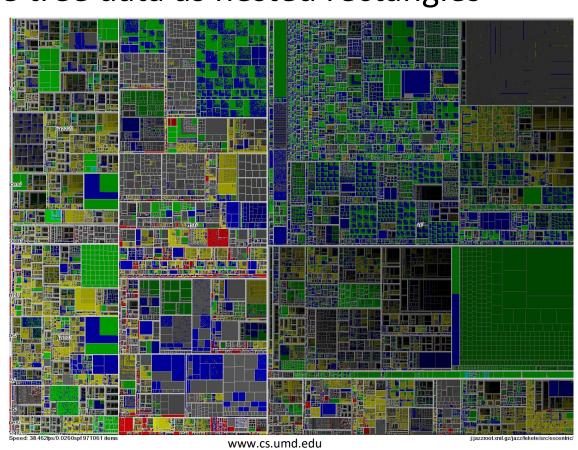




Trees

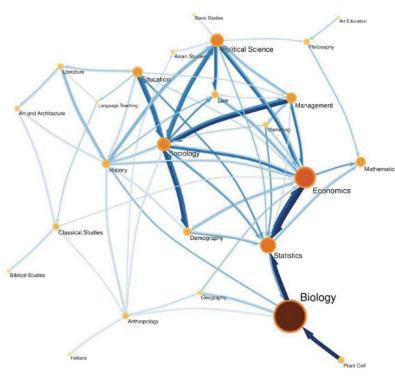
- Tree Maps
 - Displaying the tree data as nested rectangles

Tree with a million records:



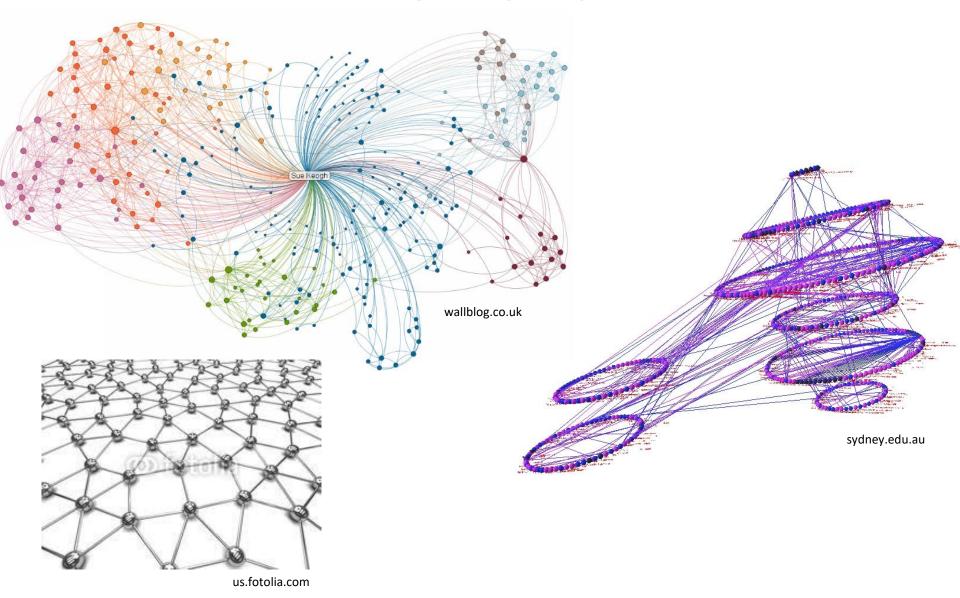
Networks

- Similar to trees we aim to display the data structure
- Networks = nodes + edges
- Design should contain:
 - Minimal edge crossings
 - Minimal edge length
 - Minimal edge bending



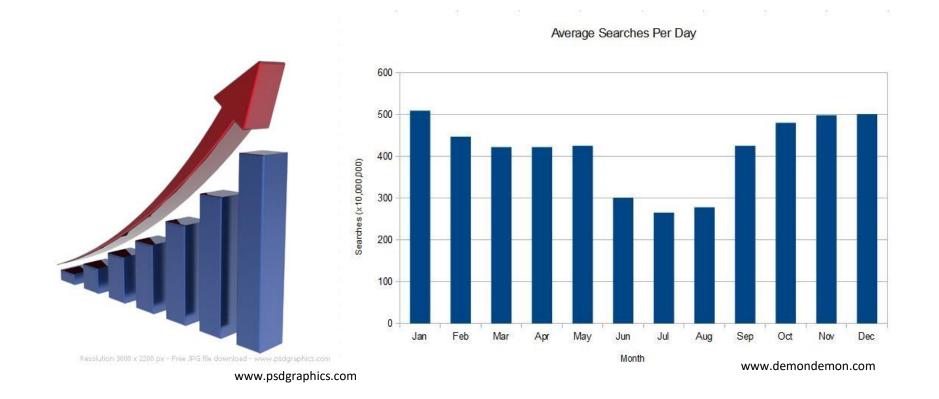
archaeologicalnetworks.wordpress.com

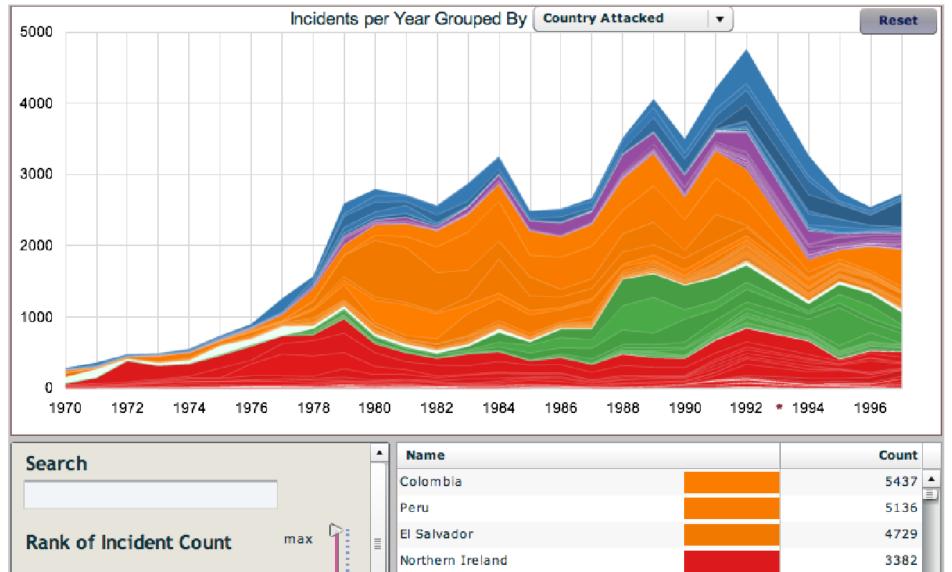
Networks



Temporal data

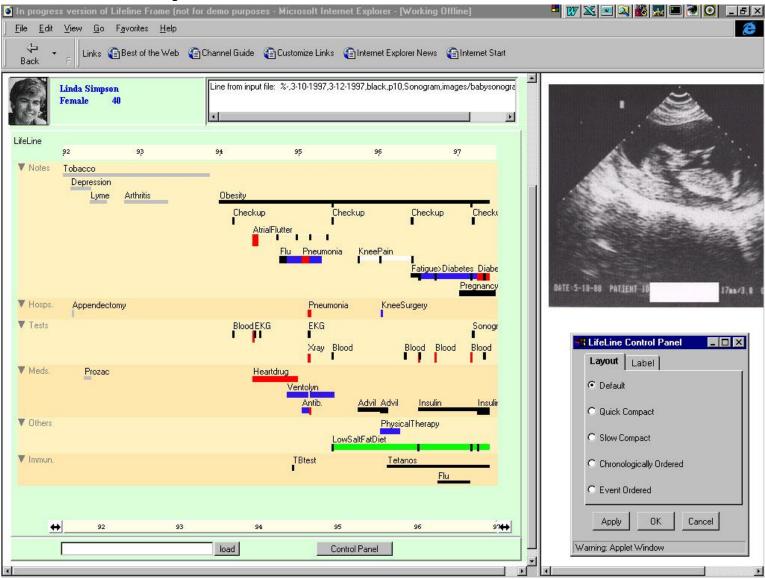
- Displaying data dependent on time
 - Trend and seasonal graphs







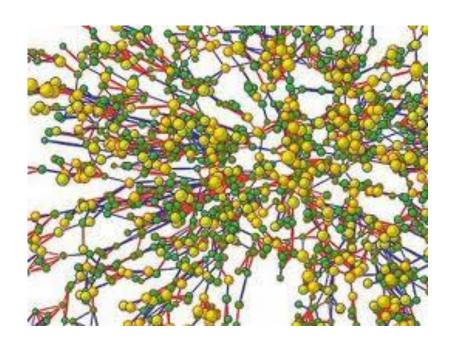
Temporal data - LifeLines



http://www.cs.umd.edu/hcil/lifelines/

Data preprocessing

- Displaying raw data = precise, identification of outliers, missing data, ...
- Sometimes preprocessing is required



Preprocessing – techniques

- Metadata and statistics
- Missing values and data "cleaning"
- Normalization
- Segmentation
- Sampling and interpolation
- Dimension reduction
- Data aggregation
- Smoothing and filtration
- Raster to vector

Metadata and statistics

- Metadata information for preprocessing
 - Reference point for measurement
 - Unit of measurement
 - Symbol for missing values
 - Resolution
- Statistical analysis
 - Detection of missing records
 - Cluster analysis
 - Correlation analysis

If you had two cans without labels, which would you eat?

Without a label, how would you know which was tuna and which was cat food?



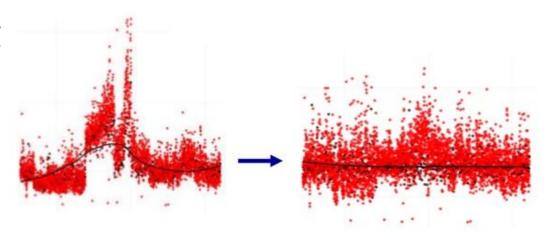


Missing values and data "cleaning"

- Removing wrong records
- Assigning a given value
- Assigning an average value
- Assigning a value derived from the nearest neighbor value
- Calculating the value (imputation)

Normalization

- Transformation of the input dataset
- Adjusting values measured on different scales to a notionally common scale
- Normalization to interval [0.0, 1.0]:
 dnormalized = (doriginal dmin)/(dmax dmin)
- Clamping according to the threshold values

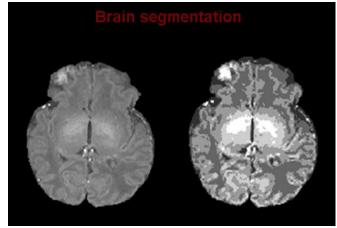


Segmentation

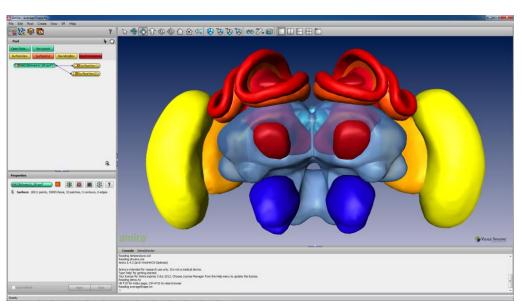
Classification of input data into given

categories

 Split-and-merge iterative algorithm



blog.campaigner.com



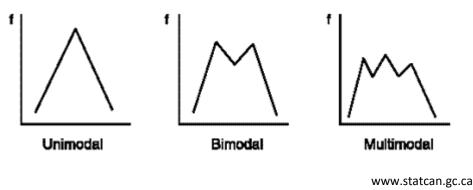
Split-and-merge

- similarThresh = defines the similarity of two regions with given characteristics
- homogeneousThresh = defines the region homogeneity (uniformity)

```
do {
           changeCount = 0;
           for each region {
                       compare region with neighboring ones and find the most similar one;
                       if the most similar one is within similarThresh of the current region {
                                   connect these two regions;
                                   changeCount++;
                       evaluate the homogeneity of the region;
                       if homogeneity of region is smaller than homogeneousThresh {
                                   split the region to two parts;
                                   changeCount++;
} until changeCount == 0
```

Complex parts of the algorithm

- Determining the similarity of two regions
- Evaluating the homogeneity of a region histogram



Splitting the region

Possible problem

 Infinite loop by repeating split and merge steps of the same region

Solution:

- Changing the threshold value for similarity or homogeneity
- Taking into account other region properties (e.g., size and shape of regions)

Sampling and interpolation

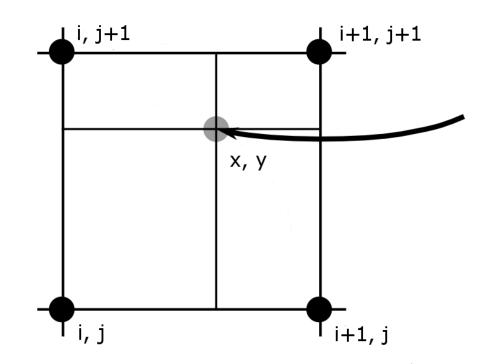
- Transformation of input data
- Interpolation = sampling method
 - Linear interpolation
 - Bilinear interpolation
 - Non-linear interpolation



inperc.com

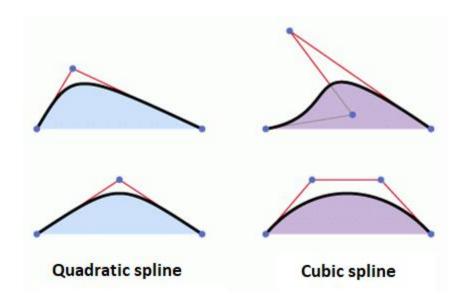
Bilinear interpolation

- Uniform grid
- Horizontal + vertical interpolation



Non-linear interpolation

- Problems with linear interpolation zero connectivity in grid points
- Solution = using quadratic and cubic splines

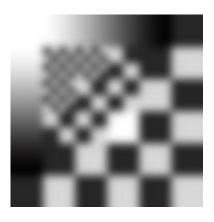


Result

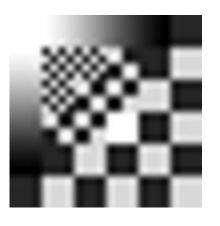
Original image (24x24 pixels)



cubic B-spline filter



Catmull-Rom



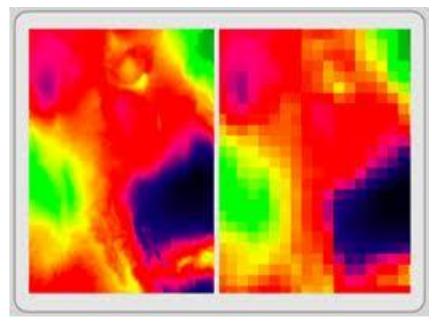
research.cs.wisc.edu

Resampling

Pixel replication

Neighbor averaging

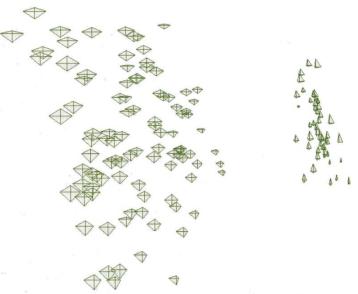
Data subsetting



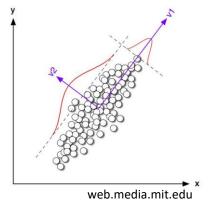
giscommons.org

Dimension reduction

- Preparing multidimensional data for displaying
- Keep as much original information as possible
- Techniques:
 - PCA (principal component analysis)
 - MDS (multidimensional scaling)
 - SOMs (Kohonen self-organizing maps)

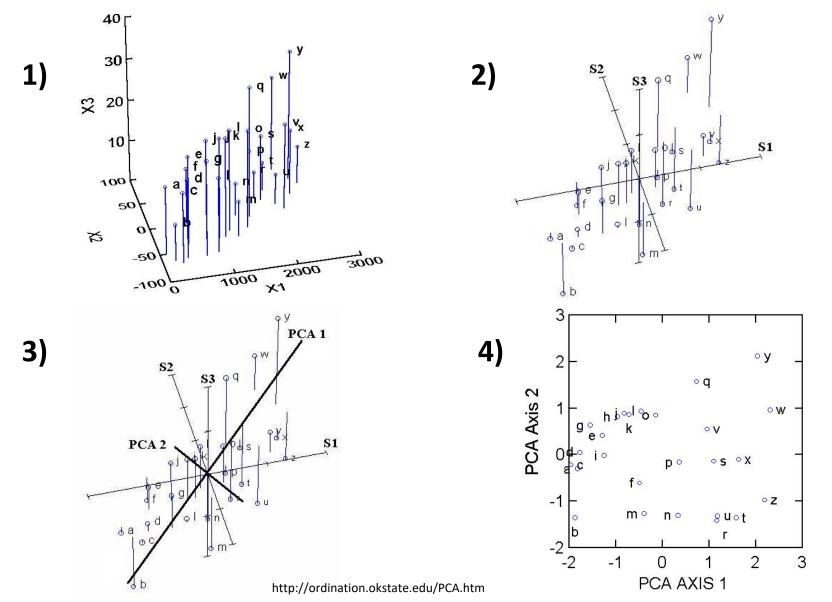


PCA intuitively



- 1. We select a line in space visualizing n-dimensional data. This line covers the most of the input data items and is called the first principal component (PC).
- 2. We select a second line perpendicular to the first PC, this forms the second PC.
- 3. We repeat this until we proces all PC dimensions or until we reach a desired number of principle components.

PCA – principal component analysis



MDS – multidimensional scaling

Based on comparing the distances between individual data items in original and reduced space

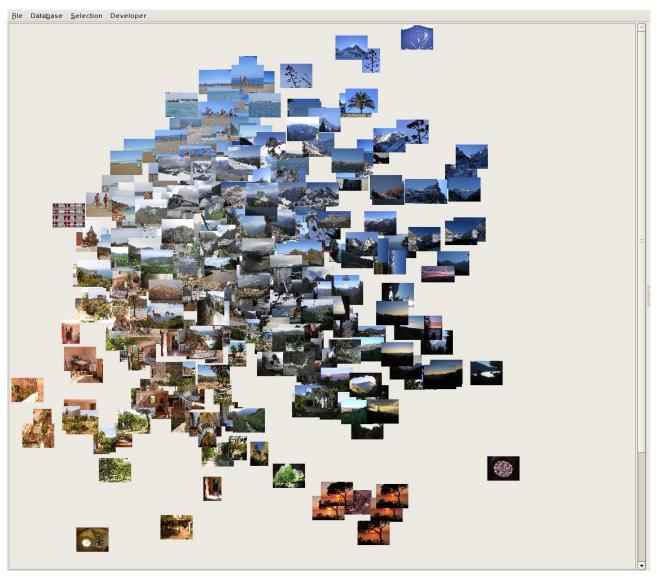
MDS embedding of the digits, (time 5.88s)

MDS embedding of the digits (time 5.88s)

MDS – multidimensional scaling

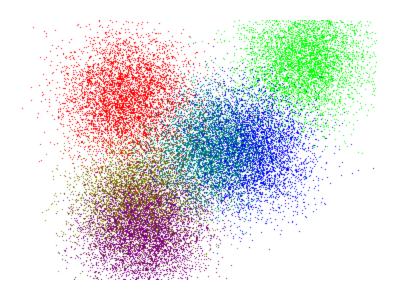
- 1) We calculate the distances between all pairs of data points in the original space. If we have n points as an input, this step requires n(n-1)/2 operations.
- 2) We transfer all input data points to points in the reduced dimension space (often randomly).
- 3) We calculate *stress*, i.e., difference in distance between points in the original and reduced space. This can be done using different approaches.
- 4) If the average and cummulated *stress* value is smaller than the user-defined threshold, the algorithm ends and returns the result..
- 5) If the *stress* value is higher than the threshold, for each point we calculate a directional vector pointing to the desired shift direction in order to reduce *stress* between this point and the other points. This is determined as the weighted average of vectors between this point and its neighbors and its weight is derived from *stress* value calculated between individual pairs. Positive *stress* value repulses the points, negative one attracts them. The higher the absolute value of *stress*, the bigger movement of point.
- 6) Based on these calculations we transform the data points to the target reduced dimension, according to the calculated vectors. Return to step 3 of the algorithm.

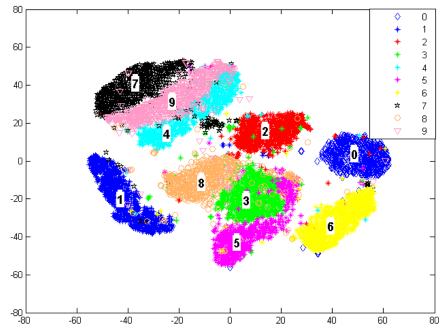
MDS – multidimensional scaling



Data aggregation

 Aggregation = clustering of similar data to groups.





Smoothing and filtration

- Signal processing techniques noise removal
- Convolution in 1D:

$$p_i = \frac{p_{i-1}}{4} + \frac{p_i}{2} + \frac{p_{i+1}}{4}$$

Converting rasters to vectors

Used for:

- Data compression
- Image comparison
- Data transformation

Methods:

- Thresholding
- Region growing
- Edge detection

— ...



Conclusion

The techniques mentioned improve the efficiency of visualization

We have to inform the user that the data has

been transformed!!!

