Reduce Items and Attributes

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Five Major Strategies for Big Data

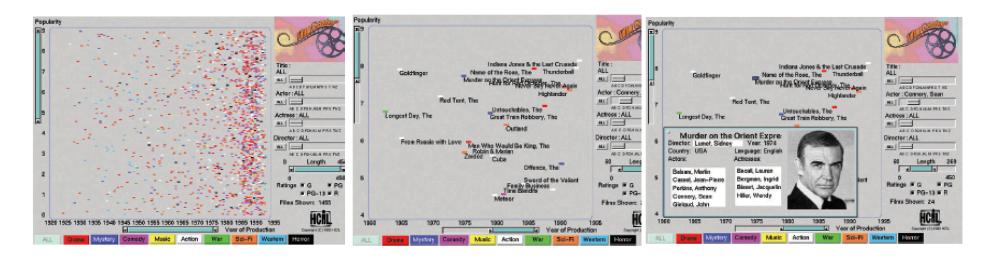
- Derive new (and more compact) data (Tamara Chapter 3)
- Change a view over time (Chapter 11)
- Facet data into multiple views (Chapter 12)
- Reduce items and attributes (Chapter 13)
- Focus+Context viewing (Chapter 14)

Filtering

- Eliminate or select some items and/or attributes to make visual exploration more effective
- Challenges:
 - Without information losses
 - Support effective dynamic queries tightly coupled with visual encoding and interaction
 - Do it efficiently

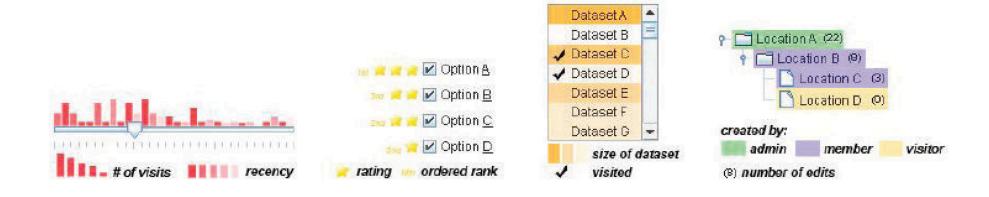
Filter Items

- Example: FilmFinder
 - Use sliders to control what items (films) to show in a scatter plot
 - The marks automatically adapt to the number of items shown (enlarged and labeled)
 - Detail information pops out with mouse over



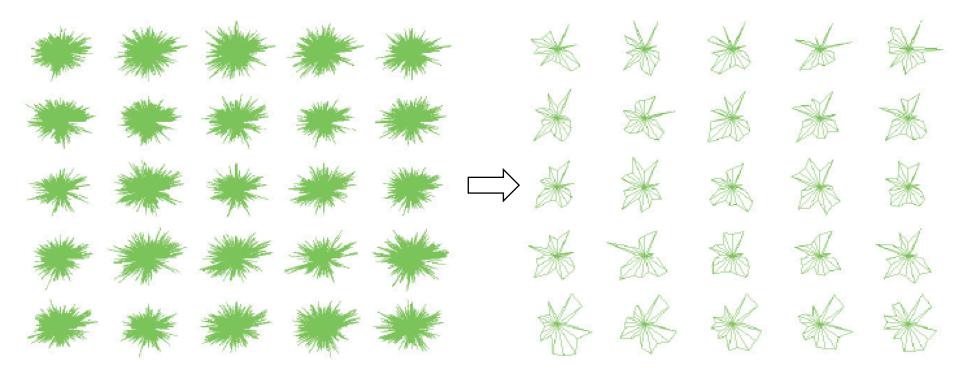
Scented Widget

 Augment the selection widget with concise visual information about the data



Attribute Filtering

- To show the same number of items, but fewer attributes
- Can be combined with item filtering
- Can also benefit from attribute ordering (or clustering) based on their similarity, and then only show the unique ones

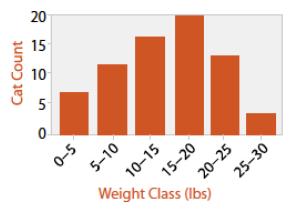


Aggregation

- A group of elements is represented by a new derived elements, e.g., average values
 - Elements are merged with aggregation, as opposed to be filtered/eliminated
- Basic aggregation: average, minimum, maximum, count, and sum
- Challenge: avoid eliminating interesting information

Example: Histogram

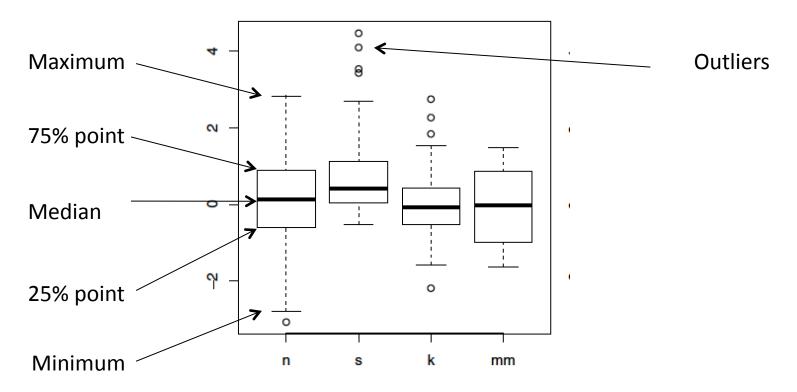
 Bin the data into different ranges, or different categorical types, and then count the number of items in each bin



Idiom	Histograms
What: Data	Table: one quantitative value attribute.
What: Derived	Derived table: one derived ordered key attribute (bin), one derived quantitative value attribute (item count per bin).
How: Encode	Rectilinear layout. Line mark with aligned position to express derived value attribute. Position: key attribute.

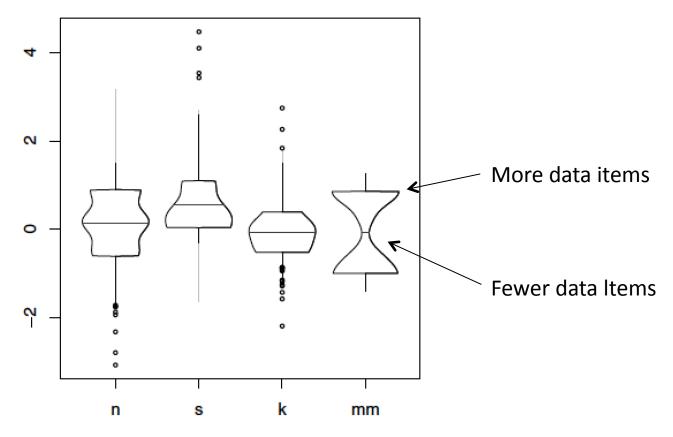
Example: Boxplots

 Compute five basic quantities: median (50% point), first quartile (25% point), third quartile (75% point), and two extremes (minimum and maximum)



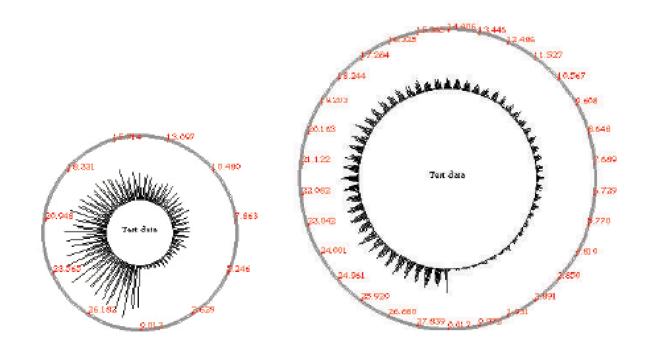
Example: Vaseplots

Augment boxplots with width to depict data density



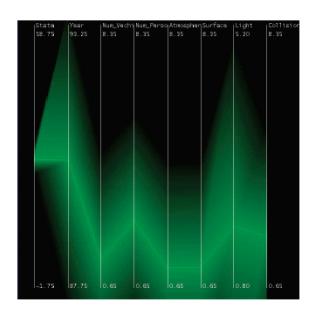
Example: Solarplots

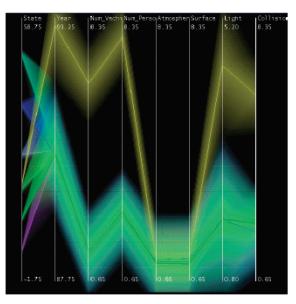
 Different rings indicate histograms of different aggregation levels, inner most being the coarsest

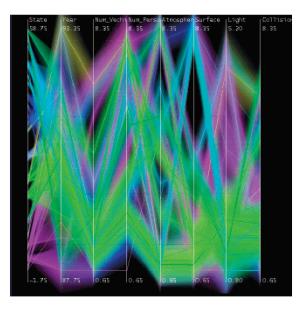


Example: Hierarchical Parallel Coordinates Plots

- Cluster the data items into different number of groups, and show the groups (mean, min, max) instead of the raw data items in PCP
- Inspect the clusters/data at different levels







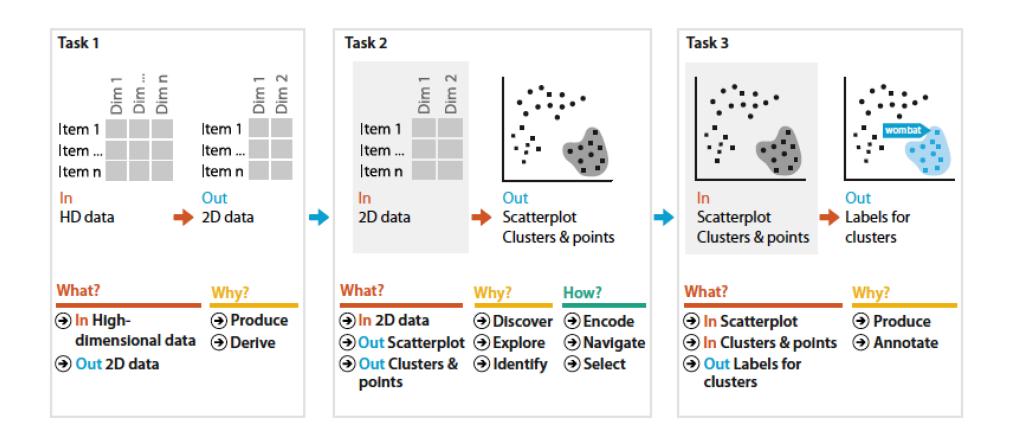
Attribute Aggregation: Dimensionality Reduction

- Use a fewer derived attributes to represent the original data attributes
 - Dimensions: number of attributes
- Goal:
 - preserve the meaningful structure in the data even with the new dimensions
 - This often means preserve the distance between the raw data points
- Common techniques
 - Multidimensional Scaling (MDS)
 - Principle Component Analysis (PCA)

Example: Document Collection

- Transform a document into a bag of words, and count the frequency of each words
 - (vis, tool, filter, aggregate, channel, ...)(75, 10, 25, 34, 50,)
 - This is called a feature vector
- The dimensionality of the feature vec typically very high, so need to be reduced
- Apply dimensionality reduction so that each document can be represented as a visualizable mark (a point for example)

Example: Document Collection



Display Dimensionality Reduction Results

- Two dimensions (e.g. output from Multidimensional Scaling, MDS) can be displayed as a scatter plot
- More than two dimensions can use scatterplot matrix (SPLOM)
- Need to allow the user to inspect the original high dimensional data by selecting the low dimensional derived attributes