

9. Text & Documents

Visualizing and Searching Documents

Vorlesung „Informationsvisualisierung“

Prof. Dr. Andreas Butz, WS 2011/12

Konzept und Basis für Folien: Thorsten Büring

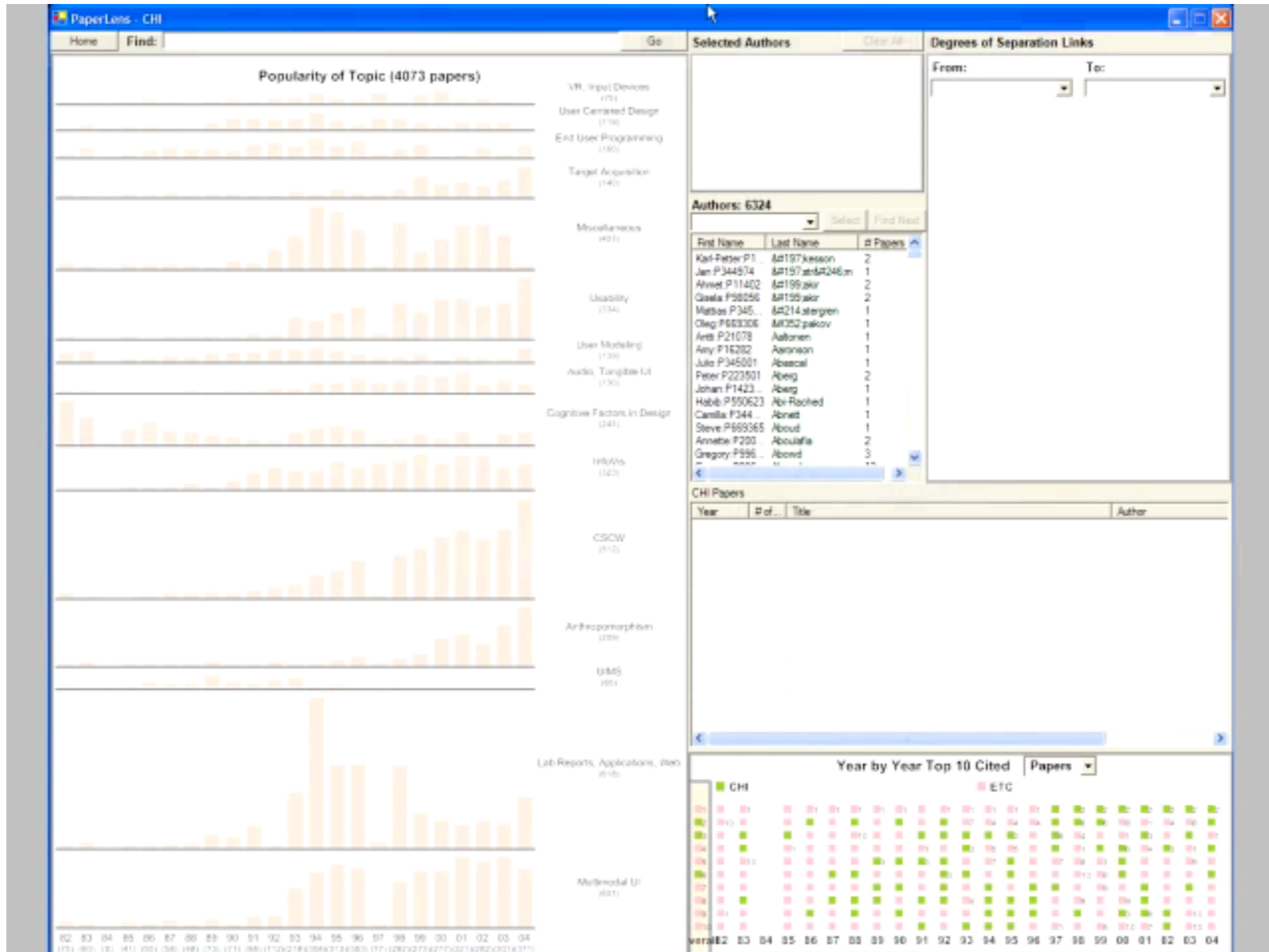
Outline

- Characteristics of text data
- Detecting patterns
 - SeeSoft
 - Arc diagrams
 - Visualizing Plagiarism
- Keyword search
 - TextArc
 - Enhanced scrollbar
 - TileBars
- Cluster Maps
 - Visualization for the document space
 - WEBSOM
 - ThemeScapes
- Cluster map vs keyword search

Text & Documents

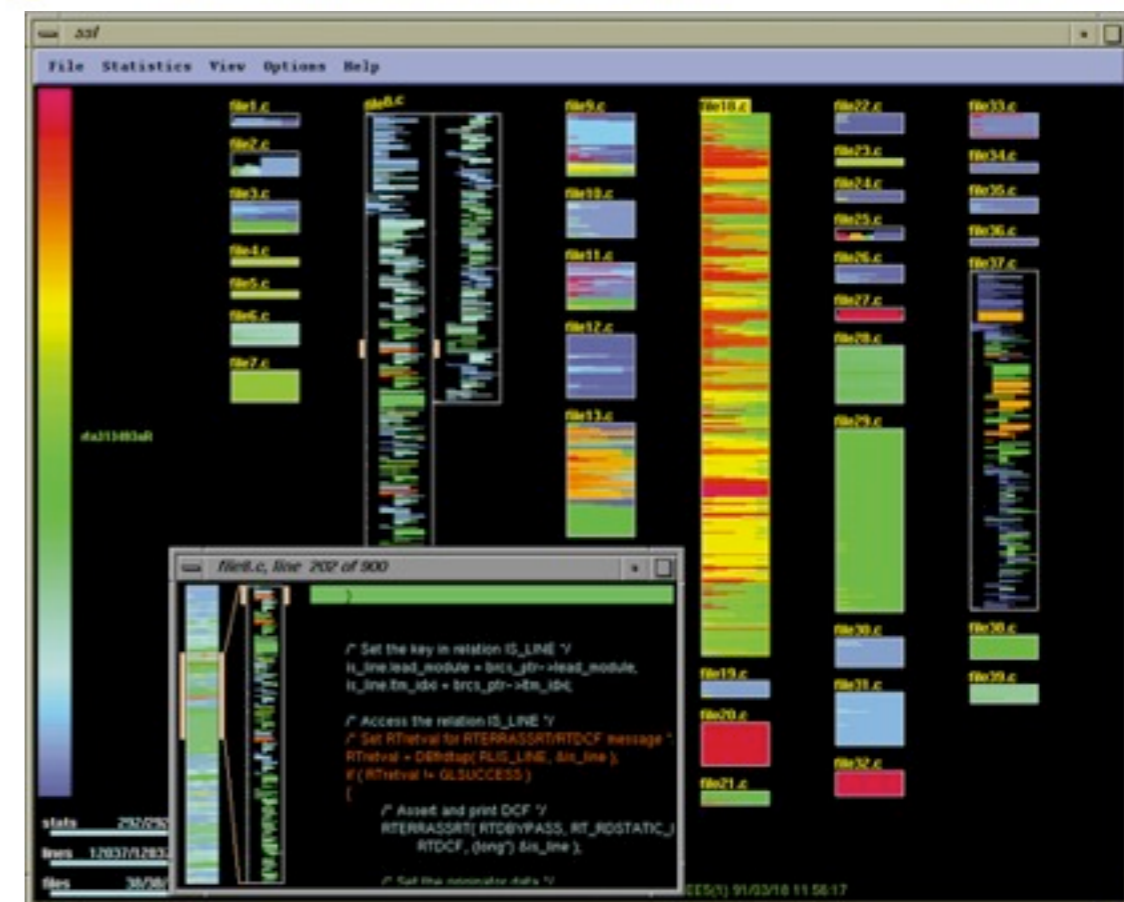
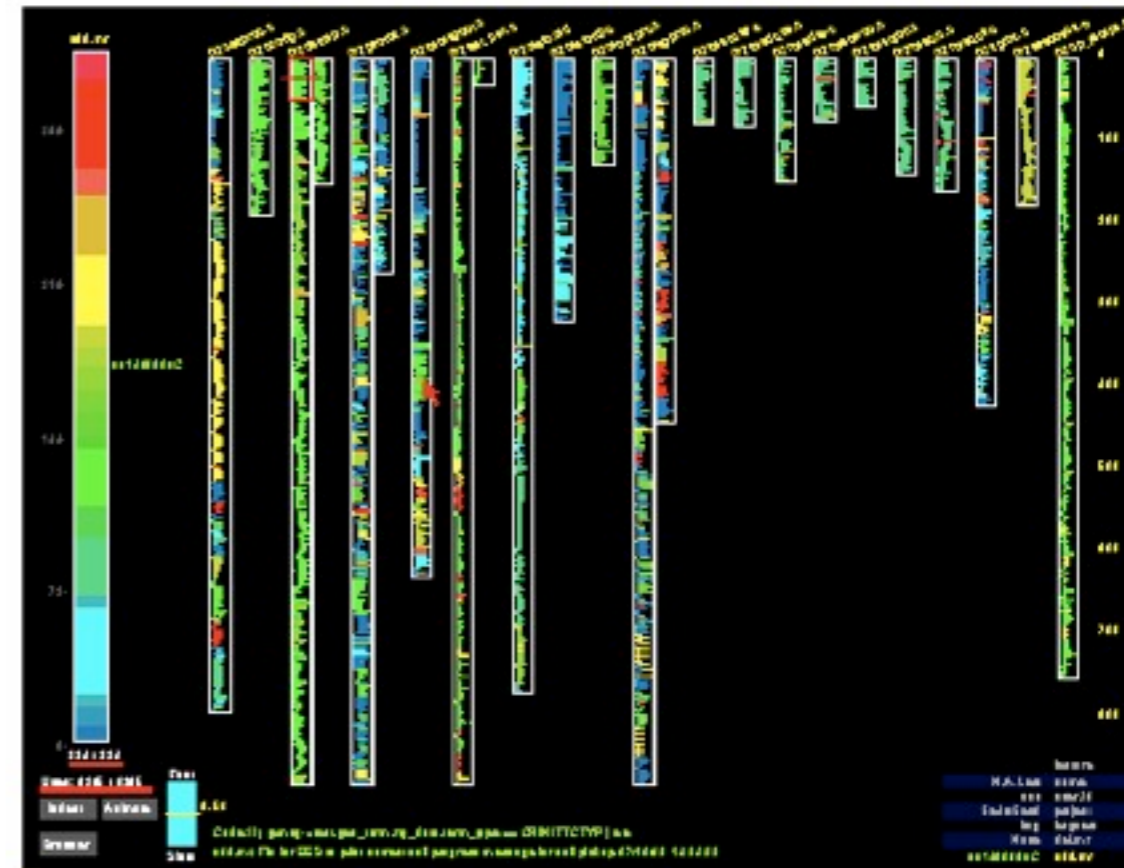
- The main vehicle for humans to store information
- Huge existing resources: libraries, WWW
- What to visualize?
- Text is a nominal data type, but with many additional and interesting properties
- Text structure
- Meta data
 - Author
 - Dates
 - Descriptions
- Relations between documents (e.g. citation, similarity)
- Relevance of documents to a query
- Text statistics (e.g., frequency of different words)
- Content / Semantics

PaperLens



SeeSoft

- Eick et al. 1993
- Software visualization tool to display code line statistics (e.g., age, programmer, number of executions in recent test, etc.)
- Encoding
 - Each column represents a file
 - Height of column: length of the document
 - Files exceeding the height of the screen are continued over to the next columns
 - Each row represents a line of code
 - Width of row: length of line
 - Color: age of the line (red: newest; blue: oldest)
- Scales up to 50,000 lines on a single screen
- Example: 20 files with 9,365 lines of code
- Reading windows controlled by virtual magnifying boxes



Arc Diagrams

- Wattenberg 2002
- Visualizes repetition in string data
- Application domains: text, DNA sequences, music
- Approach: to avoid clutter, only visualize an essential subset of all possible pairs of matching substrings
- Display string on a single line
- Connect the consecutive intervals by a semi-circular arc
 - Thickness of the arc: length of the matching substring
 - Height of the arc: proportional to the distance of substrings



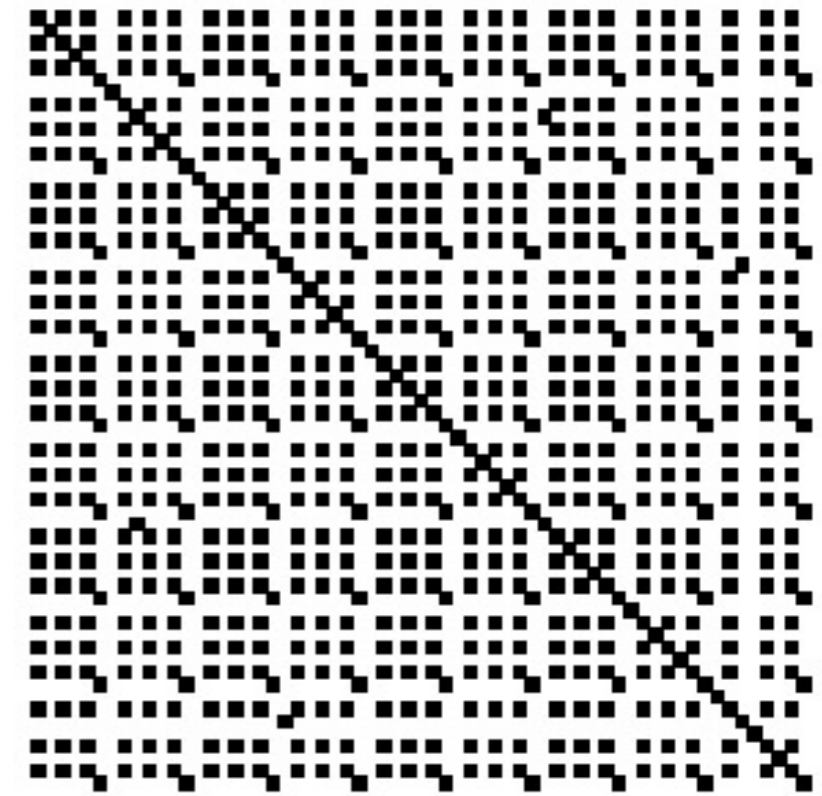
Arc Diagrams

- Apply translucency to not obscure matches
- Still: for strings with a high frequency of small repeated substrings the visualization may cause clutter
- Provide users with the ability to filter by minimum substring length to consider



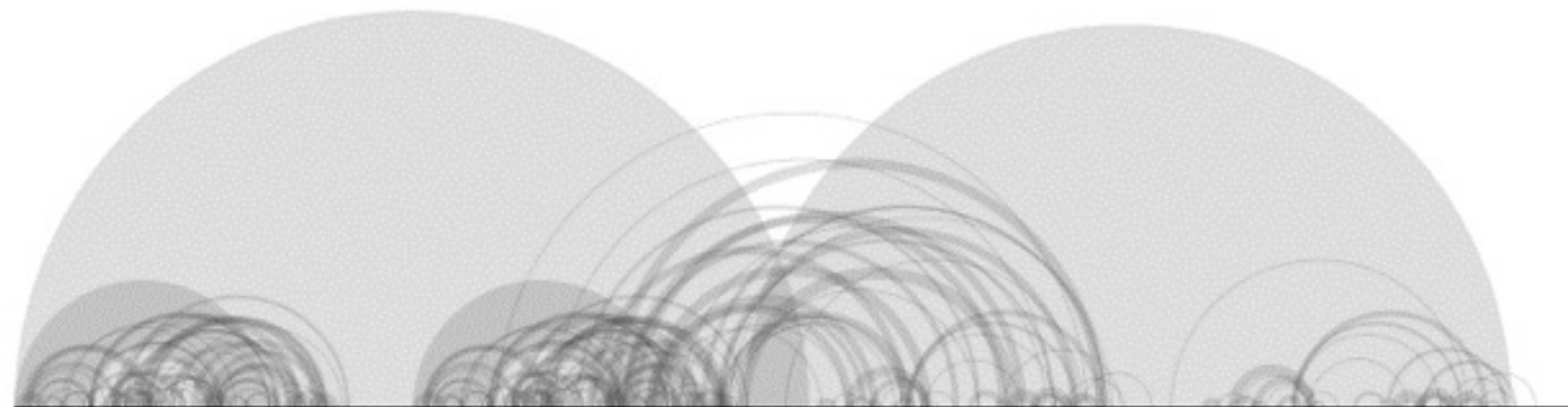
Arc Diagrams

- Comparison to a dotplot diagram
- Recap Matrix diagram
 - Correlation matrix
 - String of n symbols a_1, a_2, \dots, a_n is represented by an $n \times n$ matrix
 - Pixel at coordinate (i, j) is black if $a_i = a_j$
 - Can handle very large datasets
 - Shows both small and large-scale structures
- Heavy clutter caused by small substrings with high frequency: n repetitions of a substring lead to n^2 visual marks
- Arc Diagrams mark only similar substrings, which are subsequent



Arc Diagrams

- Applied to music, Minuet in G Major, Bach
- Shows classic pattern of a minuet: two main parts, each consisting of a long passage played twice
- Parts are loosely related: bundle of thin arcs connecting the two main parts
- Overlap of the two main arcs shows that the end of the first passage is the same as the beginning of the second passage



Visualizing Plagiarism

- Ribler & Abrams 2000
- Problem: programming assignment in a class with large number of students
- High probability of plagiarism
- Need to compare every document (code file) with every other document
- Visualization must support two steps
 - Highlight suspicious documents
 - Allow for detailed examination of the similar passages - high level of similarity between documents may not be due to cheating (e.g., headers)

Visualizing Plagiarism

- Categorical Patterngram
- Visualize frequencies of sequences of characters present in more than one document
- Remove all non-printable characters in the document collection
- Define length of character sequence to analyse (in the example: 4)
- Histogram-like approach
 - X-axis: start character of sequence
 - Y-axis: number of documents containing the sequence
 - Doc at Y = 1: base document to compare against all other documents

Toy0: This is a test.

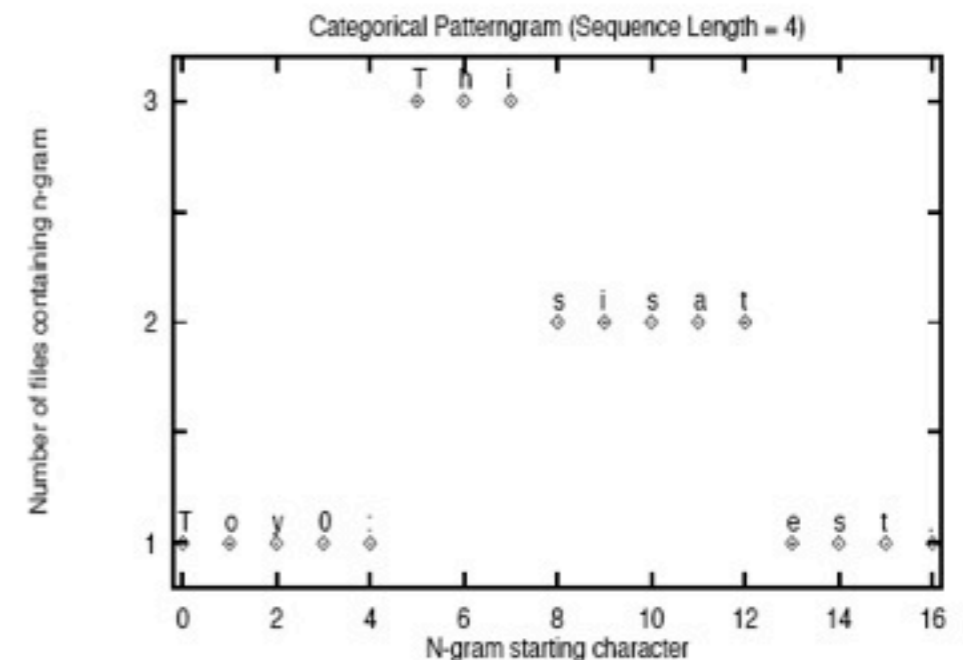
Figure 1. Toy File 0

Toy1: Oh yes. This is a test too.

Figure 2. Toy File 1

Toy2: Toy2 has little in common with the other two.
This is common.

Figure 3. Toy File 2



Visualizing Plagiarism

- Composite Categorical Patterngram
- Visualizes which particular documents are similar
- Y-axis: each value corresponds to an individual document

Toy0: This is a test.

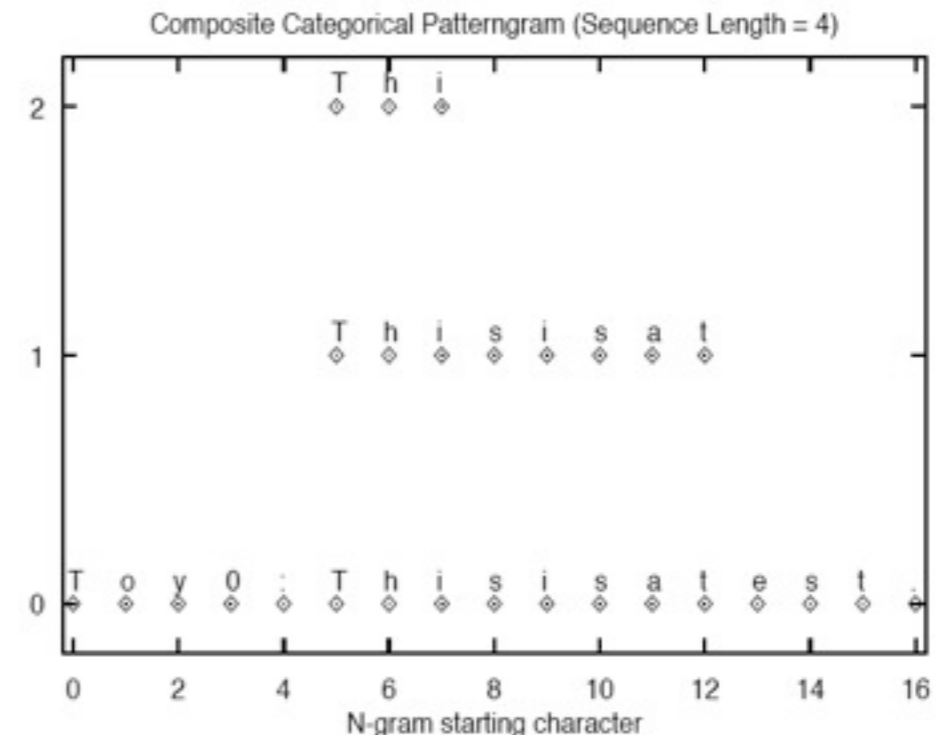
Figure 1. Toy File 0

Toy1: Oh yes. This is a test too.

Figure 2. Toy File 1

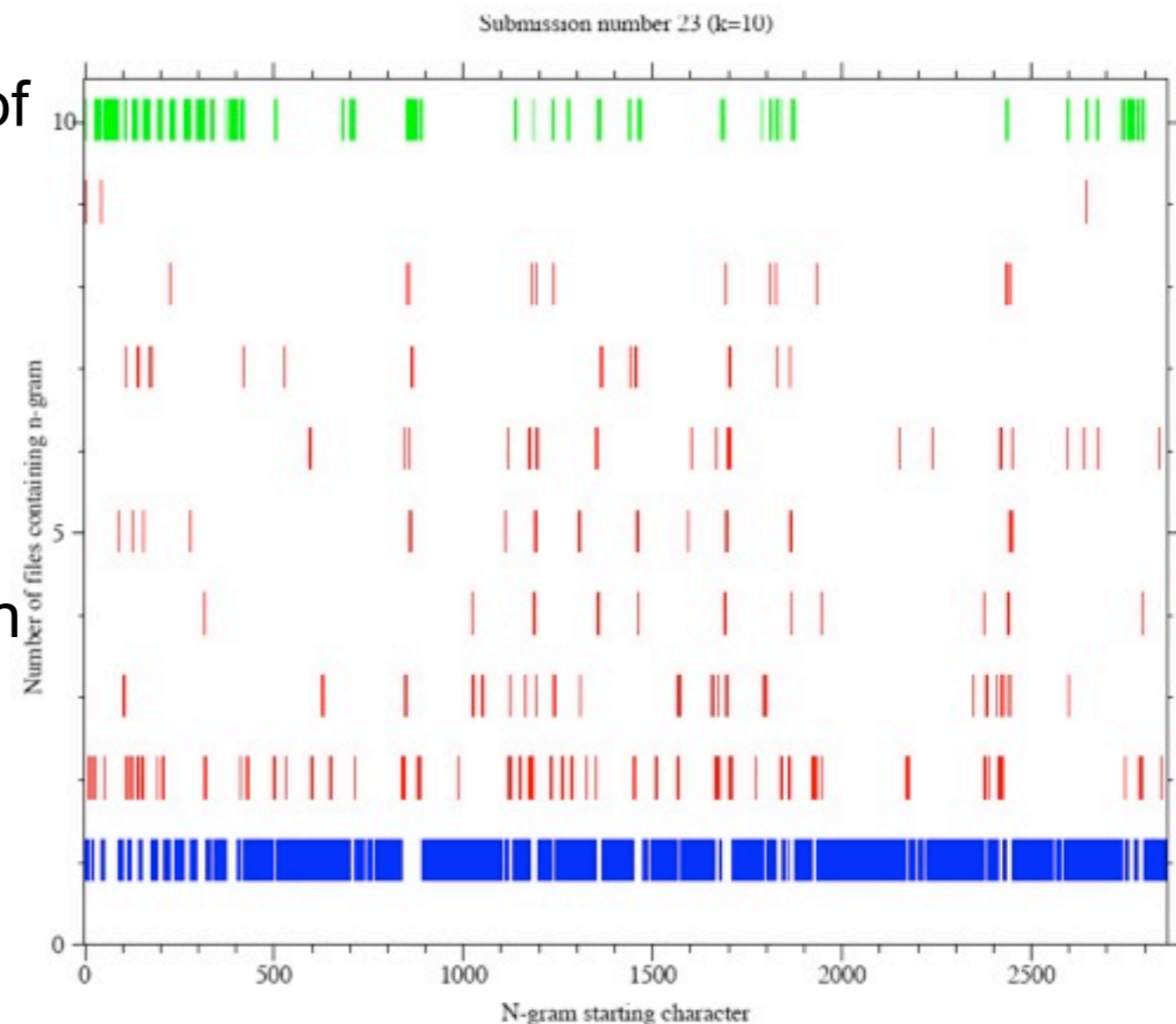
Toy2: Toy2 has little in common with the other two.
This is common.

Figure 3. Toy File 2



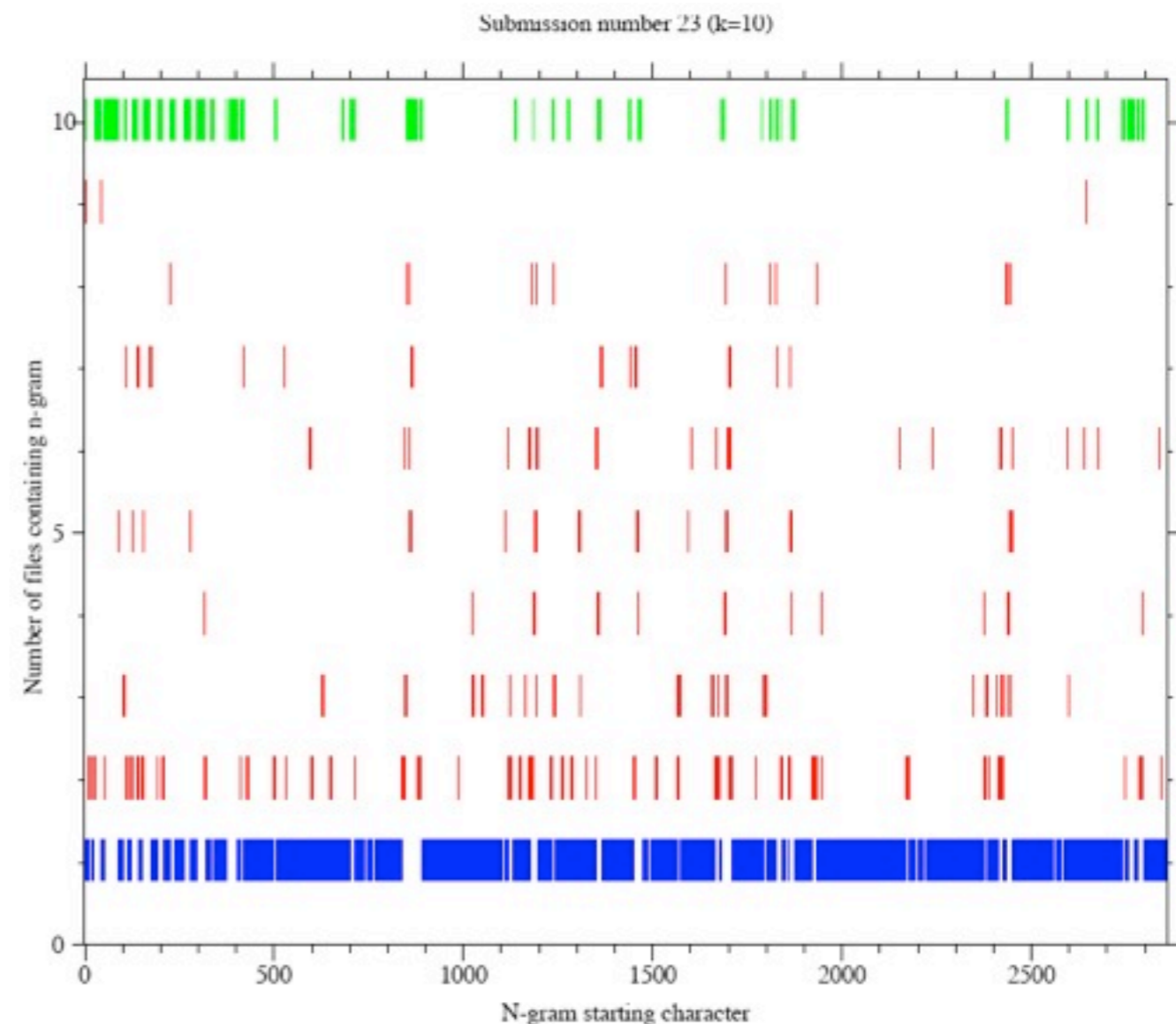
Visualizing Plagiarism

- Case study
- Students were asked to extend a sample program of about 30 lines of code
- Average completed program was about 150 lines
- Submission via email
- Graphic shows categorical patterngram for a single submission
 - Sequence length = 10
 - Lines not text due to high density
 - Rather confusing color coding
- Color coding (not very reasonable)
 - Green: frequency ≥ 10
 - Red: frequency < 10
 - Blue: base document
- Plagiarism or not?



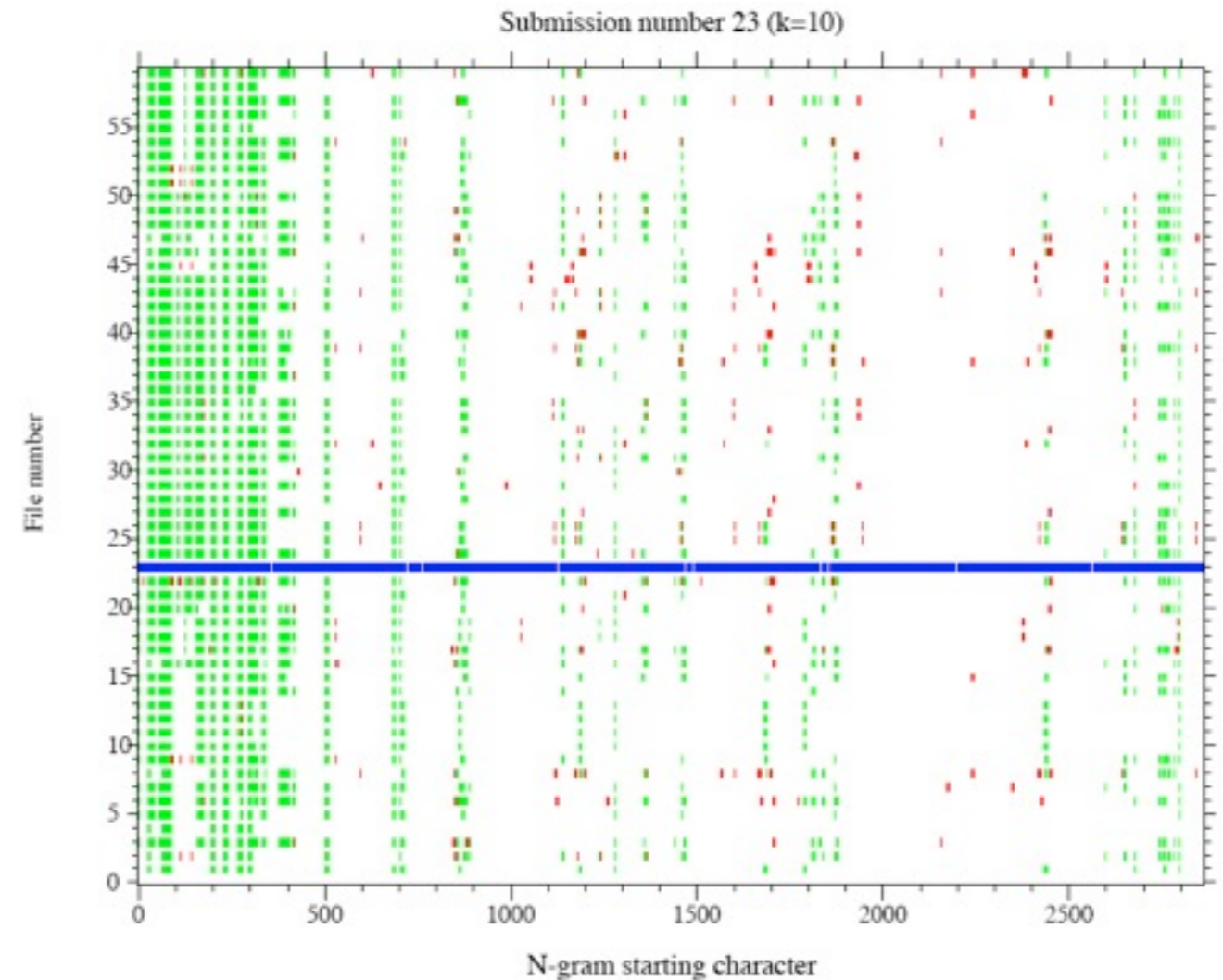
Visualizing Plagiarism

- What to look out for?
 - Sequences that occur frequently are not of interest - all points with $y \geq 10$ are plotted as $y = 10$
 - Suspicious: accumulation of points with low frequencies
- Analysis
- Majority of points are plotted at $Y = 1$
- Hence most 10-char sequences are unique to the base document
- Number of points plotted at $Y = 2$, but evenly distributed



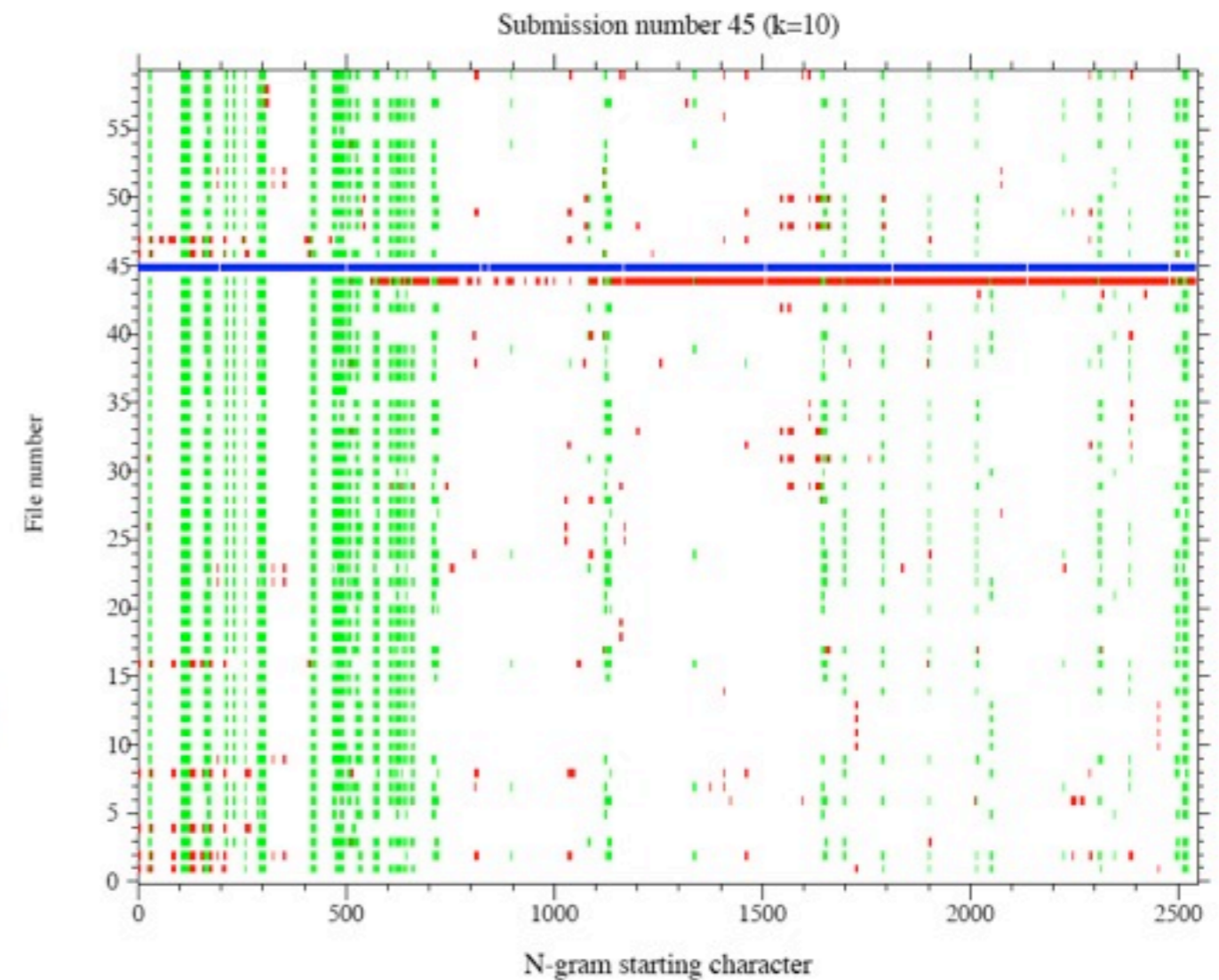
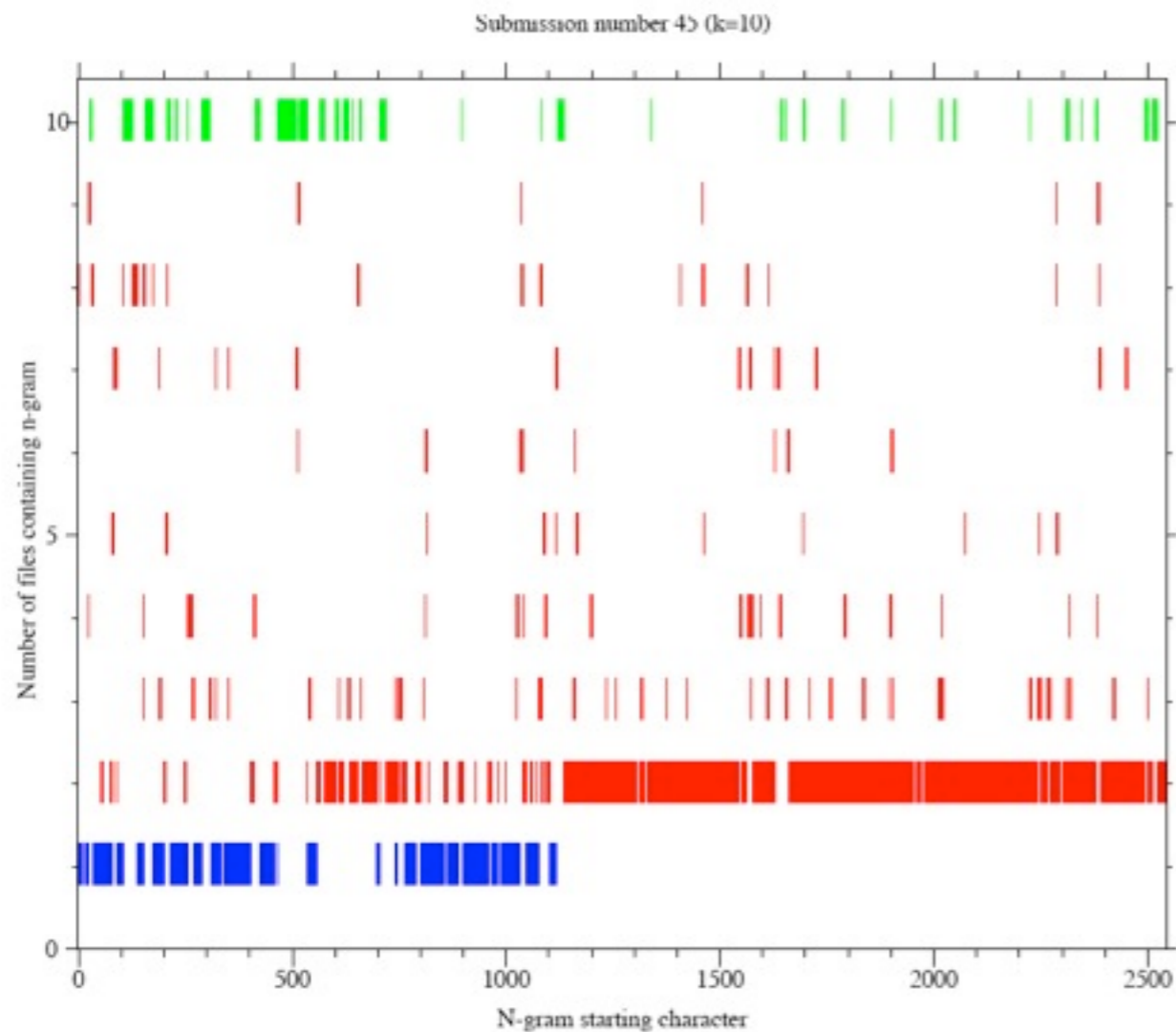
Visualizing Plagiarism

- Composite Categorical Patterngram for the submission
- Solid line represents the base document (submission number 23)
- Large number of points plotted in the range of $x = [0; 500]$: email message header
- Other frequent sequences due to the sample program
- Pattern typical for independent work



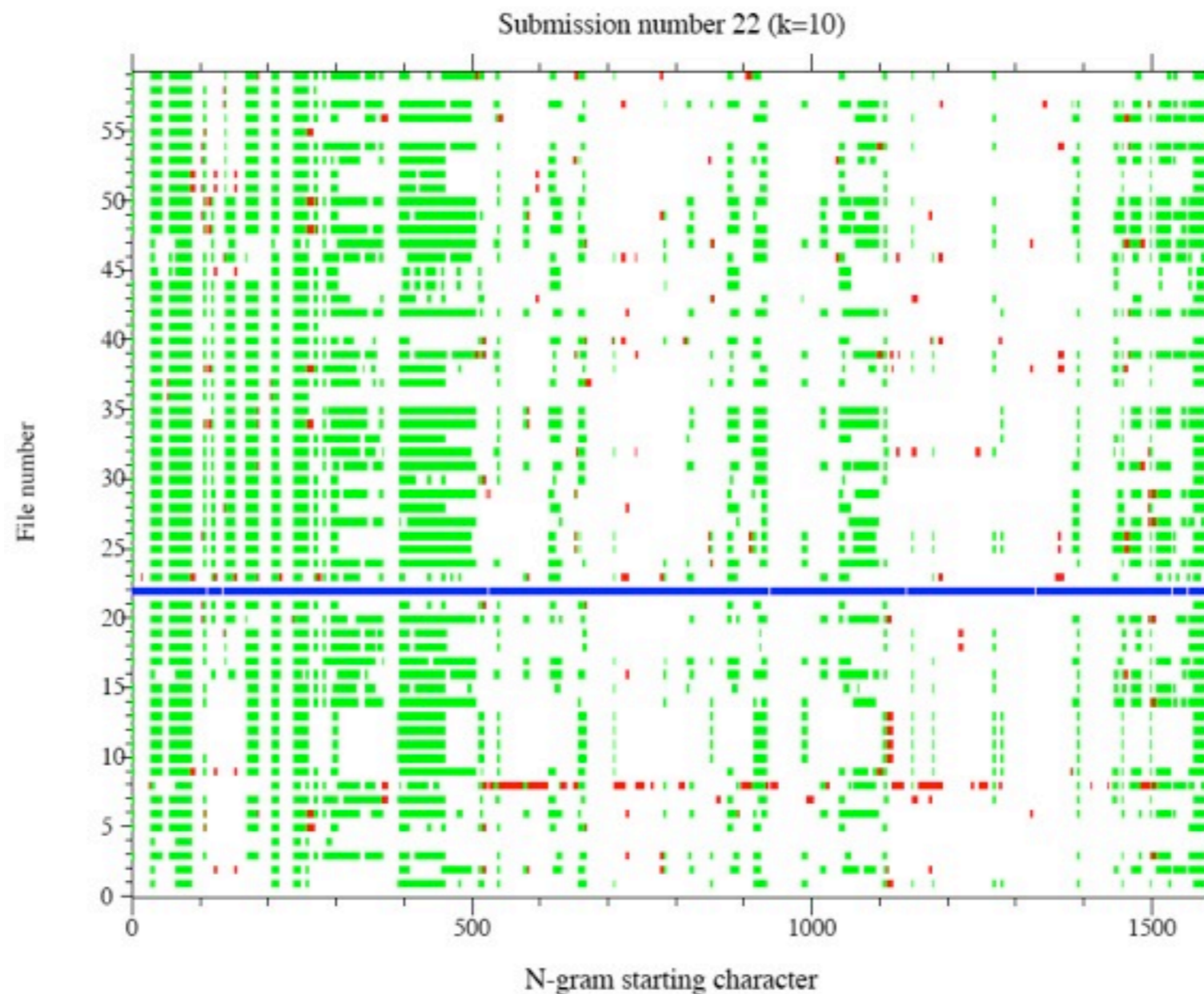
Visualizing Plagiarism

- Example of patterngrams indicating extensive plagiarism



Visualizing Plagiarism

- Patterngram of more subtle plagiarism

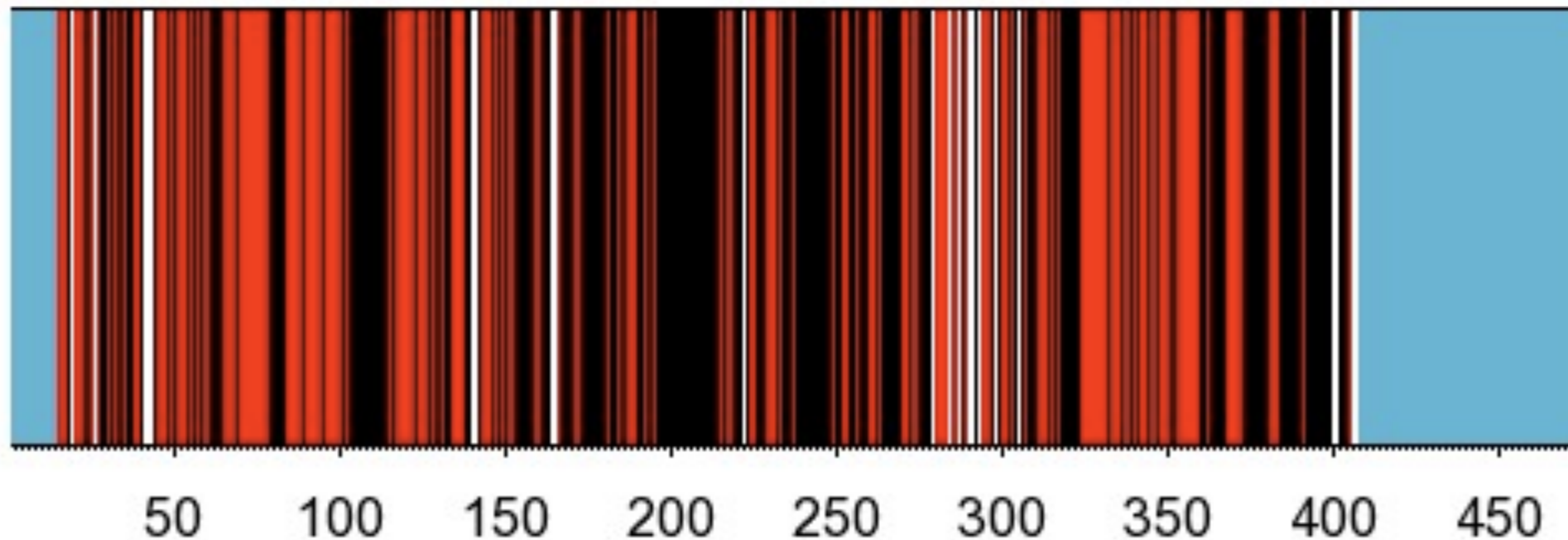


Visualizing Plagiarism

- What may a student do to mask plagiarized code
- Change variable names
- Minimize masking effect by replacing all alphanumeric strings in all documents into single characters
- Two documents with the same code but different variable names will produce identical patterngrams

A more recent example [guttenplag.wikia.com]

1218 Plagiatsfragmente aus 135 Quellen
auf 371 von 393 Seiten (94.4%)
in 10421 plagiierten Zeilen (63.8%)



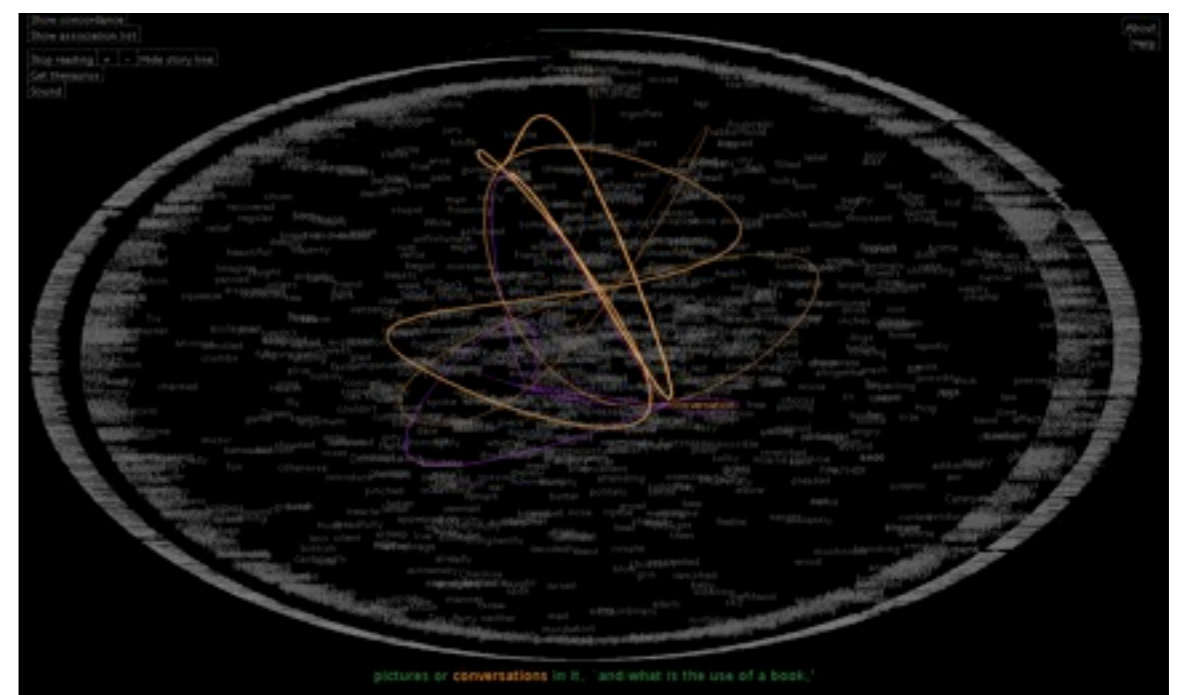
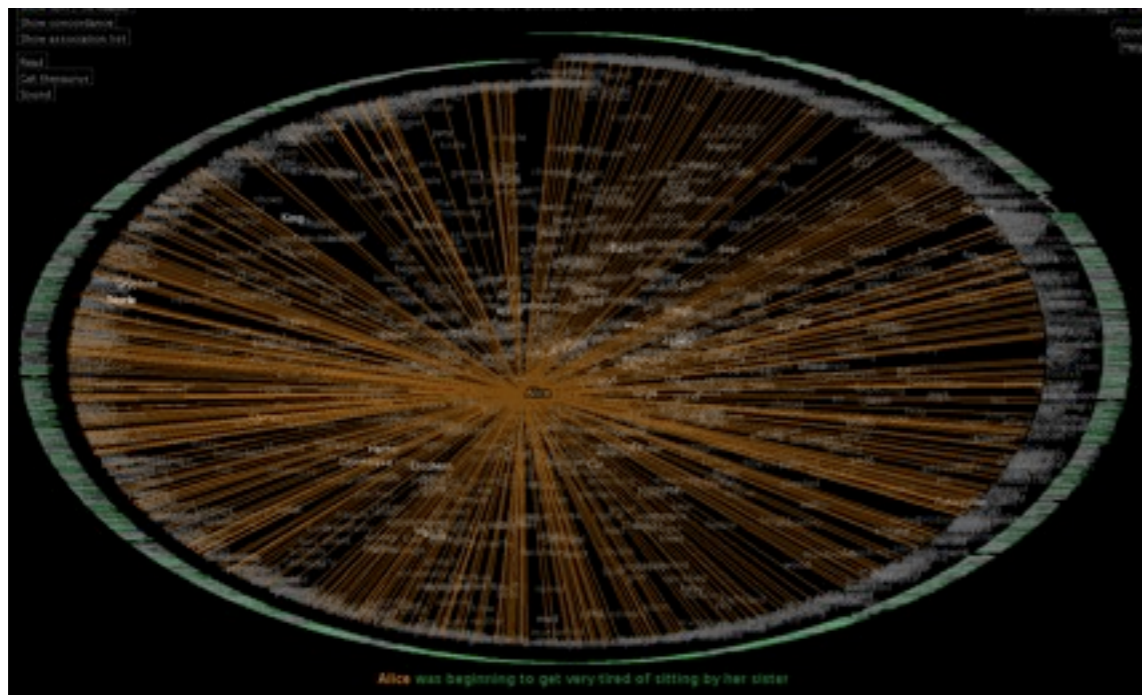
Stand: 03.04.2011 11:55 Uhr

- Seiten, auf denen Plagiate gefunden wurden
- Seiten mit Plagiaten aus mehreren Quellen
- Seiten, auf denen bisher keine Plagiate gefunden wurden
- Das Inhaltsverzeichnis (Seiten 1-14) und die Anhänge (ab Seite 408) wurden nicht bei der Berechnung des Prozentualwertes mit einbezogen



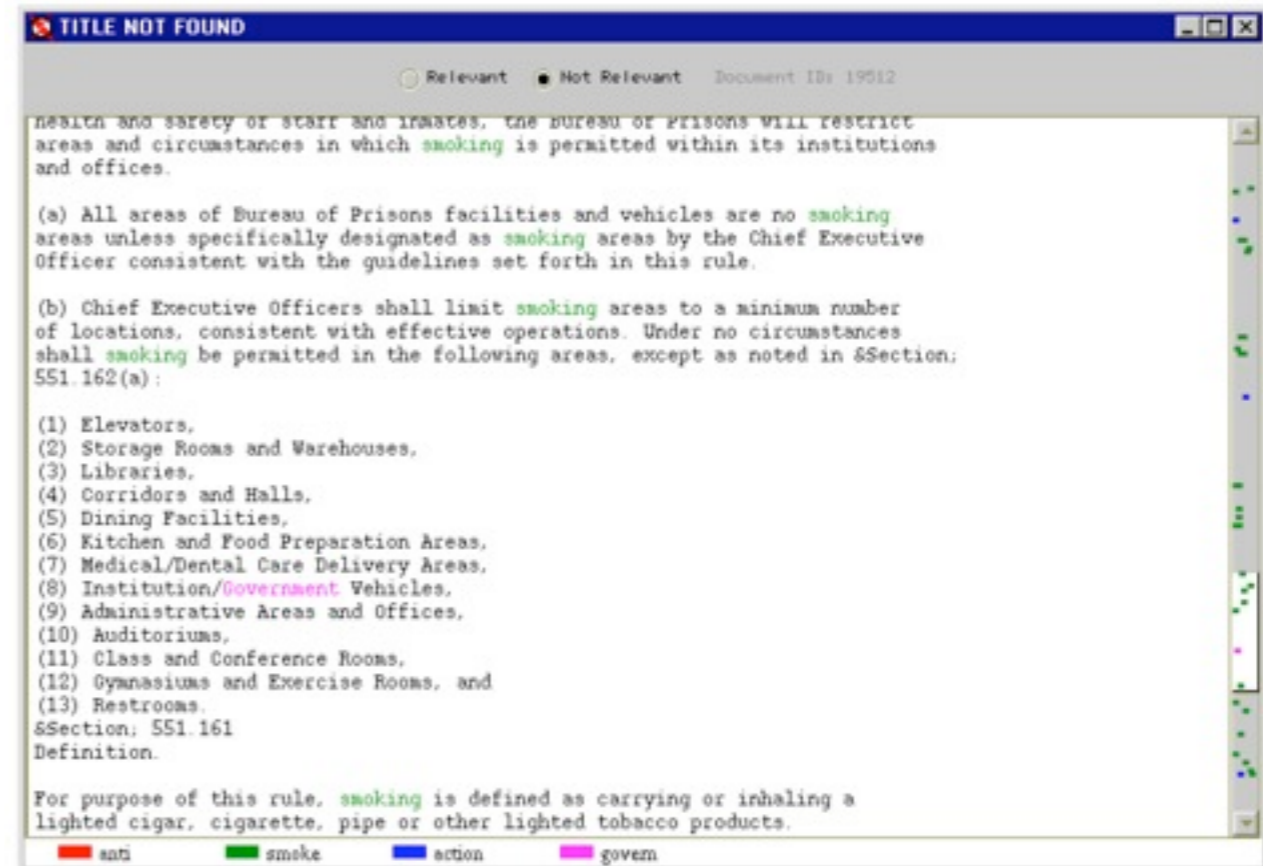
TextArc

- <http://www.textarc.org/> - demo
- Represents the entire text as 1 pixel lines in an outer circle
- Text is revealed via mouse-over
- Words are repeated in inner circle at a readable size
- Position of the words depend on where the word appears in the document
- Words that appear throughout the novel will be drawn to the center
- Frequent words stand out
- Example visualizes the novel “Alice in Wonderland”
- Various visualization features



Search Terms on a Scrollbar

- Byrd 1999
- Searching of keywords in a single document
- Color coding to map each occurrence of a keyword in the document as a small colored icon in the scrollbar
- Provides an overview of the entire document, not only of the portion currently visible
- Users can directly jump to keyword occurrences by moving the slider thumb



TileBars

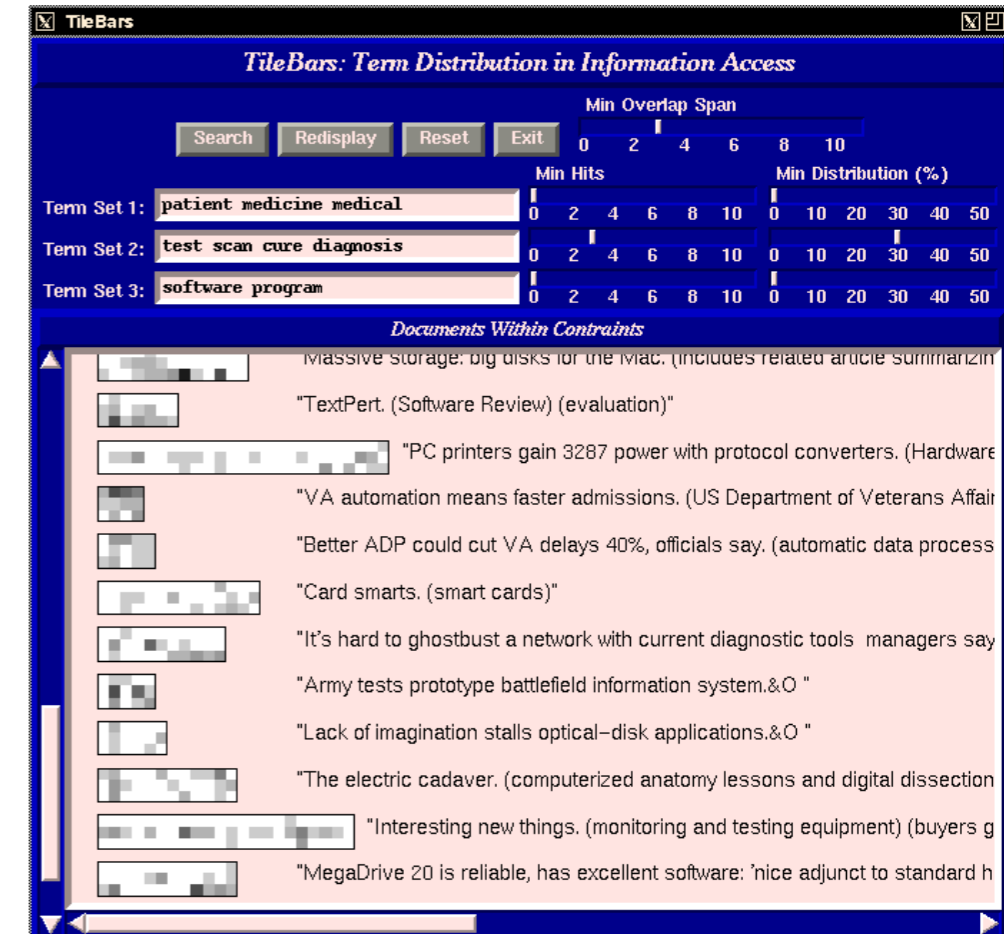
- Hearst 1995
- Problem with document ranking of common search engines?
- Ranking approach is opaque:
 - What role did the query terms play in the ranking process
 - What is the relationship between the query terms in the document
- TileBars attempts to let the users make informed decisions about which documents and passages to view



The screenshot shows a Google search interface. The search bar contains the text "visualization text ranking". Below the search bar, there are radio buttons for "Web-Suche" (selected) and "Suche Seiten auf Deutsch". To the right of the search bar is a "Suche" button and a link for "Erweiterte Einstellungen". Below the search bar, the results are displayed as "Web Ergebnisse 1 - 10 von ungefähr 334.000 für visualization text ranking. (0,17 Sekunden)". A tip suggests searching only in German. The first result is titled "Data Mining: Text Mining, Visualization and Social Media: Video ..." and includes a link to "Diese Seite übersetzen". The snippet for this result mentions TechCrunch and the launch of video ranking. The second result is titled "Data Mining: Text Mining, Visualization and Social Media: ranking" and includes a link to "Diese Seite übersetzen". The snippet for this result discusses influential ideas and blogs/bloggers. The third result is titled "Seminararbeit: User Interfaces and Visualization for ..." and includes a link to "HTML-Version". The snippet for this result mentions a seminar at the University of Würzburg. The fourth result is titled "BioMed Central | Full text | PCOGR: Phylogenetic COG ranking as an ..." and includes a link to "Diese Seite übersetzen". The snippet for this result mentions a graphical visualization of COGs.

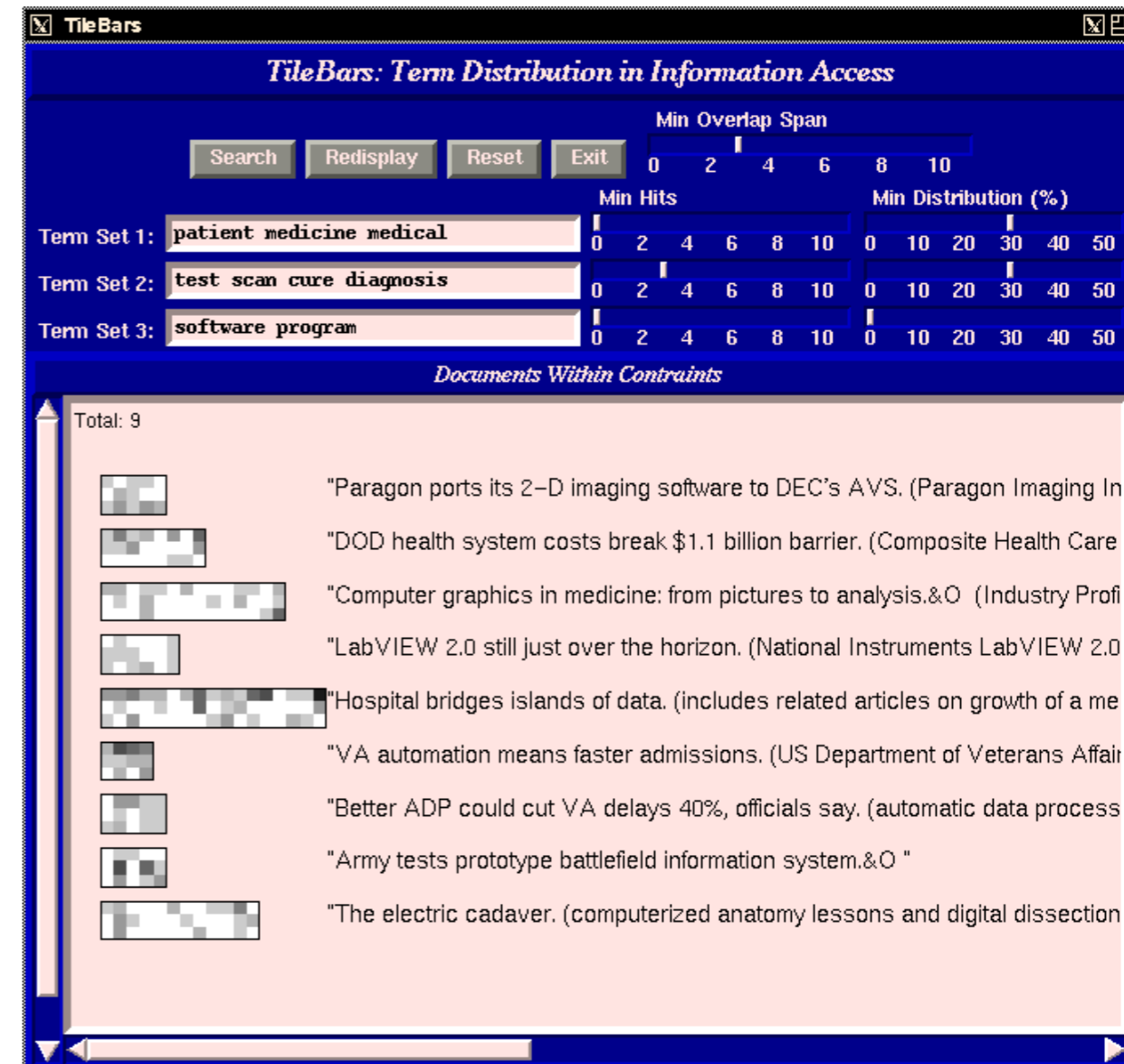
TileBars

- Users provide sets of query terms
 - OR within a set
 - AND between sets
- Documents are partitioned into adjacent, non-overlapping multi-paragraph segments
- Each document of the result set is represented by a rectangle - width indicates relative length of the document
- Stacked squares correspond to text segments
- Each row of the stack corresponds to a set of query terms
- Darkness of the square indicates the frequency of terms from the corresponding term set - (Why is this a reasonable color mapping?)
- Title + initial words appear next to each document
- Users can click on segments to retrieve the corresponding text



TileBars

- Analysis hints
 - Overall darkness indicates that all term sets are discussed in detail throughout the document
 - When terms are discussed simultaneously the tiles blend together causing an easy to spot block
 - Scattered term set occurrence show large areas of white space
 - Helps to distinguish between passing remarks and prominent topic terms
- Users may also set distribution constraints to refine the query
 - Minimum number of hits per term set
 - Minimum distribution (percentage of tiles containing at least one hit)
 - Minimum adjacent overlap span



Cluster Maps

- Downscaling of n-dimensional document space to 2D
- Map of a document collection
- Similar documents are placed close to each other
- Dissimilar documents are placed farther apart from each other
- Provide thematic overview for exploration (same concept as product arrangements in a store)
- How to - Vector space model and map construction
 - Create inverted index of document collection
 - Exclude stop words and the most frequent words (“and” may not be a good discriminator of content)
 - Matrix of indexing words versus documents gives you document vectors
 - A document vector reflects the frequency of index words occurring in the document

Cluster Maps

- How to - Vector space model and map construction (continued)
 - Compute similarity between pairs of documents (e.g. dot product of vectors)
 - Layout documents in 1D/2D/3D
- Common approaches
 - Spring model of graph layout
 - Multi-dimensional scaling
 - Clustering (e.g. hierarchical)
 - Self-organizing maps (SOM aka Kohonen map)

Document vectors

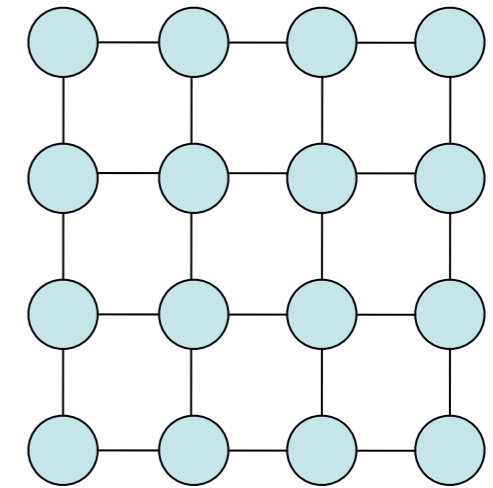
	Doc 1	Doc 2	Doc 3
“Artificial”	1	2	0
“Creativity”	2	1	0
“Java”	0	0	3

Similarity Matrix

	Doc 1	Doc 2	Doc 3
Doc 1	1	0.66	0
Doc 2	0.66	1	0
Doc 3	0	0	1

SOM

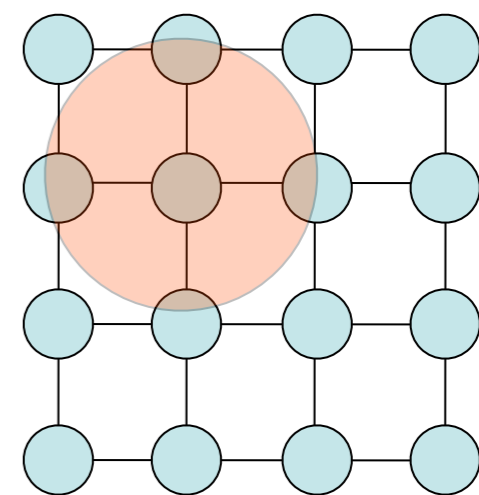
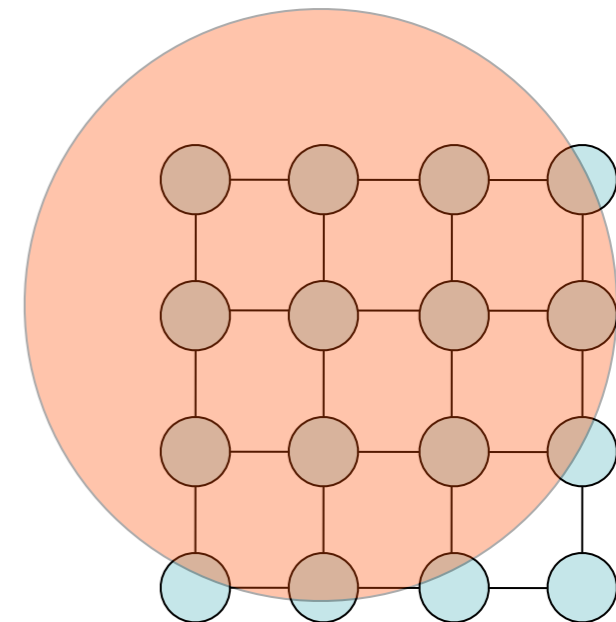
- Unsupervised learning algorithm
- SOM map is formed from a regular grid of neurons (nodes)
- Each node has
 - An $x y$ coordinate in the grid
 - A weight vector of the same dimensionality as the input vectors
- Input vectors
 - Used to train the map
 - Represent collection of objects
- In case of visualizing text, input vectors are usually equal to document vectors



Network of 4x4 nodes

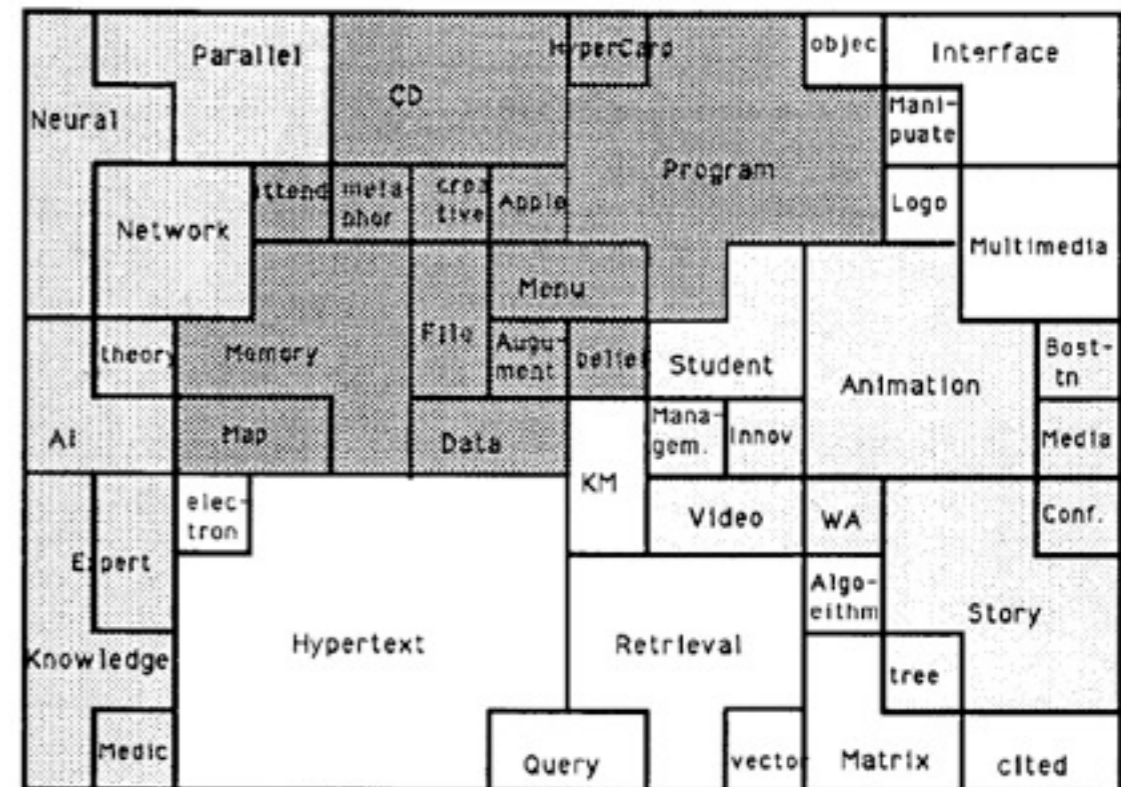
SOM - Algorithm

- 1. Start with assigning small random weights to the nodes of the grid
- 2. Chose a vector at random from the set of input vectors and present it to the grid
- 3. For each node: calculate the Euclidean distance between each node's weight vector and the current input vector - the closest node is called the Best Matching Unit (BMU)
- 4. Calculate the radius of the BMU (radius diminishes with each time-step)
- 5. For each node within the radius of the BMU: adjust the weights to make them more similar to the input vector - the closer a node is to the BMU, the more its weights get altered
- 6. Repeat step 2 for N iterations
- When training is completed each document is assigned to its BMU



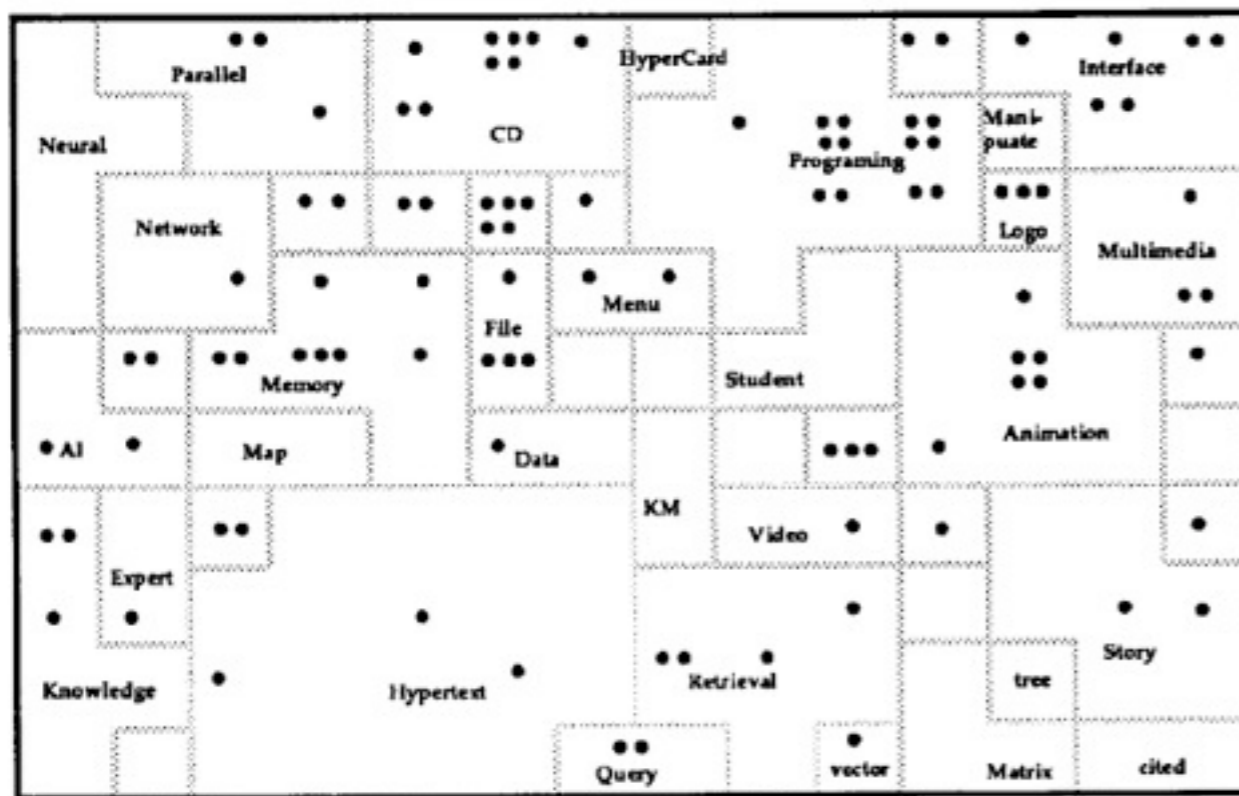
Cluster Maps

- Lin 1992
- Personal collection of 660 research documents
- 2500 learning iterations
- Labeled word show most frequent title words
- Size maps to frequencies of occurrence of the words
- Neighboring relationships of areas indicate frequencies of the co-occurrence of words

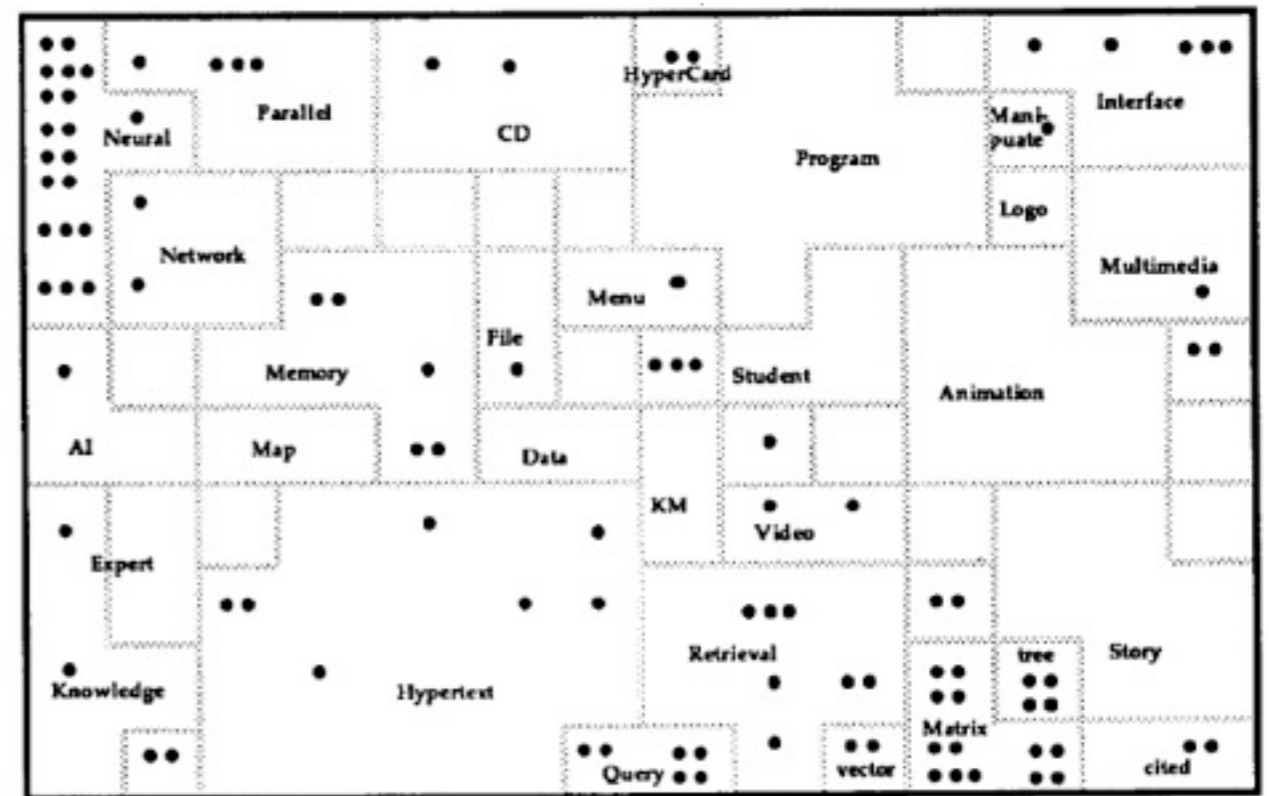


Cluster Maps

- Research interest changing over time



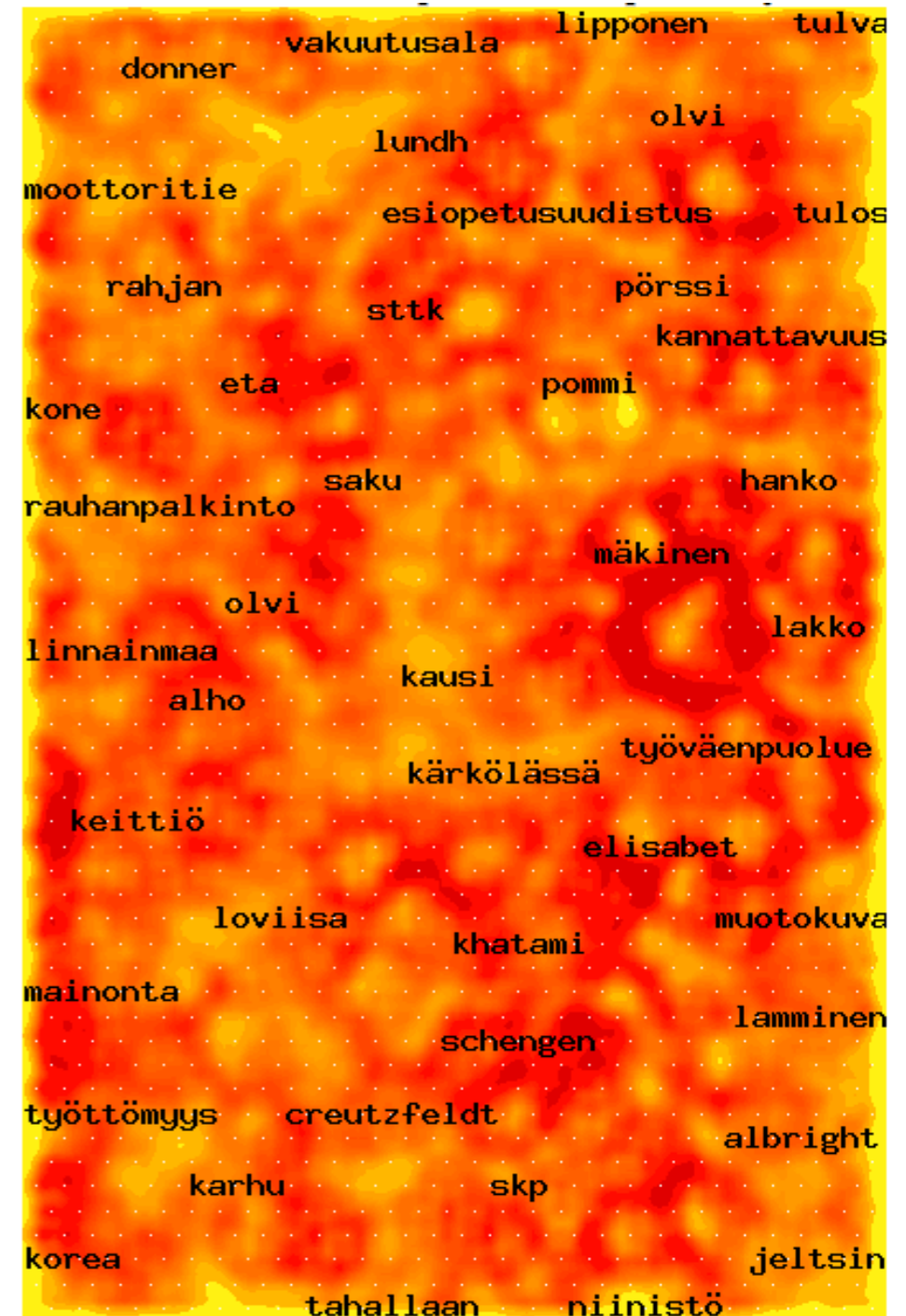
(a) Distribution of the first 100 documents in the personal collection



(b) Distribution of the latest 100 documents in the personal collection

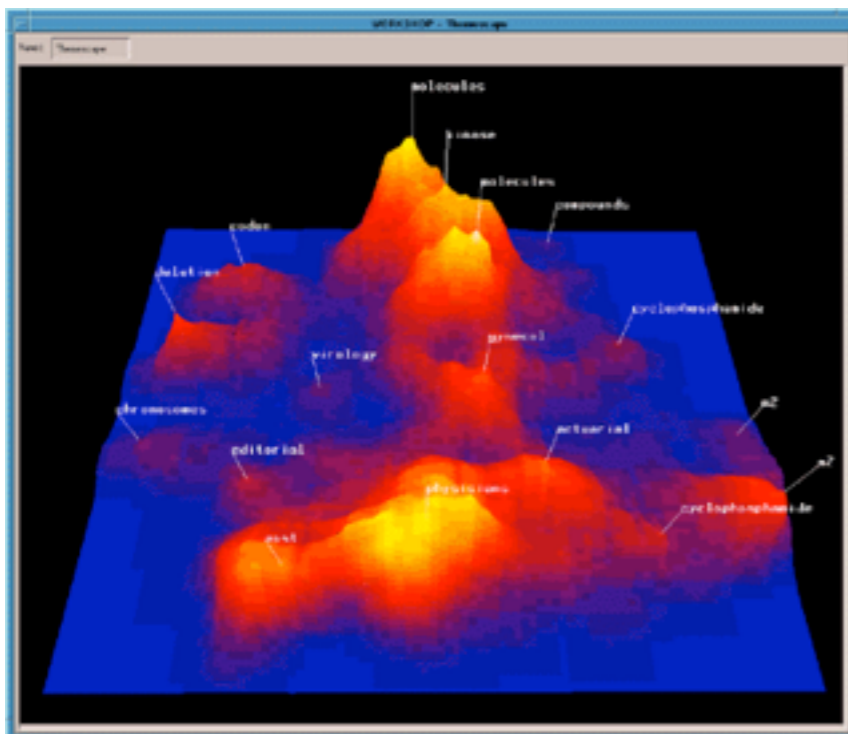
WEBSOM

- <http://websom.hut.fi/websom/>
- SOM of Finnish news bulletins for exploring and retrieving documents
- Labels show the topics of areas in the SOM
- Coloring encodes density - light areas contain more documents
- Navigation via zooming and panning
- Documents can be retrieved on the lowest level of the visualization

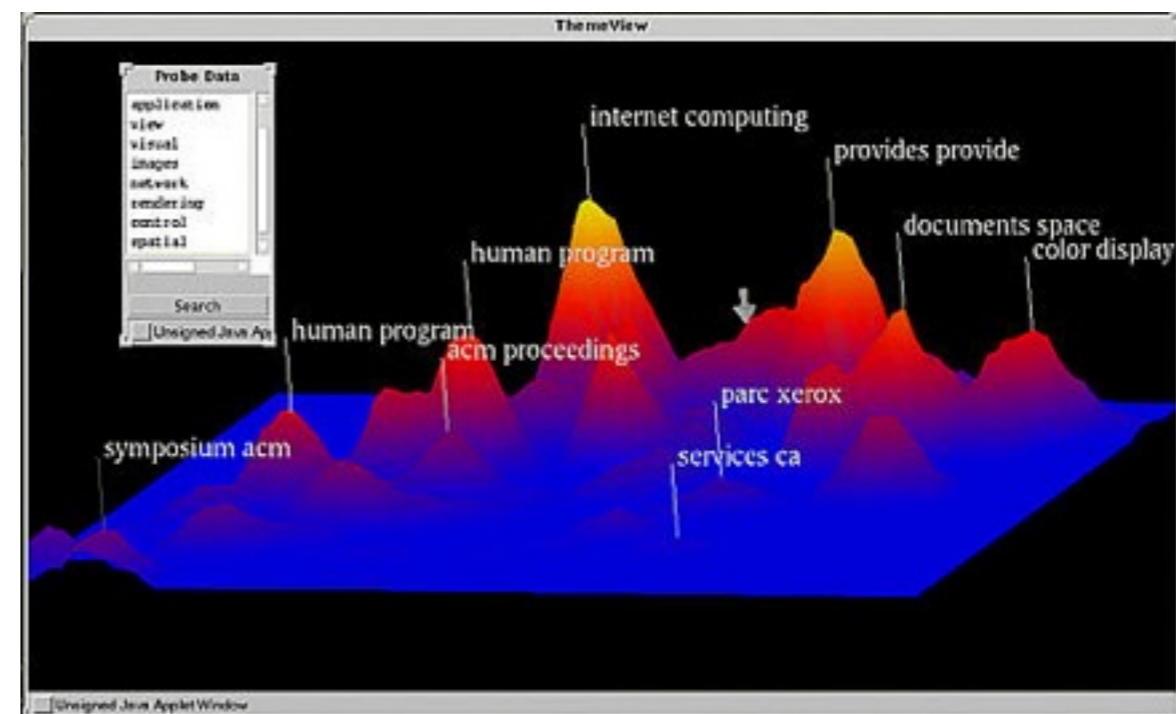


ThemeScapes

- Wise et al. 1995
- Map document density to third dimension
- News article visualized as an abstract 3D landscape
- Mountains represent frequent themes in the document corpus (height proportional to number of documents relating to the theme)
- Spatial characteristics of the map should map to interconnections of themes



<http://nd.loopback.org/hyperd/zb/spire/spire.html>



<http://infoviz.pnl.gov/technologies.stm>

Cluster Map vs Keyword Search

- Chris North
- Cluster Map pros
 - Facilitates non-targeted exploration and browsing by spatially organizing documents
 - Provides overview of document set: major themes, sizes of clusters, relationships between themes
 - Scales up
- Cluster Map cons
 - How to label groups?
 - What does the space mean? How to label space?
 - Where to locate documents with multiple themes: both mountains, between mountains, ...?
 - Relationships within documents?
 - Algorithm (SOM) is time-consuming

Cluster Map vs Keyword Search

- Chris North
- Keyword search pros
 - Reduces the browsing space according to user's interests
- Keyword search cons
 - What keywords do I use?
 - What about other related documents that don't use these keywords?
 - No initial overview
 - Mega-hit, zero-hit problem

TagClouds

- Show the frequency of words in a text
- Frequency is mapped to size and/or color
- Often found as navigation aid on web pages
- example below generated by www.wordle.net

