

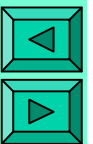


The
University
of Auckland

Experimental CBIR Systems

COMPSCI.708.S1.C

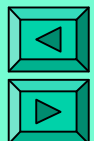
A/P Georgy Gimel'farb





Classification of CBIR Systems

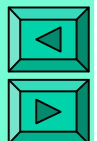
- **Level of automation** of feature extraction and index generation
- **Level of integration** of multimedia modalities
- **Level of adaptability** to the needs of users and applications
- **Level of abstraction** in which content is indexed
- **Level of generality** of visual information domain
- **Level of automation** of the database collection
- **Abilities to semantically categorise** visual information
- **Abilities to process compressed visual information**





Level of Automation

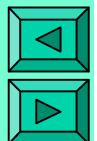
- Interactive content-based querying: on the quantitative features of extracted and indexed images/videos
 - Low-level features (colours, textures, shapes, and motions) are extracted by automatic methods
 - Generation of higher-level semantic indexes usually requires human input and/or system training
- In a semiautomatic system the user selects manually image objects and features which are then used by the system for generating the feature indexes





Integration / Adaptability

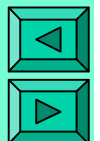
- Multimedia modalities: images, video, graphics, text, audio → typically, independent indexing
 - Integration of multiple modalities: under investigation but it is not yet fully exploited
- Most systems use a static set of previously extracted features included into image metadata and indexes
 - The features are selected by the designer on the basis of indexing costs and search functionalities
 - Dynamic features extraction and indexing is needed to adapt to subjective and changing needs of users





Level of Abstraction

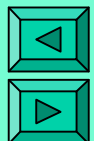
- Level of abstraction in which content is indexed:
 - ***low-level feature level*** (colour, texture, shape)
 - ***object level*** (moving foreground item)
 - ***syntactic level*** (video shot)
 - ***high-level semantic level*** (image subject)
 - **automatic indexing**: mostly low-level features
 - **higher-level indexes**: generated manually
- Interaction between the levels: still the unsolved problem





Generality / Data Collection

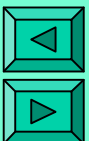
- Generality of the visual information domain:
 - specific domain knowledge: special feature sets (e.g., medical and remote-sensing applications)
 - generally: indexing unconstrained visual information, such as that on the Web
- Collection of the system's database:
 - **dynamic CBIR systems**: special software *robots* that automatically traverse the Web
 - **other systems** (*online news archives, photo stock houses, etc.*): manual addition of visual information





Categorisation / Processing

- Semantic classes of visual information
 - Navigation through the classes in large repositories of visual information is very useful
 - Effective categorisation is not yet developed
- **Compressed Data Processing:** direct extraction of the features from compressed images and videos in order to avoid expensive data decompression
 - compression standards (JPEG, MPEG) → features can be computed in the compressed domain

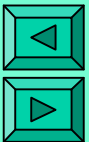




The
University
of Auckland

IBM: CBIR System "QBICTM"

- Query by Image Content System
 - "...When keywords alone cannot locate that special "something" to fit a specific taste, users can turn to IBM's patented Query By Image Content or QBIC, a new search engine that sorts through database images according to colors, textures, shapes, sizes and their positions" **(from the IBM's description)**
- The first CBIR system exploring feature-based image and video retrieval

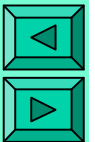
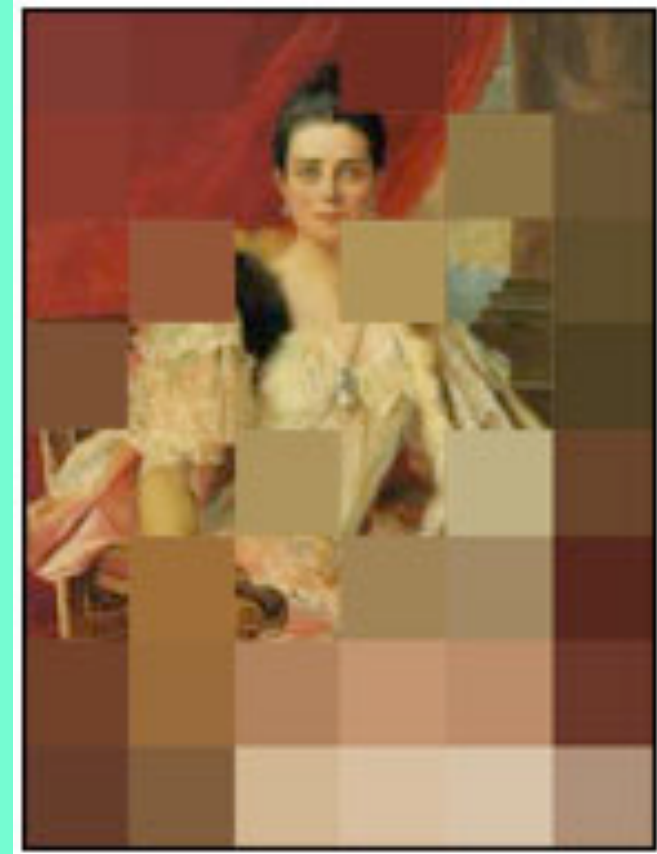




The
University
of Auckland

QBICTM in the Hermitage

- The Hermitage (Russian Museum)
- **Browsing** of art categories: paintings, drawings, sculptures, arms, etc
- **Advanced Search:** by artist, title, subject, style, genre, theme, or date
- **QBIC:** to refine the search by requesting all artwork with comparable visual attributes





Colour Search Interface in QBIC

QUICK SEARCH

- BROWSE +
- QBIC SEARCHES +
- COLOUR SEARCH -
- LAYOUT SEARCH -
- ADVANCED SEARCH +

QBIC COLOUR SEARCH

The interface includes a color palette, a bucket for color selection, and RGB sliders. The bucket is currently filled with a blue color. The RGB sliders are set to R=128, G=255, and B=252. There are buttons for 'delete', 'clear all', and 'search'.

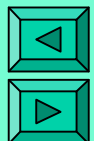
1. Use your mouse to select a colour from the palette.

2. Click the arrow button to add the colour to the bucket.

3. Slide the triangular handles on the bucket to adjust the percentage of this colour.

4. You may repeat this process until the bucket is full. When you are ready, click Search.

You may also use the Colour Mixer to adjust RGB (red, green, blue) values to use in your search.





Colour Search Interface in QBIC

QUICK SEARCH

BROWSE •

- QBIC SEARCHES •
- COLOUR SEARCH -
- LAYOUT SEARCH -
- ADVANCED SEARCH •

QBIC COLOUR SEARCH

The interface shows a color palette with a crosshair. To the right is a bucket with a stack of colored blocks. Below the bucket are three sliders for Red (R), Green (G), and Blue (B) with numerical values. At the bottom are buttons for 'delete', 'clear all', and 'search'.

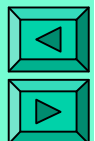
1. Use your mouse to select a colour from the palette.

2. Click the arrow button to add the colour to the bucket.

3. Slide the triangular handles on the bucket to adjust the percentage of this colour.

4. You may repeat this process until the bucket is full. When you are ready, click Search.

You may also use the Colour Mixer to adjust RGB (red, green, blue) values to use in your search.





Layout Search Interface in QBIC

QUICK SEARCH

BROWSE •
QBIC SEARCHES •
COLOUR SEARCH -
LAYOUT SEARCH -
ADVANCED SEARCH •

QBIC LAYOUT SEARCH

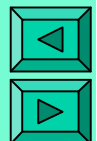
1. Use your mouse to choose a colour from the palette.

2. Select either the round tool or the square tool.

3. Hold down your mouse button and drag the cross on the canvas to create a coloured shape.

4. Repeat this process until you complete your custom layout. When you're ready, click Search.

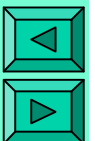
To perform other actions, click the shape to make it active. Drag the edges to Resize. Click Send to Back and Bring to Front to layer shapes. Click Delete to remove a shape. Click Clear All to empty the layout.





The QBIC Prototype System

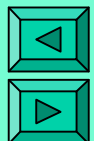
- Two data types: a scene (image or video frame) and an object (a part of a scene)
- Objects are outlined manually and characterised by:
 - colour descriptors: average colour and histogram with 64 or 256 elements of the quantised colour space
 - modified Tamura's texture descriptors: coarseness, contrast, and directionality
 - shape descriptor (moments, heuristics, parametric splines with Hausdorff distance between the curves)





QBIC: Object / Scene Retrieval

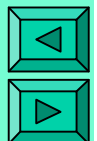
- All objects in the images of a given database have to be outlined and features of all images and objects have to be computed at the stage of preparing the database for queries to be processed. Each query can be based on objects, entire scenes, or a combination of both.
- **Object-based query:** images containing objects with certain features such as colour percentage, colour layout, or/and texture.
- **Scene-based query:** full scenes with certain features (because the objects need not be outlined, the query interface is simpler)





QBIC: Object / Scene Retrieval

- Normalised similarity (distance) functions for each feature to be meaningfully combined
- **Weighted Euclidean distance** \Rightarrow for 3-component average colour vectors, 3-component texture vectors, and 20-component shape vectors
 - Weights - the inverse variances of each component in the database
- **Quadratic distance** $(\mathbf{X}-\mathbf{Y})^T \mathbf{S} (\mathbf{X}-\mathbf{Y}) \Rightarrow$ to match colour histograms \mathbf{X} and \mathbf{Y}
 - \mathbf{S} : a symmetric colour dissimilarity matrix with the components $S(i,j)$ indicating the dissimilarity of the colours i and j in the histogram

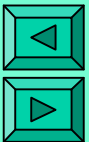




The
University
of Auckland

QBIC: Object / Scene Retrieval

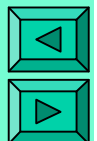
- Pre-computed stored features are compared to a given query in order to determine which images match it
- **Filtering and R-trees** \Rightarrow to index the feature vectors in the database
 - 64-element colour histogram: access by filtering
 - 3D average colour: access by an R-tree
 - 20-dimensional moment-based shape vector \Rightarrow PCA to map onto a 3D space to use an R-tree index





MIT Media Lab: Photobook

- Queries based on features related to particular models fitted to each image
- Commonly: colour, texture, and shape models, although features from any model may be used (e.g., results of face recognition)
- Features comparisons with a library of matching algorithms:
 - Euclidean or Mahalanobis distance
 - Kullback-Leibler divergence
 - Vector space angle (correlation)
 - Histogram distance
 - Peaks in Fourier space
 - Wavelet tree distance, or
 - User-defined ones, as well as any linear combination of these





The University of Auckland

MIT Media Lab: Photobook

Database: keyframe
 Display mode: image
 Search metric: texture-ev
 Working Set: 365
 Status: ready
 Left button to select
 Middle button to search
 Right button for info

 boston,09610	 boston,18410	 boston,16010	 boston,18510
 boston,18310	 boston,09310	 boston,40106	 boston,09510
 boston,16210	 boston,18010	 boston,09410	 boston,15910
 boston,11310	 boston,18610	 boston,18210	 boston,12610

Operators: AND, NOT, EXCEPT, OF, Clear Filter, Controls: Search, Shuffle, Initialize, Load, Rearrange, Probe, Toggle windows, Primitives, Searched On, Icon, Metric Data, Page Up/Down, Page 1 of 23

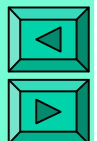
Query for images similar to the upper left image





Photobook + FourEyes

- Interactive learning subsystem
 - selecting and combining models on the basis of positive and negative examples from the user
- Due to the learning abilities of FourEyes, Photobook learns to select and combine features to satisfy a query
 - **Continuous learning:** each retrieval session adds to improvement of the system's performance
 - Learning to select and combine similarity measures, rather than design and understand them

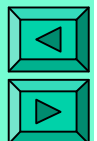
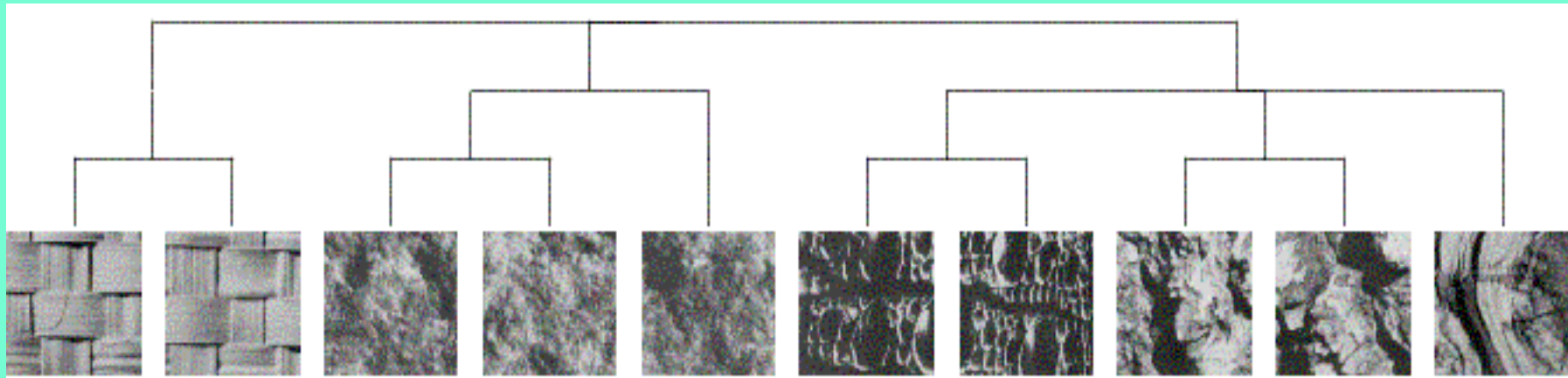




The
University
of Auckland

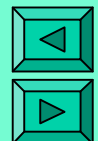
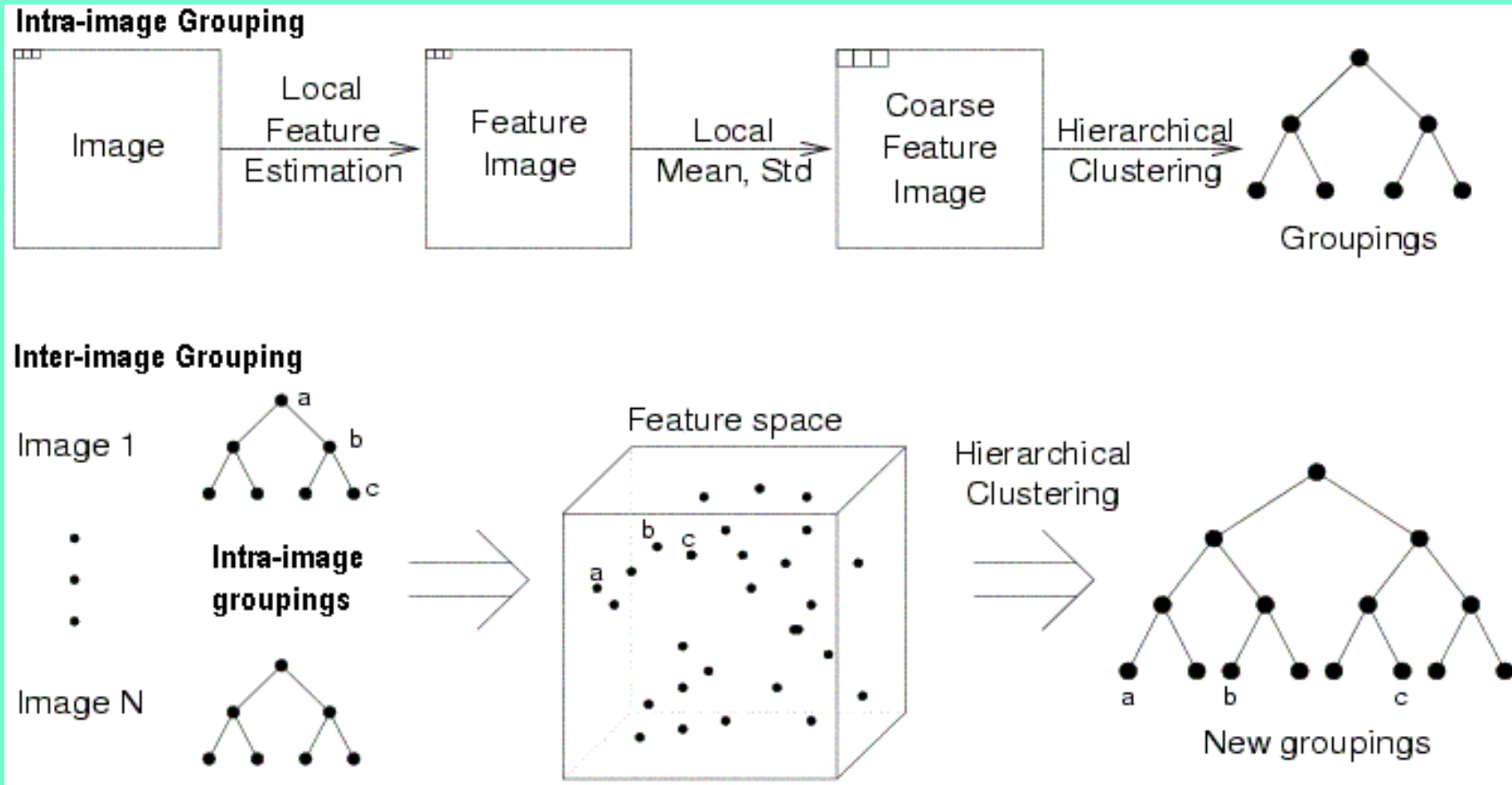
Data Grouping in FourEyes

- The data are first organised by grouping (clustering)
- The grouping allows for different similarity measures and can be produced manually, with due account of colour / texture models, optical flow information, etc:





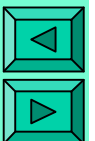
Data Grouping in FourEyes





Data Grouping in FourEyes

- **Intra-image** (within-image) groupings and **inter-image** (across-image) groupings composed of these:
 - Let grouping “a” in a single image contain “b”, which in turn contains “c”
 - In the feature space the groupings are individual, so that the resulting clustering may specify “a” looks more similar to “b” rather than to “c”
 - The user specifies queries by examples and relates most appropriate similarity measures to different parts of the query





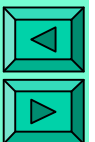
The University of Auckland

Computer-assisted Annotation

The user has selected a few sky patches in the two right images and labelled them "sky"

Intra-image grouping: the labelled patches grow into larger cross-hatched "sky" regions

Inter-image grouping places tentative labels on the two left images





The
University
of Auckland

Computer-assisted Annotation

- By pointing out the false labels, the user provides the negative examples to refine the decision rule
- Three decision strategies:
 - **Set covering** (SC): the smallest union of sets in the feature space which covers all of the positive examples and none of the negative ones
 - **Decision list** (DL): sets of either positive but no negative or negative but no positive examples
 - **Decision tree** (DT): the best division of examples into two parts for a possible further division

