

ECE624

Week 3-b

Performance Parameters for an Image Retrieval System

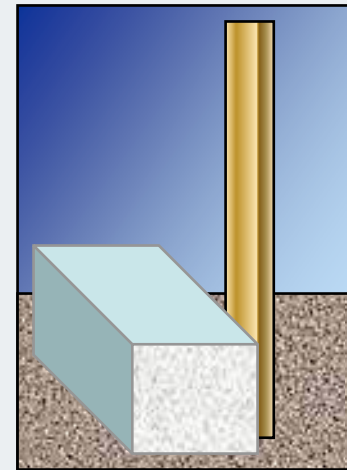
Recall: Ratio of relevant items retrieved to the total number of items in the database. It measures the ability of a system to present all relevant items

Precision: Ratio of relevant items retrieved to the total number of items retrieved. It measures the ability to present only relevant items.

Note: Number of false positive = $1 - \text{Precision}$

Indexing Problem

- Feature extraction
- Data structures for fast retrieval



Challenges in Image Indexing

- Application dependencies
- Similarity measures
- Scalability issues

Existing Solutions

- Raw image domain methods
 - QBIC
 - VisualSEEK
- Transform domain methods (Wavelets only)
 - *Fast Multiresolution Image Querying*
 - *Wavelet-based Image Indexing and Searching(WBIIS).*

QBIC (IBM)

- Features:

- Color histogram
- Texture
- Shape
- Bit map
- Edge map

- Problems:

- Linear search for all images in the DB is expensive
- Feature vector computation is costly

VisualSEEK (Columbia University)

- Features:
 - Color histogram
 - Objects and regions in the image
 - Inter-region relationships
- Search based on
 - Single/multiple region matching
 - Complex feature vector matching

Fast Multiresolution Image Querying

- Haar wavelets, color space (YIQ)
- 128x128 image size
- Feature vector uses 40-60 most significant coefficients from each color space
- Linear search in DB
- Query time depends on DB size

Wavelet-based Image Indexing and Searching(WBIIS) (Stanford University)

- Daubechies Wavelet transform
- Feature Vector:
 - Standard deviation of upper left 8x8 block of coefficients for each color
 - 8x8x3 coefficient matrix
 - 16x16x3 dimensional final feature set

Wavelet-based Image Indexing and Searching (WBIIS) (Stanford Univ, cont.)

- Performance results of search on 10,000 images
 - Standard deviations to obtain best 2000
 - Euclidean distance comparison of $8 \times 8 \times 3$ matrix
 - Final selection using $16 \times 16 \times 3$ matrix
 - 3.3 secs to select best 100 matches
- Problems: Not scalable

Liang, Kuo (University of Southern California)

- Features
 - Frequency of important coefficients in each sub-band
 - Luminance histogram
 - Binary quantization of lowest subband
 - Color histogram
- Independent or joint use of features for retrieval

Wavelet Domain Methods

| System | Image Size | Database Size | Feature Vector Size | Search Time |
|---------------|-------------------|----------------------|----------------------------|--------------------|
| Liang Kuo | 192x128 | 2119 | 212 bytes | NA |
| WBIIS | 128x128 | 10,000 | ~1000 bytes | 3.3secs |
| QBIC | 100x100 | 1000 | NA | 2-40secs |
| U. Washington | 128x128 | 1093 | >150 bytes | 47.46 secs |

Issues

- Scalability issues
- Image keys are long (typically 256 bytes)
- Ratio of number of images retrieved to the number of images selected is high
- Feature extraction and indexing are treated as two separate problems in general in a non-uniform manner

Wavelet-Based Approach

- Features based on the properties of the wavelet coefficients
- Utilization of multiresolution property of the wavelet transform
- Classification of images to reduce the search space
- Efficient data structures for fast image indexing and retrieval

Image Databases: Summary

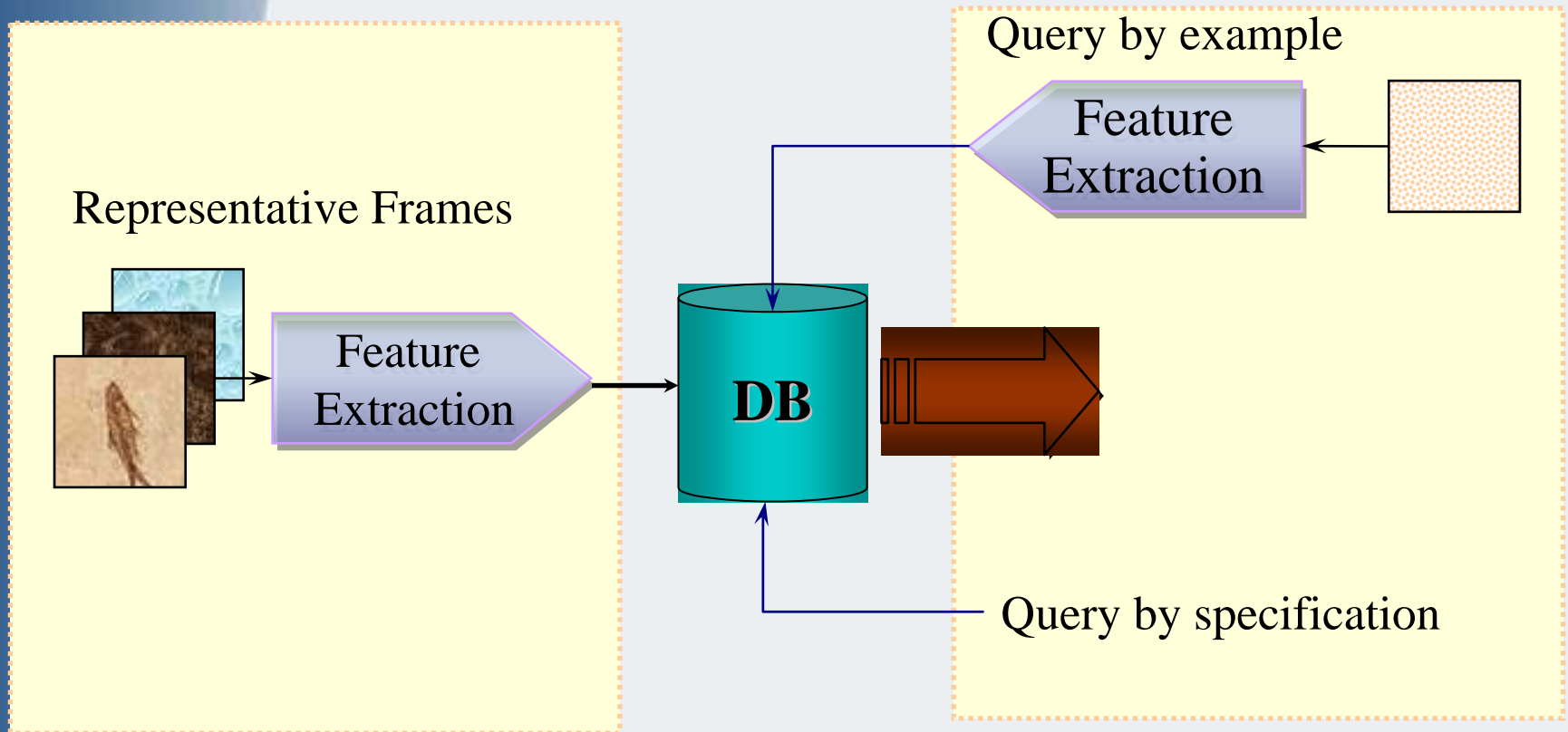
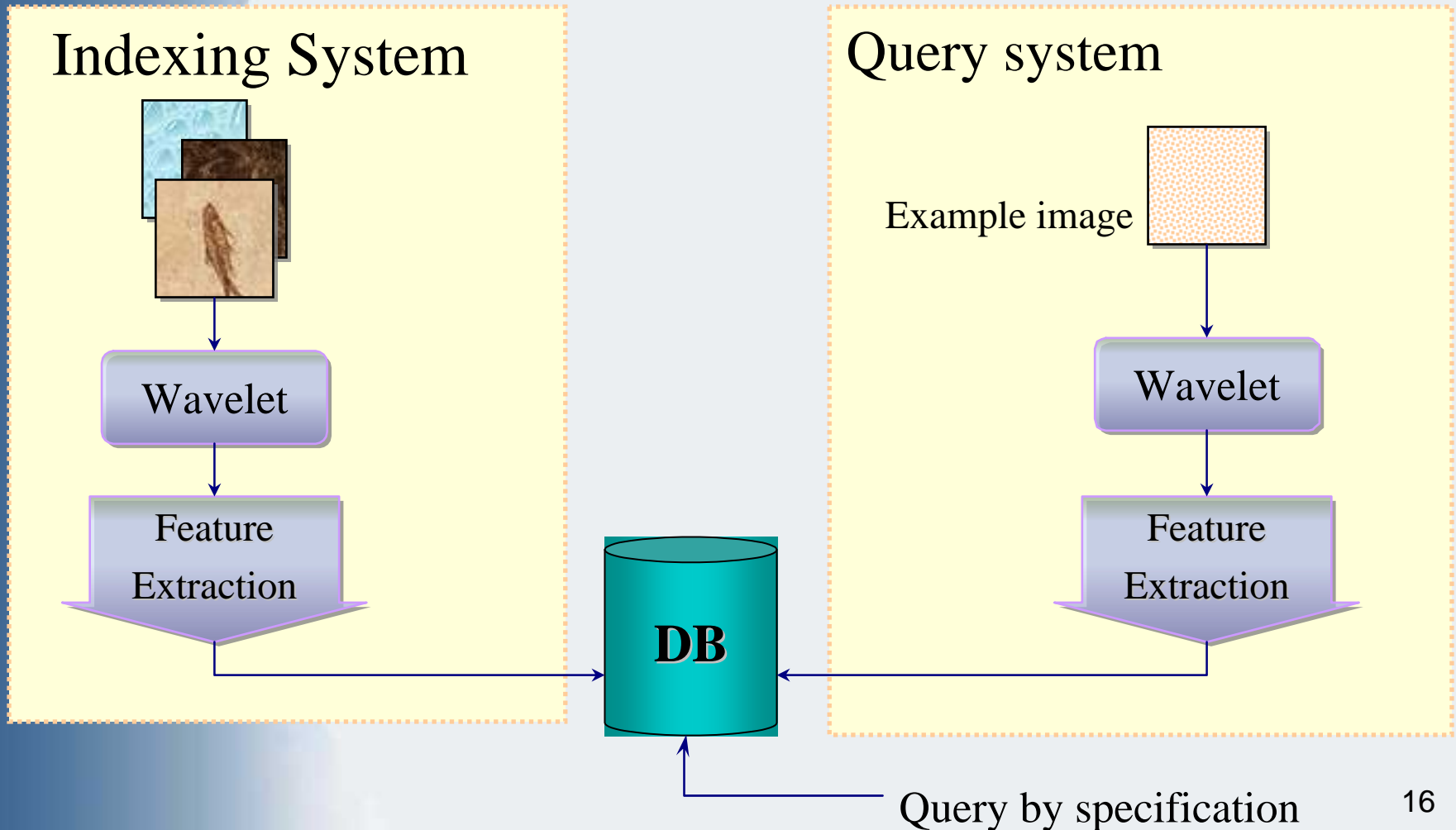


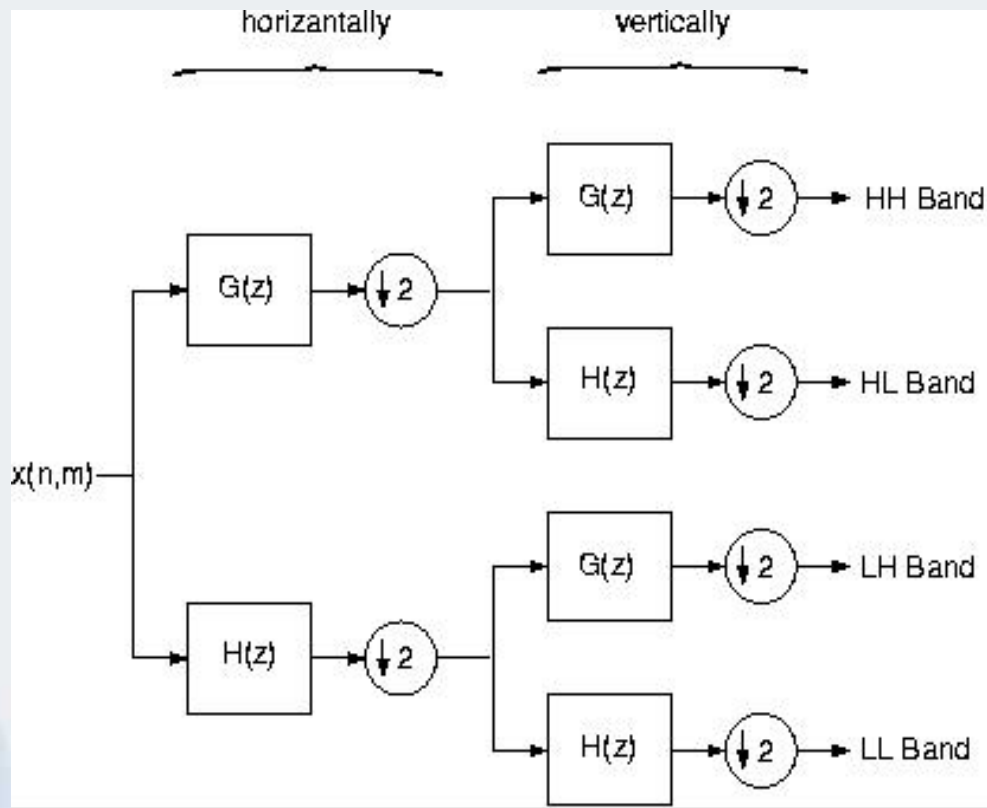
Image Database System: Summary



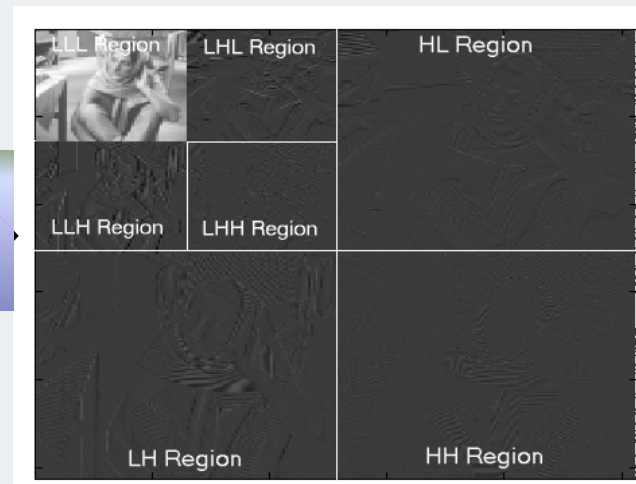
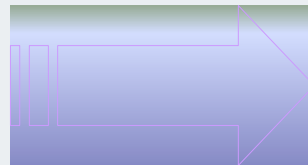
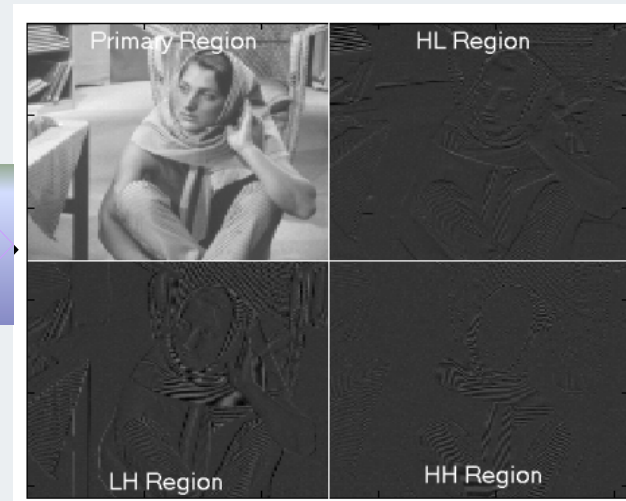
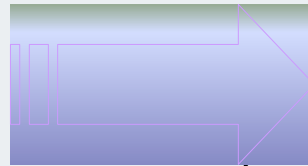
Why Wavelet Transform?

- Real-life data, such as audio and images, is not totally random but tends to have certain level of correlation.
- Correlation is local
- Better space-frequency localization
- Signal/image analysis at multiple resolutions
- Low computational complexity

2D Wavelet Transform (Mallat)



2D Wavelet Transform



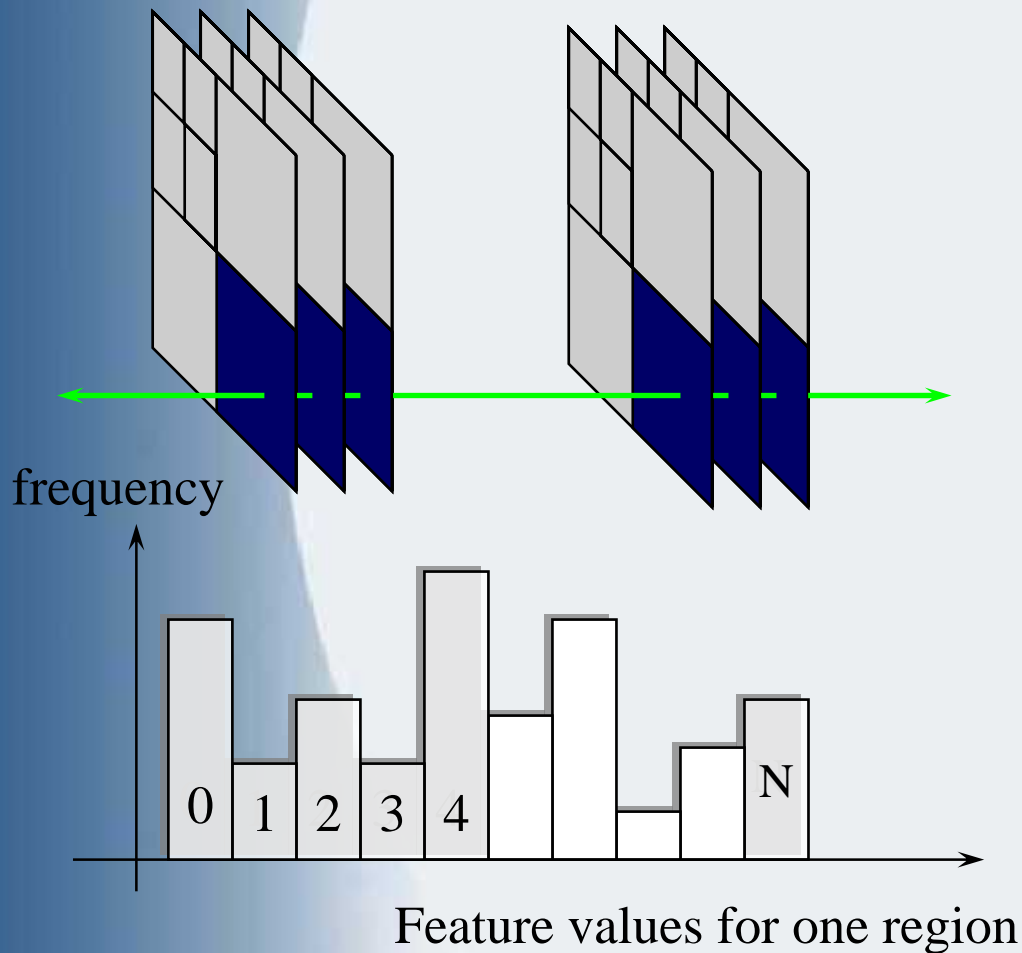
Indexing of Images

| | | | |
|-----|-----|-----|------|
| FV1 | FV2 | FV5 | FV8 |
| FV3 | FV4 | | |
| FV6 | | FV7 | |
| FV9 | | | FV10 |

Steps:

- Apply wavelet transform to each image
- For each region compute a feature value

Class (bucket) Construction



- Construct a histogram using the same region for each image.
- Determine similar images according to feature value.
- Assign class names or numbers to each image.

Feature Vector Computation

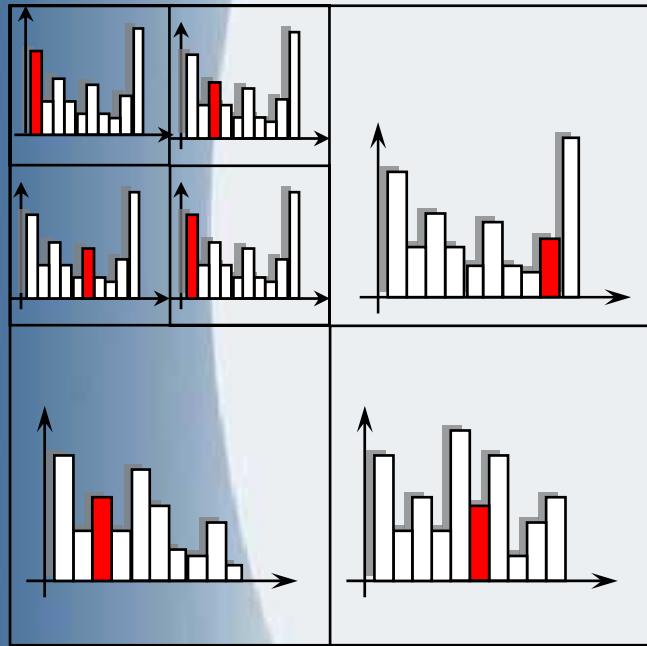


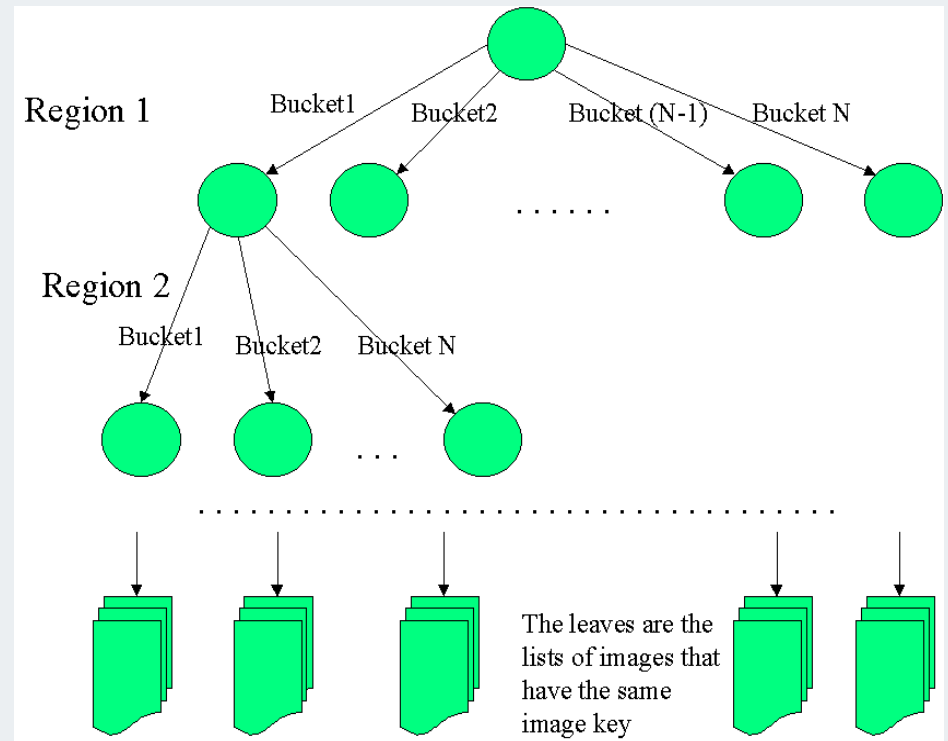
Image Key=

| | | | | | | |
|---|---|---|---|---|---|---|
| 1 | 3 | 6 | 1 | 9 | 3 | 6 |
|---|---|---|---|---|---|---|

- For a specific image, a class number is obtained from each frequency sub-band.
- Class names are stored for future insertions and used in processing queries.

Index Structure

- K-ary tree structure
- Leaves contain sets of similar images.
- Tree traversal is governed by the feature vector



Properties of Indexing Method

- Images are classified at each frequency sub-band.
- The bucket mechanism allows distribution of images over an index function.

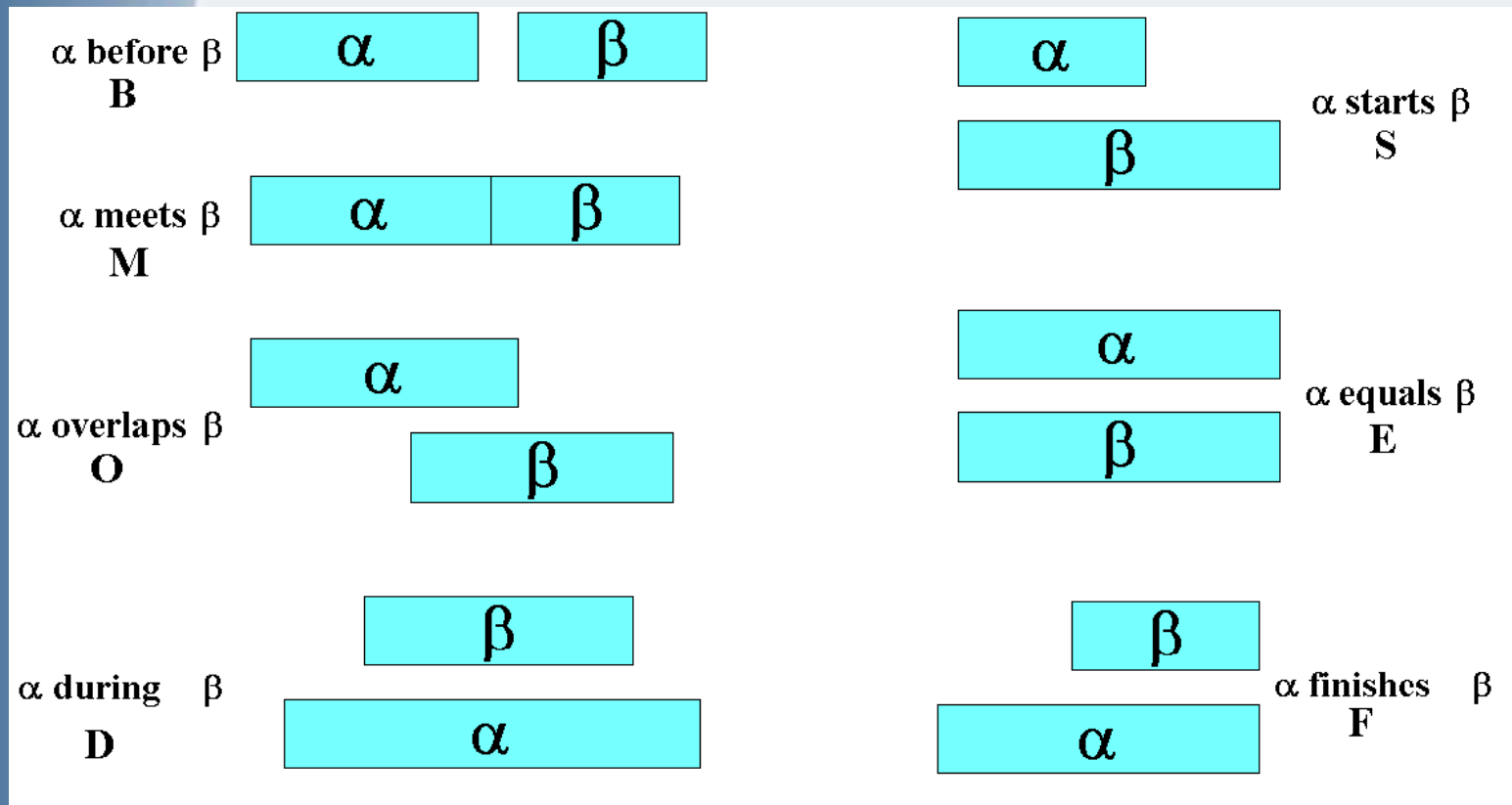
Study “Knowledge-Based Image
Retrieval with Spatial and Temporal
Constructs”, W. Chu et al
IEEE TKDE Nov/Dec 1998

Spatial Semantic-based Search

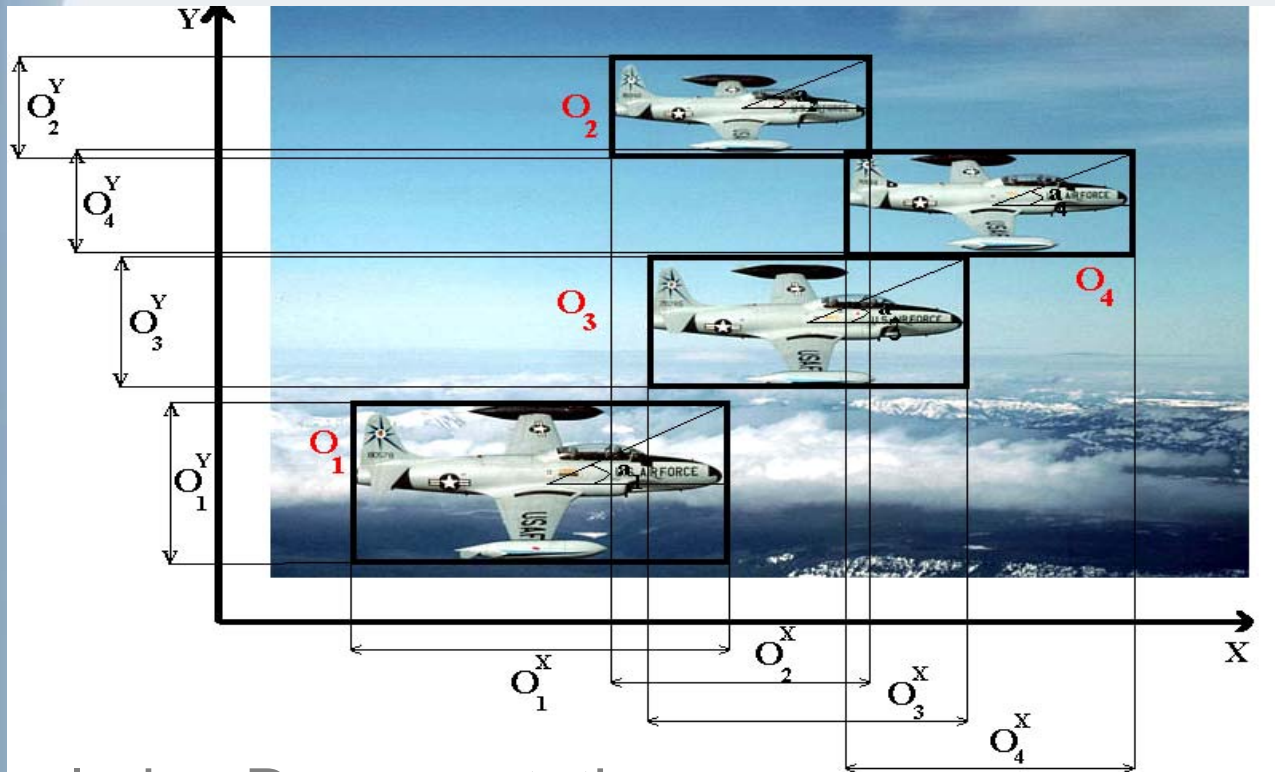


Spatial semantics: Arrow formation

Binary Spatio/Temporal Relations



Example of Spatial Meta-Knowledge



Meta-Knowledge Representation:

$$\mathbf{O} (O_1^x, O_2^x, O_3^x, O_4^x, \tau_{12}, \tau_{23}, \tau_{13}, \tau_{34}) \wedge \mathbf{B} (O_1^y, O_3^y, O_4^y, O_2^y, \tau_1, \tau_2, \tau_3)$$