

# Image Databases

- Indexing Content-based features
- Querying the exact vs Similar matches

# Content Based Search and Retrieval

- **Index** multimedia data based on the *contents of the data*, not the description
- **Contents** may include, e.g. in images:
  - Structure of physical objects
  - Patterns of color intensities
  - Spatial relationship among objects
- **Search** indexed data archives using contents of the query.
- **Retrieval** based on *similarity* NOT exact match

# Characteristics of Internet Data

- Documents consist of multiple media: **audio, image, video**
- Volume: **huge**
- Manipulation: **requires diverse computational techniques**, ranging from simple algebra to signal processing to artificial intelligence.

# Content Based Search and Retrieval

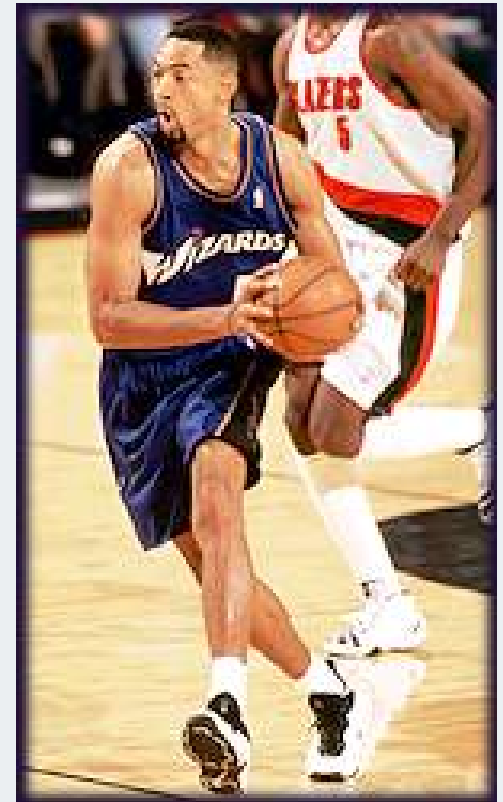
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# Search and Retrieval in Today's Internet

- **Keywords:** okay for text documents
- **Images, audio, and video:** textually annotated
- **Search** based on textual descriptions

# Difficulties of Annotation

- Perceptual differences
- Limited information can be attached to each picture
- Necessary information may not be covered
- Annotation time and effort



# Challenges

- **Similarity Criteria**
  - closer to the human perception
  - difficult in the presence of multiple features
- **Information Abstraction**
  - indexing and retrieval of real time media requires abstractions at multiple levels
  - unbiased abstractions are difficult
- **Query Modeling and Interface**
  - expressing knowledge and concepts

# Challenges

- **Scalability**

- response time

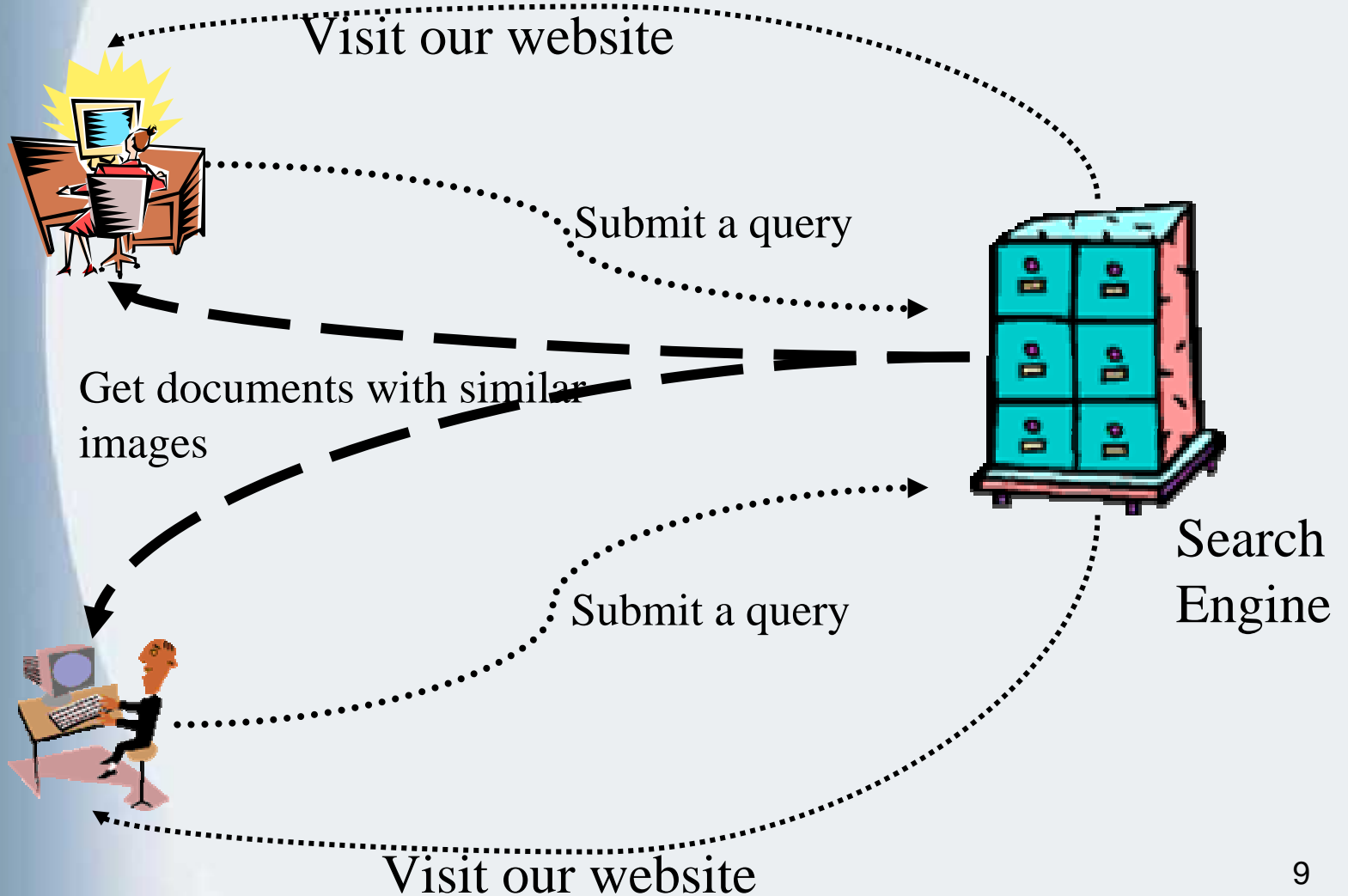
- response time should scale linearly only with the number of objects similar to the query NOT with the database size.

- quality

- robustness of the feature spaces should scale linearly with the database size.



# Search Engine

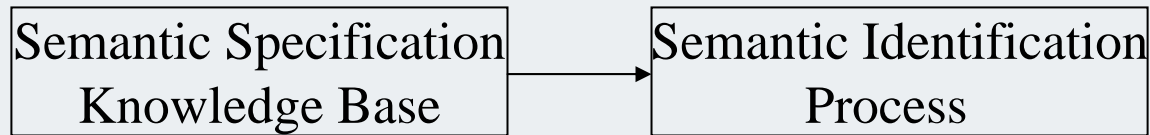


# Issues Involved

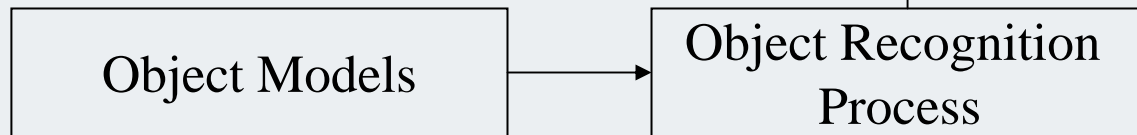
- Multiplicity of users
  - (250,000 unique users/day for Hotbot)
- Database size !
  - Hotbot (110 million documents)
- Multiple Databases

# Semantic Modeling and Knowledge Representation in Image Databases

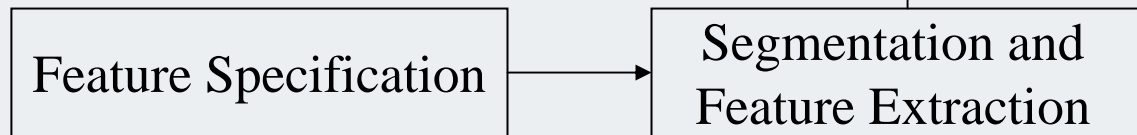
Semantic Modeling  
and Knowledge  
Representation  
Layer



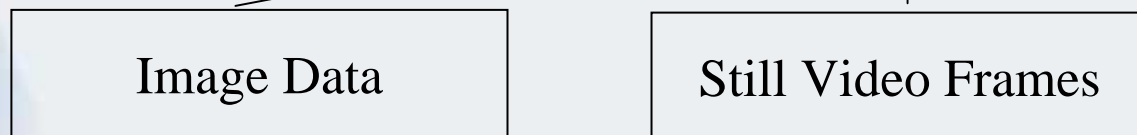
Object  
Recognition  
Layer



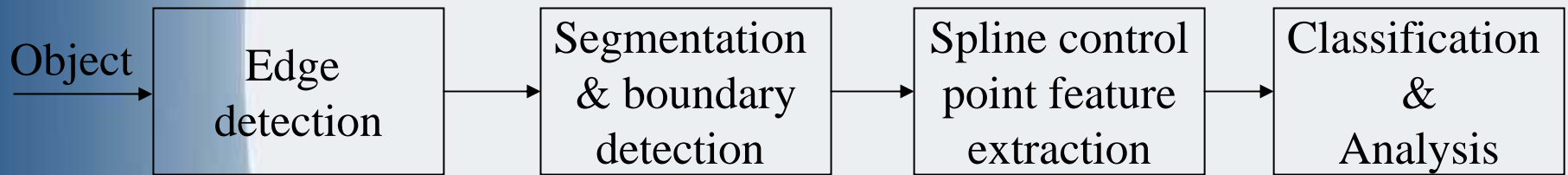
Feature  
Extraction  
Layer



Multimedia  
Data



# Steps in Image Processing



# Feature Extraction Layer

- Image Features: colors, Textures, Shapes, Edges, ..etc.
- Features are mapped into a multidimensional feature space allowing similarity-based retrieval.
- Features can be classified into two types: Global and local.
- Global features generally emphasize “coarse-grained” pattern matching techniques.
  - Transform the whole image into a functional representation.
  - Finer details within individual parts of the image are ignored.
  - Color histograms, Fast Fourier Transform, Hough Transform, and Eigen values are well known functional techniques.

# Images Databases

- Requirements:
  - Image Processing Capability
  - Image understanding Capability (knowledge-based)
- Image Representation
  - Local Features:
    - Pixels
    - Edges
    - Shape
    - Texture
    - Colors
  - Global Features:
    - Histograms
    - FFT
    - Hough Transform
    - Eigenvalues

Different features are useful for different types of queries.

# Global Features

- Advantages
  - Simple
  - Low computational complexity
- Disadvantage
  - Low accuracy

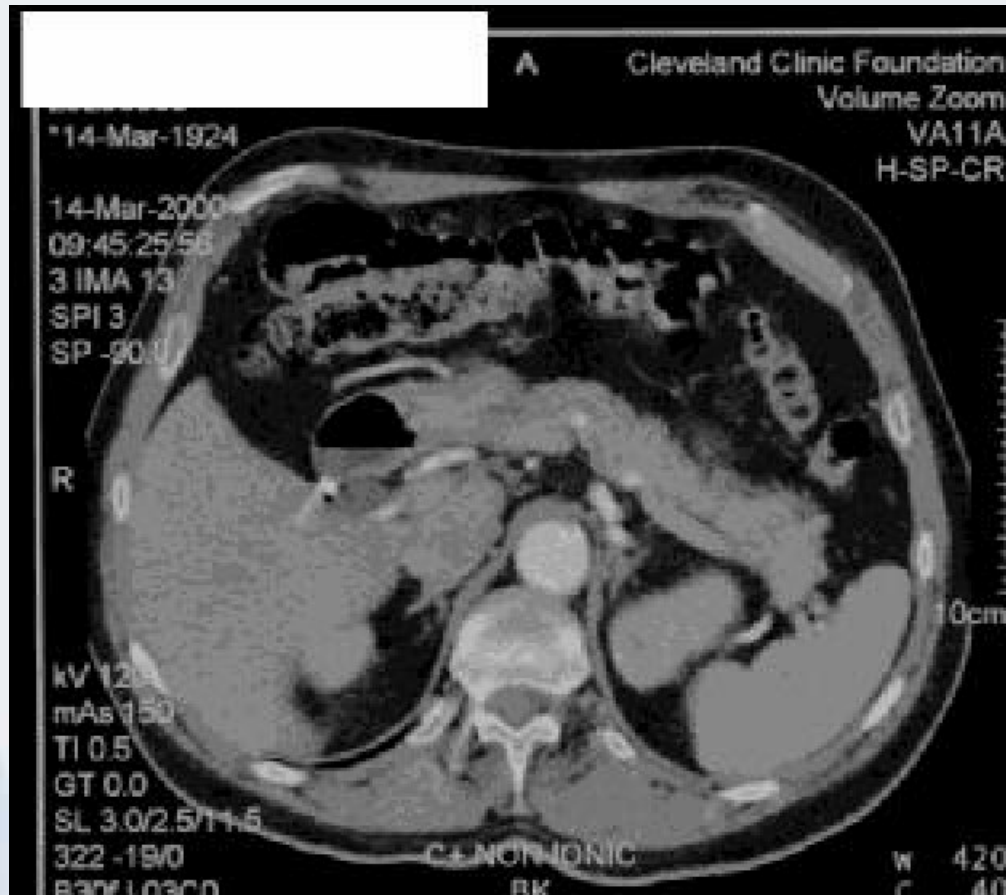
# Local Feature

- Images are segmented into a collection of smaller regions, with each region representing a potential object of interest (fine grained)
  - An object of interest may represent a simple semantic object (e.g. a round object)
  - Choice of features is domain specific.
    - X-ray imaging, GIS, ..etc require spatial features (e.g. shapes and dimensions).
    - Paintings, MMR imaging, ..etc may use color features.

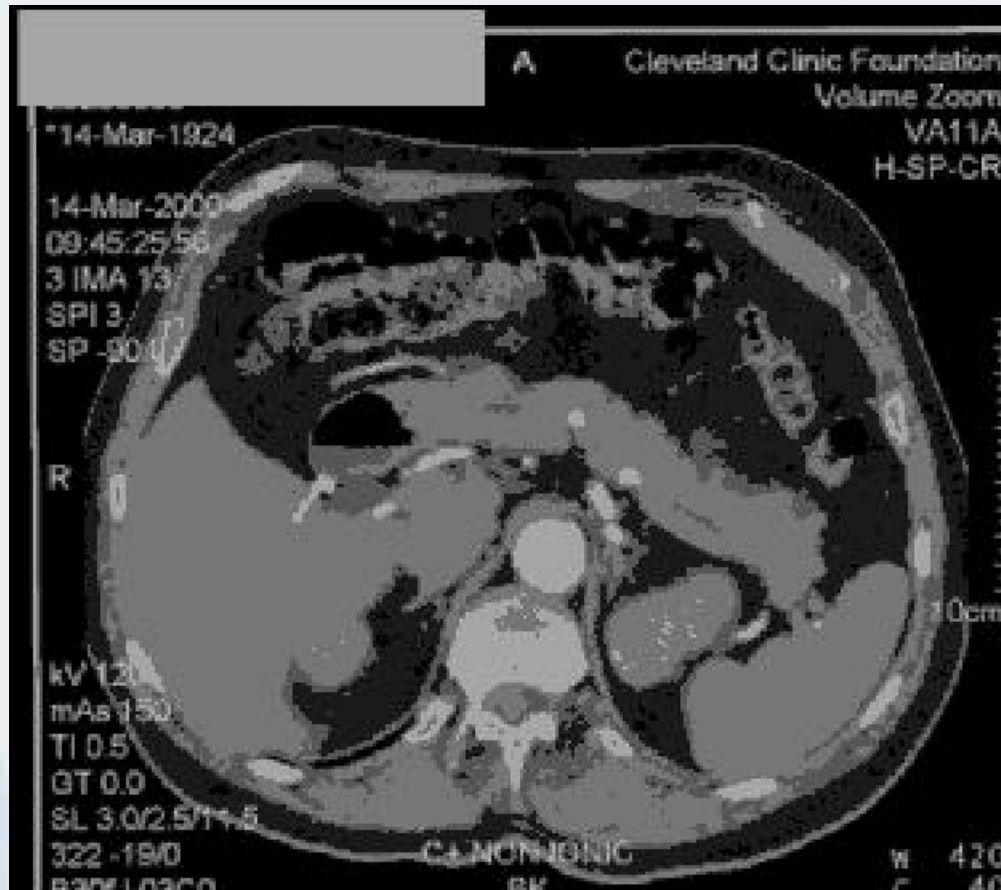


# Segmentation and feature extraction (example)

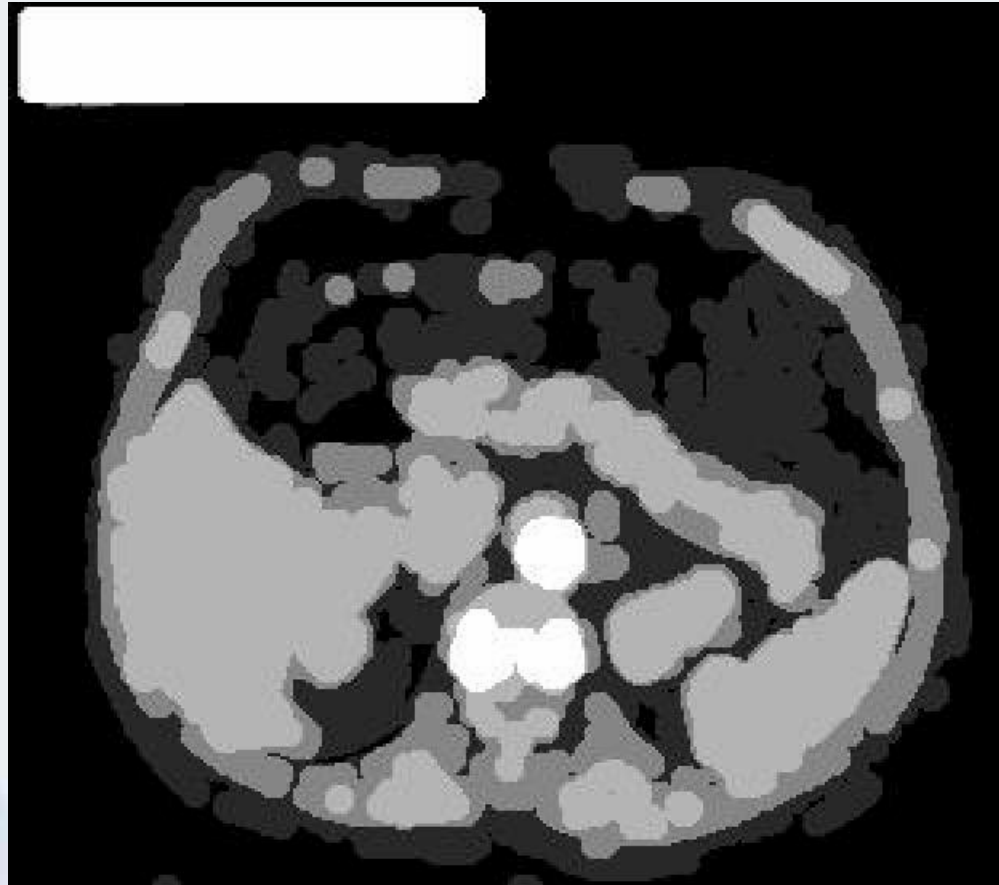
## Original Image



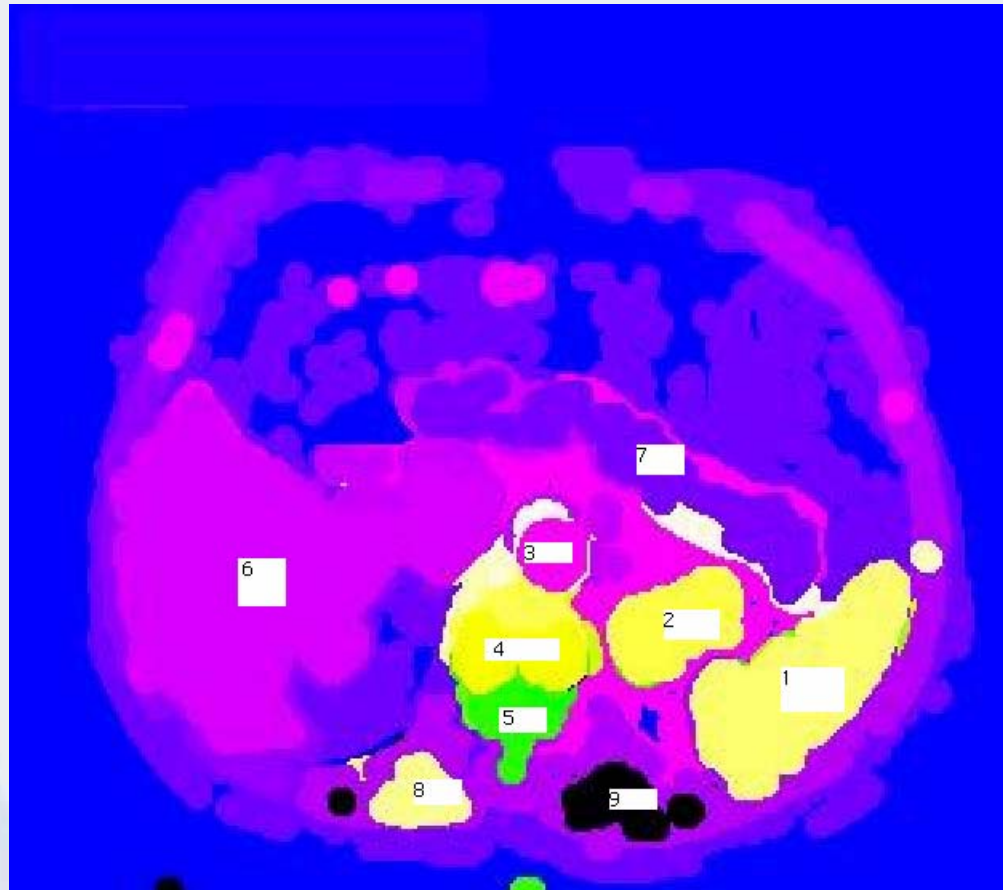
# Segmented Image



# After Morphological Filtering



# Labeled Image



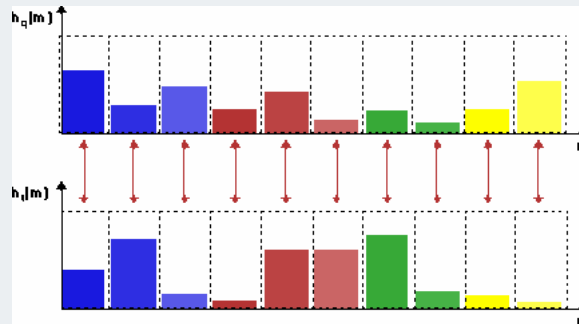
# Binary Object Features

Object	Area	Xc	Yc	Orientation	Perimeter	Euler No.	Aspect Ratio
1	2588	213	247	-0.703	200	202	0.932
2	855	194	207	-0.752	100	190	1.114
3	406	172	170	0.780	63	170	0.957
4	787	203	159	0.663	114	202	1.615
5	468	225	157	0.554	145	231	0.892
6	7449	177	79	0.393	413	182	1.061
7	2554	144	193	-0.625	343	134	1.730
8	521	247	128	0.393	78	243	1.333
9	693	250	193	0.625	115	249	1.800
10	4338	16	77	0.000	338	19	4.833
11	16496	142	168	-0.687	1615	94	1.199

# Similarity Based Retrieval

- Minkowsky Distance

- $d(H_q, H_t) = \left\{ \sum_{i=1}^n |h_q(i) - h_t(i)|^p \right\}^{1/p}$   
–  $h_q$  and  $h_t$  are the histograms of query and target images.
- ‘ $n$ ’ is the total number of bins.
- $h_x(i)$  is the average height of the  $i$ th bin.



# Similarity Based Retrieval

- Histogram intersection

- $$d = 1 - \frac{\sum_{i=1}^n \min(h_q[i], h_t[i])}{\min(|h_q|, |h_t|)}$$
- used to find known objects within images using color histogram
  - The object (q) size is less than the image (t) size.

# Similarity Based Retrieval

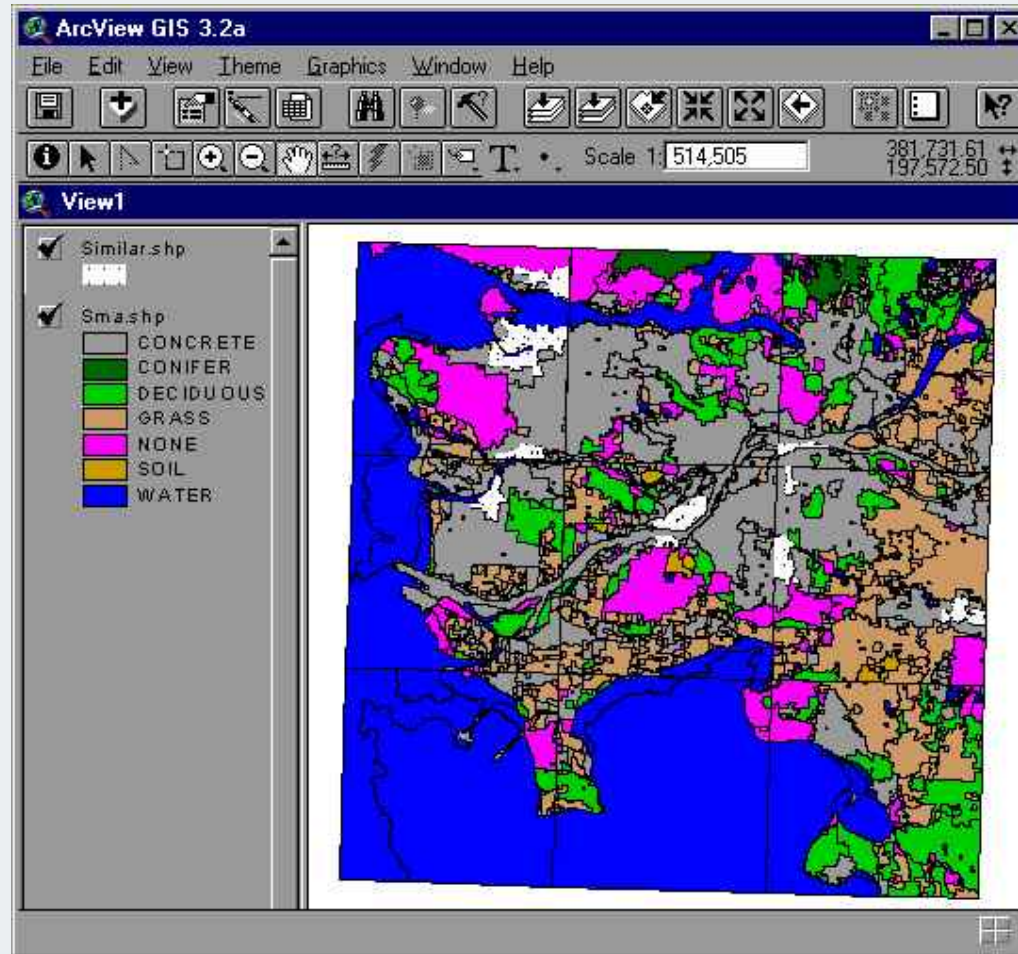
- Histogram Euclidean Distance

$$d = (h_q - h_t)^T (h_q - h_t) = \sum_{i=1}^n \left( h_q(i) - h_t(i) \right)^2$$

- $h_q$  and  $h_t$  are the histograms of the query and target images respectively.



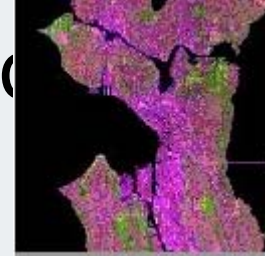
# Information Retrieval from Satellite Imagery and GIS Data



Segmented Image

# Querying GIS Data (Example)

**Query:** Find the satellite Images of all downtowns that are similar to  
(Seattle Downtown)



*Similarity is based on level of concrete, vegetation, terrain etc.*

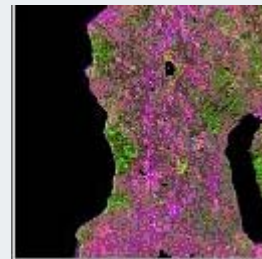
## Result



Vancouver  
Downtown



South Seattle  
Downtown



North Seattle  
Downtown



Victoria  
Downtown

# Existing Systems

- Query by Image Content (QBIC) (IBM)
- VisualSEEK (EE, Columbia Univ)
- Blobworld (CS, UC-Berkeley)
- **WBIIS (CS, Stanford Univ)**
- **Multiresolution Wavelets (CS, Univ of Washington)**
- Virage (CS, UC-San Diego)
- Chabot
  - Uses color and textual annotation.
  - Improved performance due to textual annotation (concept Query).
- KMeD
  - Uses shapes and contours features.
  - Features are extracted automatically in some cases and manually in other cases.