Content-Based Image Retrieval

Selim Aksoy
Department of Computer Engineering
Bilkent University
saksoy@cs.bilkent.edu.tr

Image retrieval

- Searching a large database for images that match a query:
 - What kind of databases?
 - What kind of queries?
 - What constitutes a match?
 - How do we make such searches efficient?

Applications

- Art Collections
 - Fine Arts Museum of San Francisco
- Medical Image Databases
 - CT, MRI, Ultrasound, The Visible Human
- Scientific Databases
 - Earth Sciences
- General Image Collections for Licensing
 - Corbis, Getty Images
- The World Wide Web
 - Google, Microsoft, Flickr

Corel data set



118011 WATER HARBOR SKY CLOUDS



TIGER CAT WATER GRASS



1090 SUN CLOUDS WATER SKY



1015 SUN TREE PLAIN SKY



143078 MOUNTAINS TREES aspens VALLEY



102042 MUSEUM memorial FLAGS GRASS



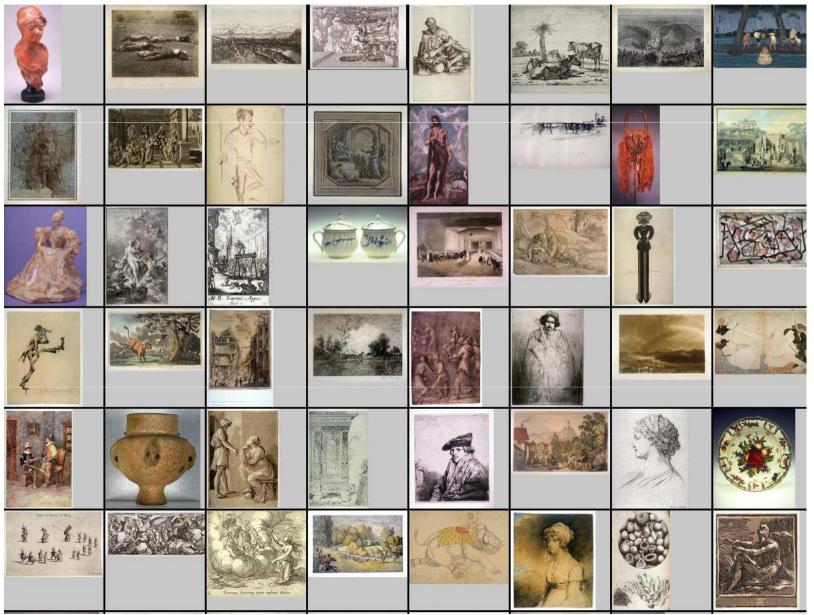
119094 GARDEN BUILDING FLOWERS TREES



131007 GARDEN FLOWERS HOUSE TREES

60,000 images with annotated keywords

Fine Arts Museum of San Francisco



80,000 images

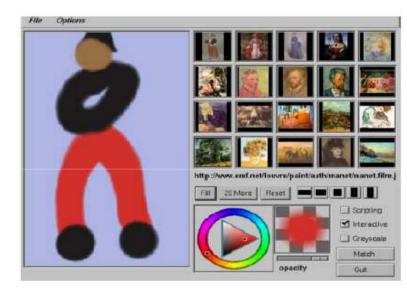
CS 484, Fall 2012 ©2012, Selim Aksoy 5

Query formulation

- Text description (keywords)
- Query by example



Query by sketch



- Symbolic description (man and woman on a beach)
- Relevance feedback

Google query on "rose"



Two-Tone Rose 500 x 375 - 95k - jpg gardening.about.com [More from z.about.com]



Welcome to the Corpus Christi Rose 351 x 334 - 51k - gif www.geocities.com



Frosty Rose 600 x 451 - 70k - jpg blogs.warwick.ac.uk



400 x 313 - 40k - jpg www.telegraph.co.uk



Dominic Cavendish reviews The Rose We've been in the business of rose. 400 x 539 - 78k - jpg www.witherspoonrose.com



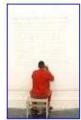
Rose 512 x 512 - 19k - ipg www.parabola.me.uk



ROSE Online 800 x 640 - 393k - ipg www.gameogre.com



Tracey Rose The Thinker, 1996 500 x 451 - 26k - ipg www artthrob co za



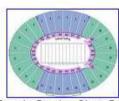
Tracey Rose Span II, 1997 317 x 500 - 18k - ipg www artthrob co za



The rose apple is occasionally ... 350 x 307 - 57k - jpg www.tradewindsfruit.com



347 x 348 - 18k - jpg lewisrose.com



449 x 354 - 18k - gif www.barrystickets.com



Lewis Rose also known as Lewis N. ... Rose Parade Seating Chart Seating ... Rose Quartz point shape in Sterling ... 456 x 478 - 15k - jpg www.wehug.com



The Exorcism of Emily Rose 300 x 375 - 29k www.smh.com.au



rose rose 6499 hits 574 x 576 - 33k - ipeq www.firstmonday.org

Corel query on "rose"



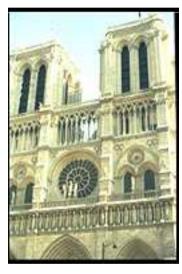












Corbis query on "rose"



42-17895937 (RM)
Season of Perfume by
Yang Ping
Yang Ping
>>> Price Image

☐ Lightbox ☐ Cart



42-17895408 (RM)
Autumn Beauties by Wang
Chingho
Wang Chingho
Price Image
Lightbox Cart



42-17895882 (RM)
Little Black Cat by Wu
Yeizhao
Wu Yeizhao
Se Price Image
Lightbox Cart





42-17895421 (RM)
Butterflies Among
Flowers by Wang Chingho
Wang Chingho
Price Image
Lightbox Cart



42-17895445 (RM)
Red Roses by Lu Bisa
Lu Bisa
>>> Price Image

Lightbox Cart



42-17895962 (RM)
Beautiful Flowers by Cao
Jingen
Cao Jingen
>> Price Image
Lightbox Cart



42-16247767 (RM)
Peppers Stuffed with
Rosebuds
J.Bilic
>> Price Image



Parid Bowie Smoking
Cigarette
1976
Steve Schapiro
>> Price Image

Lightbox Cart



42-17165934 (RM)
Woman Taking a Bath in
Rose Petals
Fendis
>> Price Image

Lightbox Cart



42-16246447 (RM)
Oriental-Style Litchi
Salad
J.Riou
>> Price Image

Lightbox Cart





42-16801959 (RM)
Bouquet of Red Roses
May 27, 2005
Owen Franken
>> Price Image
Lightbox Cart





42-17529104 (RM)
Rose Red #9 Series by
Elisa Lazo de Valdez
April 1, 2006
Elisa Lazo de Valdez
>> Price Image



42-17529137 (RM)
Rose Red #46 Series by
Elisa Lazo de Valdez
April 1, 2006
Elisa Lazo de Valdez
>> Price Image



42-17529136 (RM) Rose Red #57 Series by Elisa Lazo de Valdez April 1, 2006 Elisa Lazo de Valdez >> Price Image



42-15766292 (RM)
Roses in Kohinoor Suite
Bathroom at Amarvilas...
January 31, 2003
Remi Benali
>> Price Image





42-15944324 (RM)
Bridal Bouquet
2005
Robert Levin
Drice Image



42-16248521 (RM)
Red Onion
J.Garcia

>>> Price Image

Lightbox

Cart





42-16246499 (RM)
Glass of Rose Wine
G.Flayols

Price Image
Lightbox Cart



42-15766394 (RM) Rose Petal Bath at Vanyavilas Resort January 31, 2003 Remi Benali >> Price Image

Difficulties with keywords

- Images may not have keywords.
 - (An image is worth ... how many key-words?)
- Query is not easily satisfied by keywords.
 - "A casually dressed couple gazing into each others eyes lovingly with dramatic clouds in the background."
 - "Pretty girl doing something active, sporty in a summery setting, beach - not wearing lycra, exercise clothes - more relaxed in tee-shirt. Feature is about deodorant so girl should look active - not sweaty but happy, healthy, carefree - nothing too posed or set up nice and natural looking."
- Content-based image retrieval (CBIR)

Content-based image retrieval

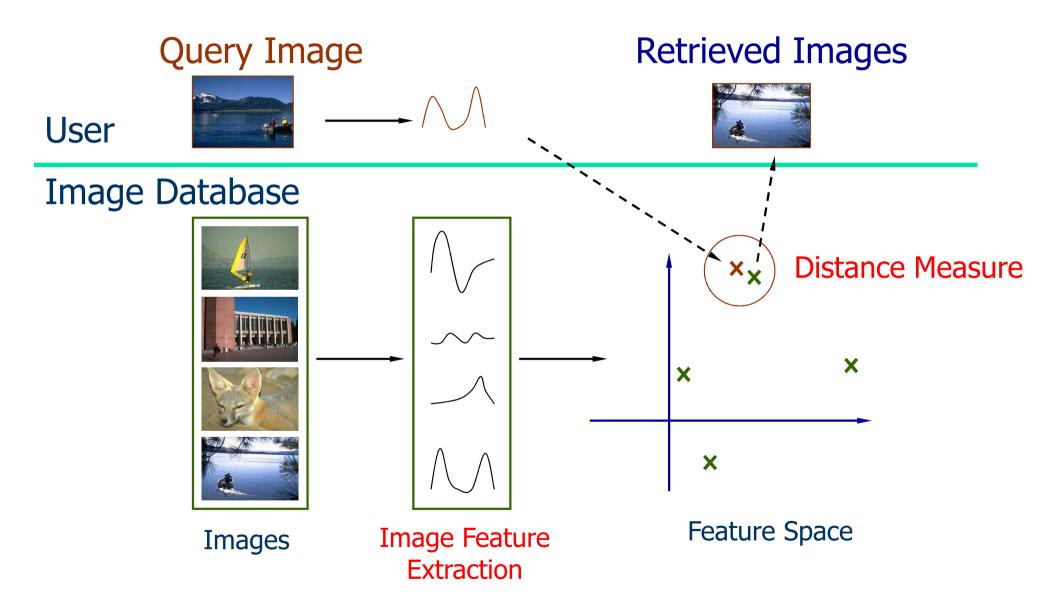


Image representations and features

- Image representations:
 - Iconic
 - Global
 - Region-based
 - Object-based
- Image features:
 - Color
 - Texture
 - Shape
 - Objects and their relationships (this is the most powerful, but you have to be able to recognize the objects!)

Image similarity

Distance measures:

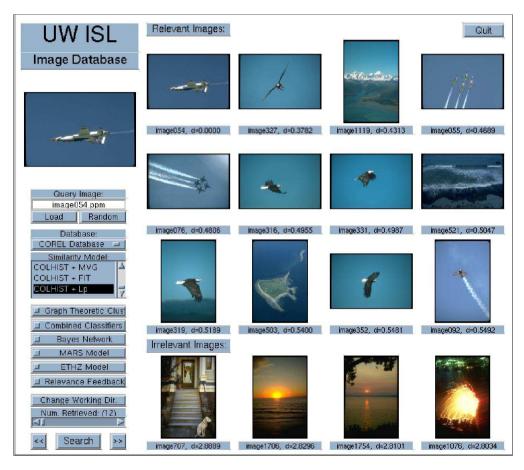
- Euclidean distance
- Other L_p metrics
- Histogram intersection
- Cosine distance
- Earth mover's distance
- Probabilistic similarity measures:
 - P(relevance | two images)
 - P(relevance | two images) / P(irrelevance | two images)

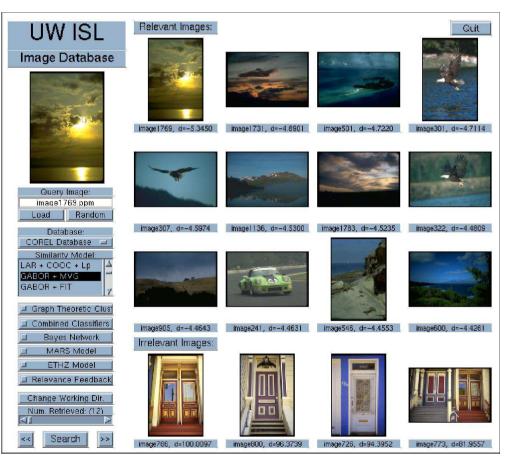
Global histograms

Searching using global color histograms



Global histograms





"Airplanes" using color histograms (4/12)

"Sunsets" using Gabor texture (3/12)

Query by image content (QBIC)



First commercial system

QBI€[™]

- Search by:
 - color percentages
 - color layout
 - texture
 - shape/location
 - keywords

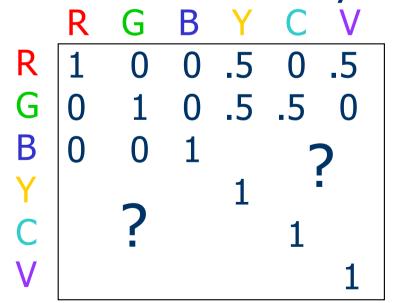
Try their demo: http://wwwqbic.almaden.ibm.com

Color histograms in QBIC

The QBIC color histogram distance is:

$$d_{hist}(I,Q) = (h(I) - h(Q))^T A (h(I) - h(Q)).$$

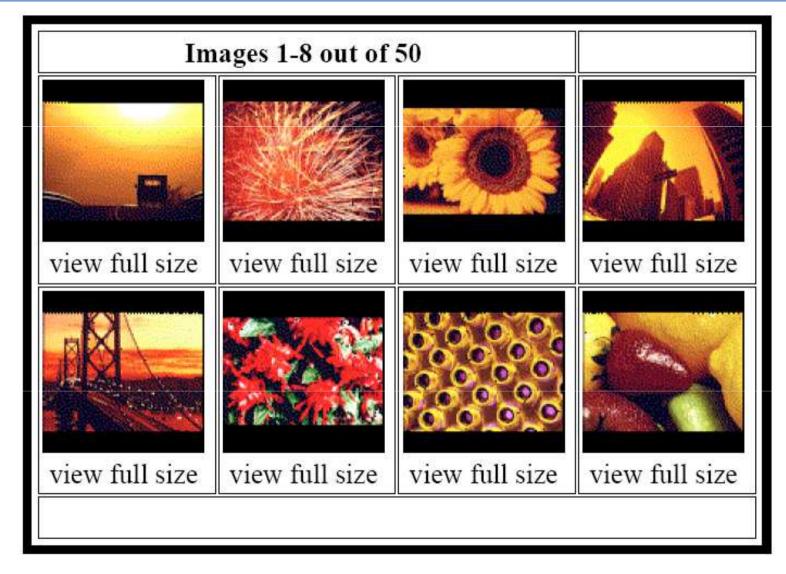
- h(I) is a K-bin histogram of a database image.
- h(Q) is a K-bin histogram of the query image.
- A is a K x K similarity matrix.



How similar is blue to cyan?

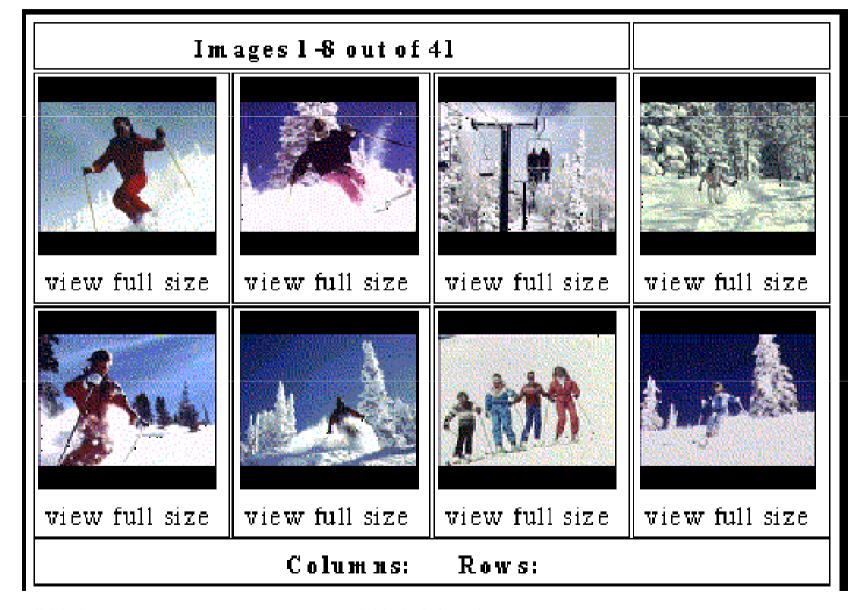
CS 484, Fall 2012 ©2012, Selim Aksoy 17

Color percentages in QBIC



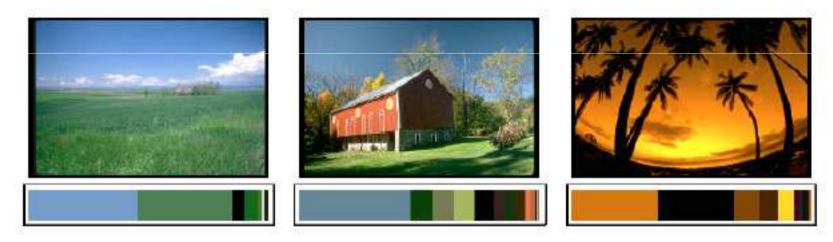
%40 red, %30 yellow, %10 black

Color layout in QBIC



Earth mover's distance

For each image, compute color signature:



Define distance between two color signatures to be the minimum amount of "work" needed to transform one signature into another.



Earth mover's distance

Visualization using EMD and multidimensional scaling



Probabilistic similarity measures

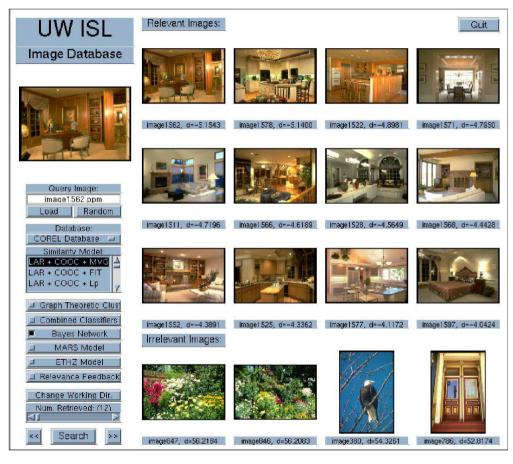
- Two classes:
 - Relevance class A
 - Irrelevance class B
- Bayes classifier

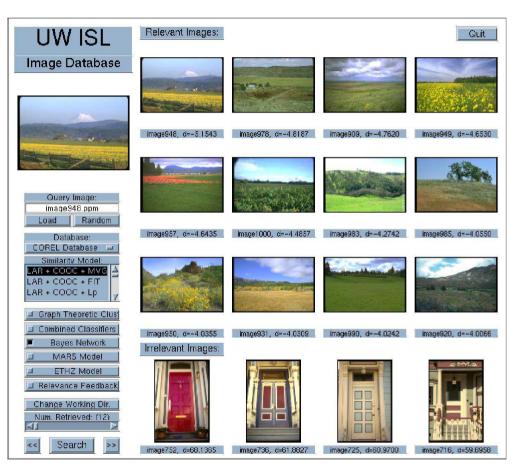
 Assign (ξ_i, ξ_j) to $\begin{cases}
 A & \text{if } P(A | (\xi_i, \xi_j)) > P(B | (\xi_i, \xi_j)) \\
 B & \text{otherwise}
 \end{cases}$
- Discriminant function for classification

$$\Delta(\xi_{i}, \xi_{j}) = \frac{P(A | (\xi_{i}, \xi_{j}))}{P(B | (\xi_{i}, \xi_{j}))} = \frac{P((\xi_{i}, \xi_{j}) | A) P(A)}{P((\xi_{i}, \xi_{j}) | B) P(B)}$$

 Rank images according to posterior ratio values based on feature differences.

Probabilistic similarity measures





"Residential interiors" (12/12)

"Fields" (12/12)

Shape-based retrieval

Find more shapes like this





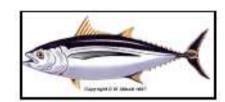












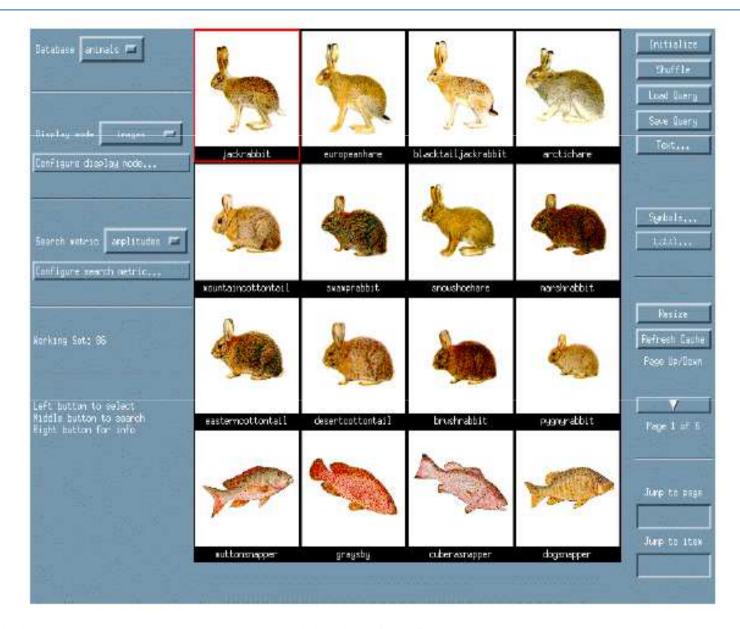




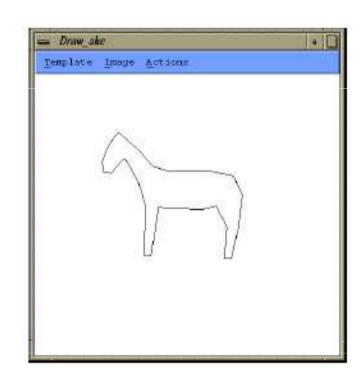




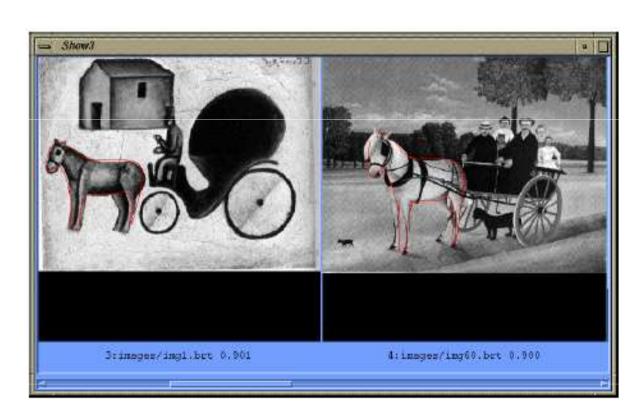
Shape-based retrieval



Elastic shape matching



Sketch-based query



retrieved images

Iconic matching

Example applications:

- Copyright and trademark protection
- Duplicate removal
- Linking images used in evidence, for example child pornography

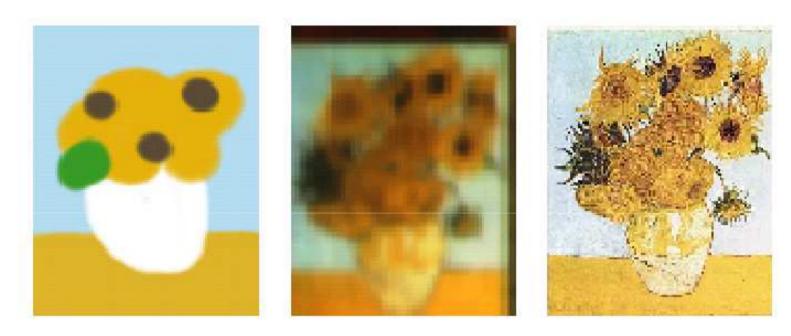
Problems in finding "exact" matches:

- Lossy compression, image scanning
- Color space conversion
- Photoshop-style transforms: blur, scale, rotate, warp, crop, cut, etc.

Iconic matching

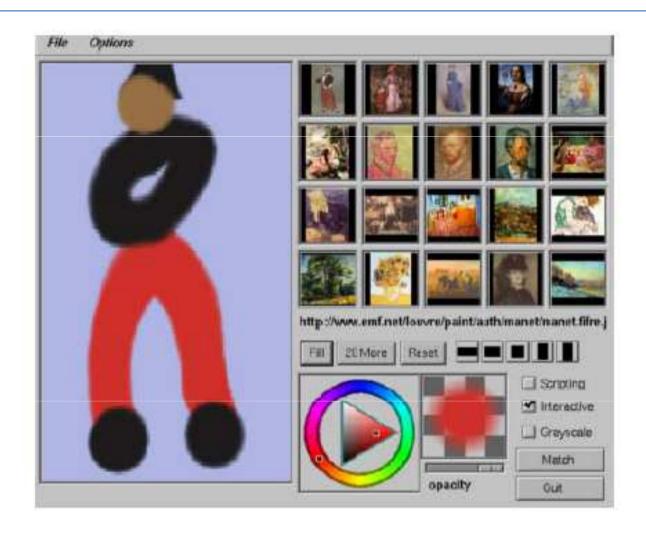
- Wavelet-based image compression
- Quantization of coefficients

painted scanned target



From Jacobs, Finkelstein, & Salesin
Fast Multi-Resolution Image Querying, SIGGRAPH 1995

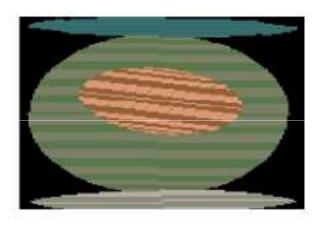
Iconic matching

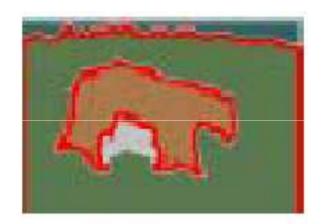


Example taken from Jacobs, Finkelstein, & Salesin Fast Multi-Resolution Image Querying, SIGGRAPH 1995

Region-based retrieval: Blobworld



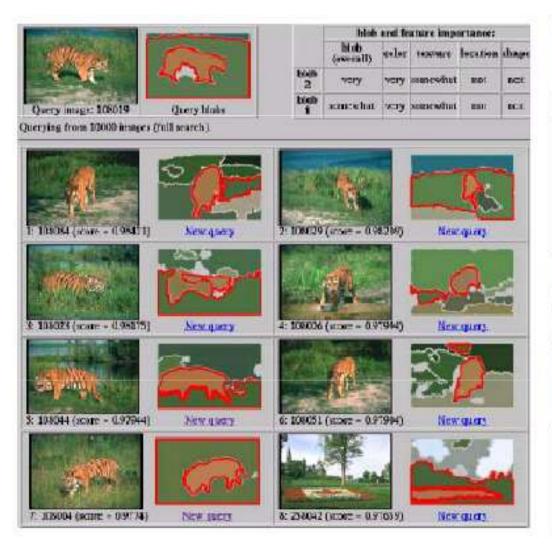


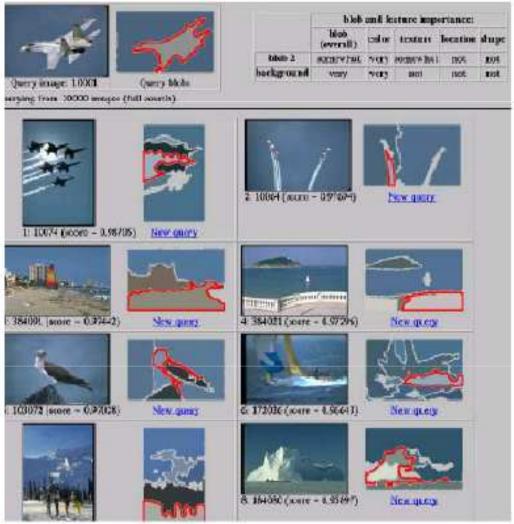


- Images are segmented on color plus texture
- User selects a region of the query image
- System returns images with similar regions
- Works really well for tigers and zebras

Demo: http://elib.cs.berkeley.edu/photos/blobworld

Region-based retrieval: Blobworld

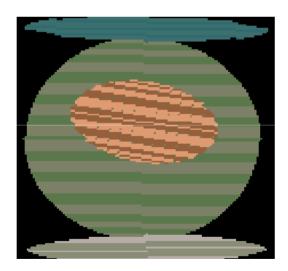




Retrieval using spatial relationships

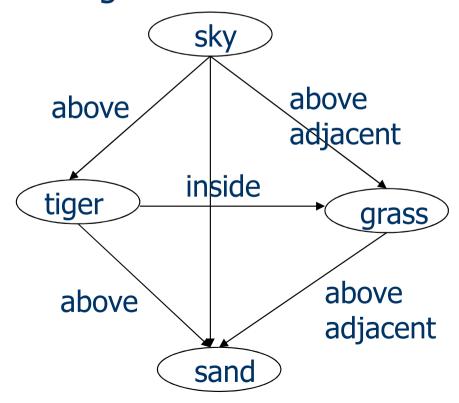


image



abstract regions

- Build graph using regions and their spatial relationships.
- Similarity is computed using graph matching.



Combining multiple features

Text query on "rose"















Combining multiple features

Visual query on













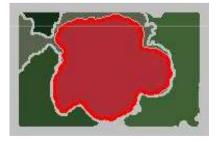






Combining multiple features

Text query
on
"rose"
and visual query
on





Video Google: object matching



36

Video Google



Viewpoint invariant descriptors

Visual vocabulary

CS 484, Fall 2012 ©2012, Selim Aksoy 37

Video Google

Document 1

Now is the time for all good men to come to the aid of their country.

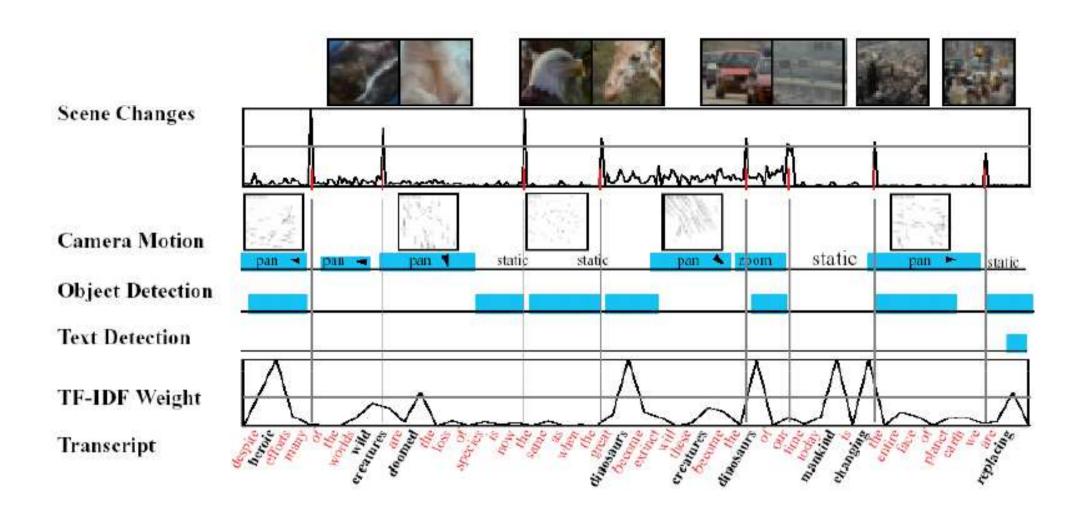
Document 2

Summer has come and passed. The innocent can never last.

Inverted index

Word	Document
aid	1
all	1
and	2
can	2
come	1, 2
country	1
for	1
good	1
the	1, 1, 2

Video skimming



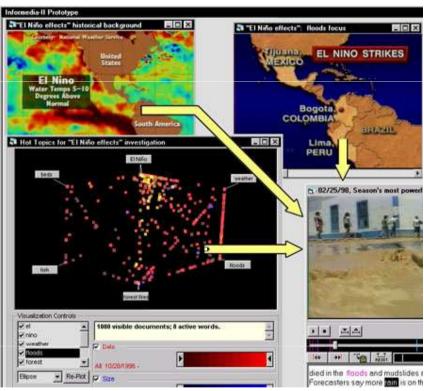
Event detection, indexing, retrieval

Assign semantic labels to significant events in video:

- Explosion, car crash, door slam (audio/video track)
- Marilyn Monroe enters scene
- Pele scores goal
- Jay Leno tells joke and then delivers punch-line
- Two people exchange a briefcase in park
- etc.

Informedia Digital Video Library





IDVL interface returned for "El Nino" query along with different multimedia abstractions from certain documents.

Informedia Digital Video Library



IDVL interface returned for "bin ladin" query.

The results can be tuned using many classifiers.

In real interactive CBIR systems, the user should be allowed to interact with the system to "refine" the results of a query until he/she is satisfied.



Example methods:

- Query point movement
 - Query point is moved toward positive examples and moved away from negative examples.
- Weighting features
 - The CBIR system should automatically adjust the weight that were given by the user for the relevance of previously retrieved documents.
- Weighting similarity measures
- Feature density estimation
- Probabilistic relevance feedback

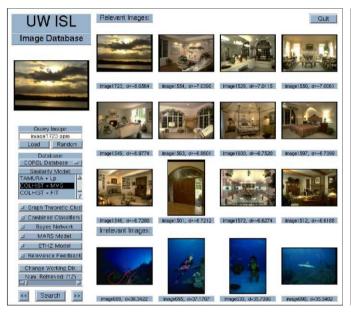
Positive feedback

$$p(A \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n)}^+) \propto p(\xi_{(n)}^+ \mid A) p(A \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n-1)}^+)$$

$$p(B \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n)}^+) \propto p(\xi_{(n)}^+ \mid B) p(B \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n-1)}^+)$$

Negative feedback

$$\begin{split} p(A \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n)}^+, \xi_{(1)}^-, \dots, \xi_{(m)}^-) & \propto \\ p(\xi_{(m)}^- \mid B) p(A \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n)}^+, \xi_{(1)}^-, \dots, \xi_{(m-1)}^-) \\ p(B \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n)}^+, \xi_{(1)}^-, \dots, \xi_{(m)}^-) & \propto \\ p(\xi_{(m)}^- \mid A) p(B \mid \xi_{(0)}, \xi_{(1)}^+, \dots, \xi_{(n)}^+, \xi_{(n)}^-, \dots, \xi_{(m-1)}^-) \end{split}$$



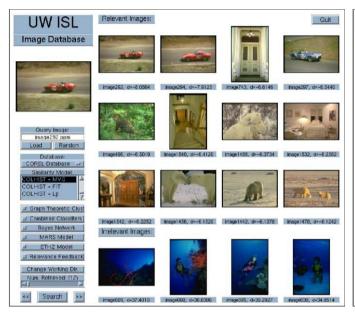




"Sunsets" using color histograms (1/12)

Using combined features (6/12)

After 1st feedback (12/12)





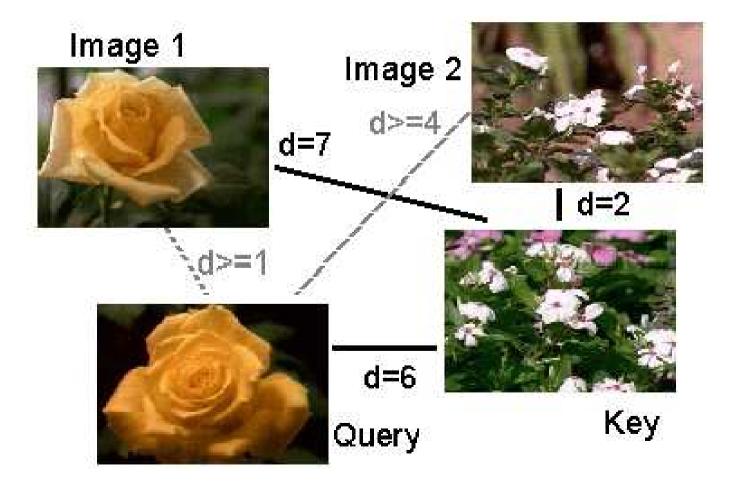


"Auto racing" using color histograms (3/12)

Using combined features (9/12)

After 1st feedback (12/12)

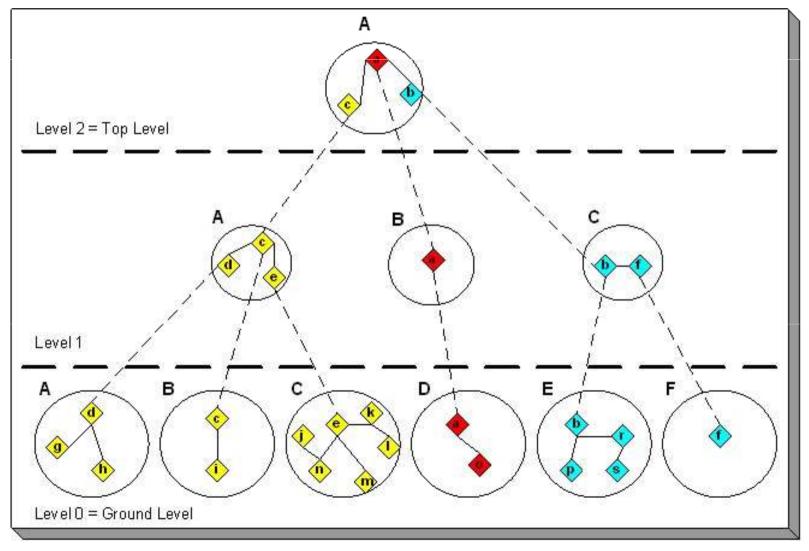
 Use of key images and the triangle inequality for efficient retrieval.



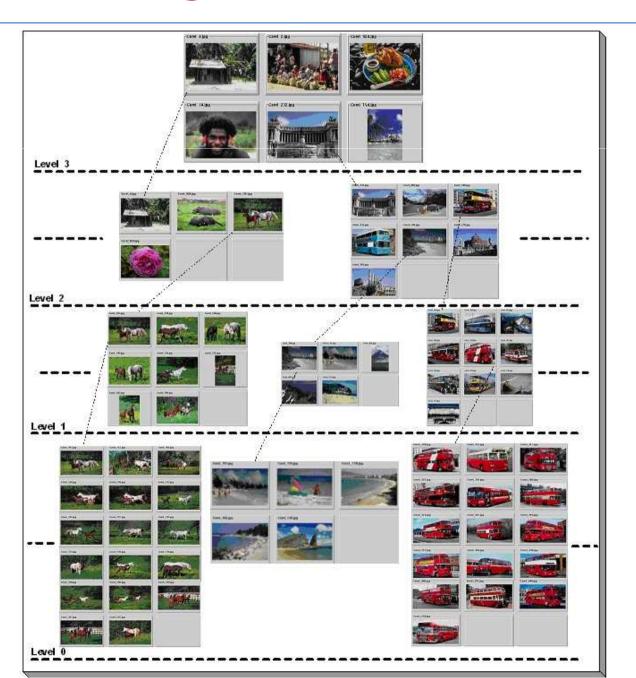
Offline

- 1. Choose a small set of key images.
- 2. Store distances from database images to keys.
- Online (given query Q)
 - 1. Compute the distance from Q to each key.
 - 2. Obtain lower bounds on distances to database images.
 - Threshold or return all images in order of lower bounds.

Hierarchical cellular tree



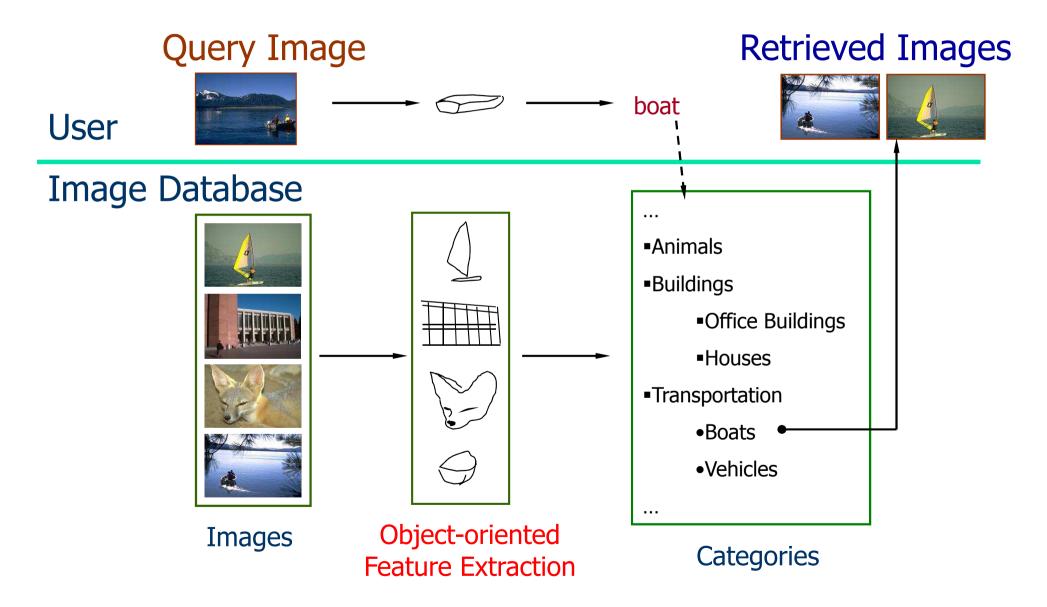
CS 484, Fall 2012 ©2012, Selim Aksoy 50



Performance evaluation

- Two traditional measures for retrieval performance in the information retrieval literature are precision and recall.
- Given a particular number of images retrieved,
 - precision is defined as the percentage of retrieved images that are actually relevant, and
 - recall is defined as the percentage of relevant images that are retrieved.

Current research objective



Demos

- Blobworld (http://elib.cs.berkeley.edu/blobworld/)
- Video Google (http://www.robots.ox.ac.uk/~vgg/ research/vgoogle/index.html)
- FIDS (http://www.cs.washington.edu/research/ imagedatabase/demo/fids/)
- Like Visual Shopping (http://www.like.com/)
- Google Image Search (http://images.google.com/)
- Yahoo Image Search (http://images.search.yahoo.com/)
- Flickr (<u>http://flickr.com/</u>)
- The ESP game (http://www.espgame.org/)

Demos

- Vitalas
 - http://vitalas.ercim.eu/
- Google Similar Images
 - http://googleblog.blogspot.com/2009/10/similarimages-graduates-from-google.html
- Google Image Swirl
 - http://googleresearch.blogspot.com/2009/11/exploreimages-with-google-image-swirl.html
- Microsoft Bing
 - http://www.bing.com/
 First use keywords, then mouse over an image and click on show similar images