

# ECE 417 MP6

## Content-Based Image Retrieval & Relevance Feedback

ECE 417 – Multimedia Signal Processing  
University of Illinois at Urbana-Champaign  
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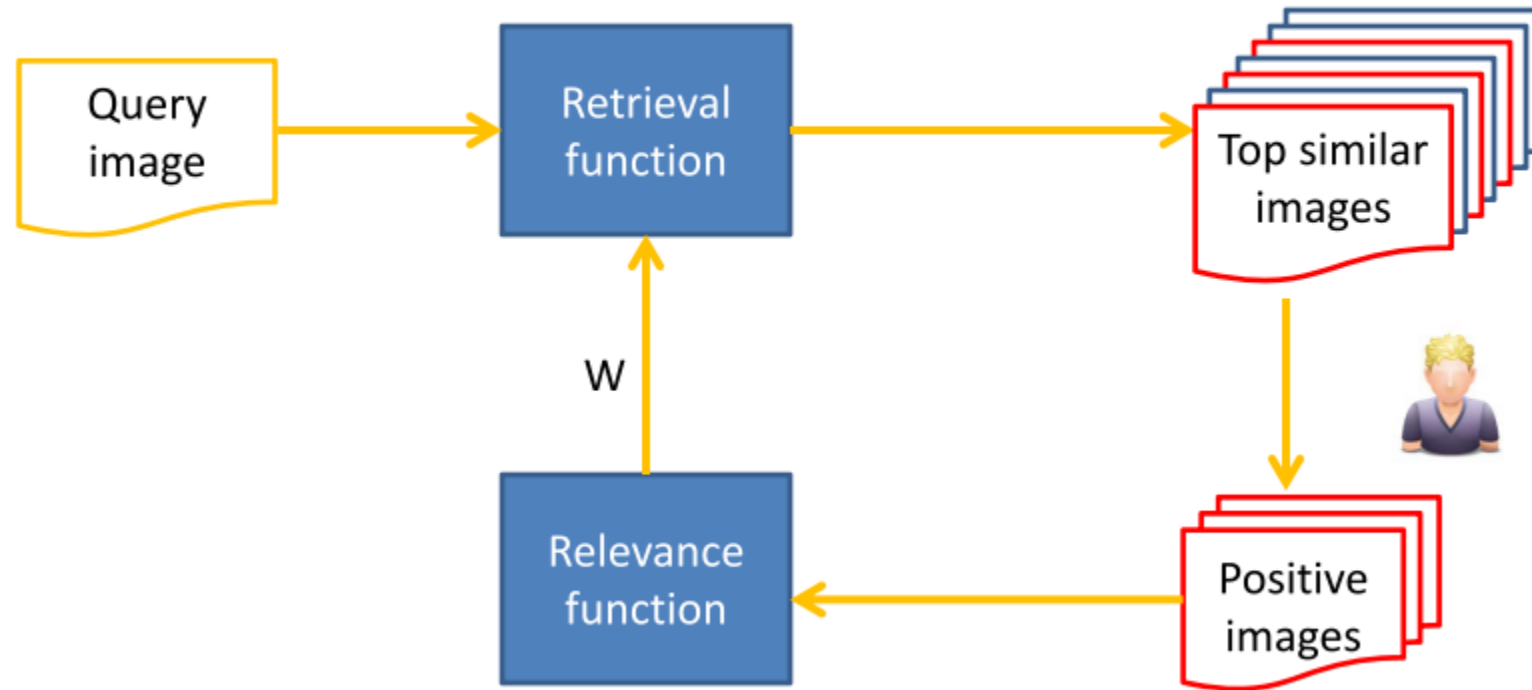
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Prepared by Po-Sen Huang

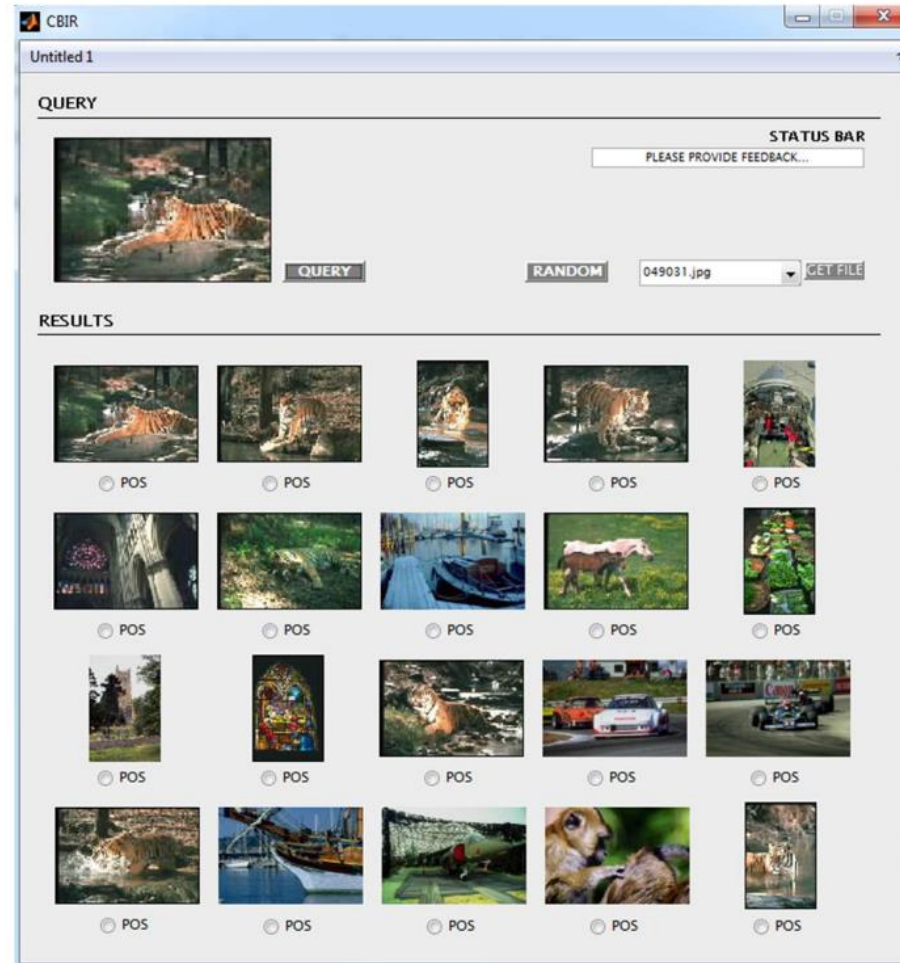
# Overview

- Data
  - You are given 1400 images with their 47-dim visual feature (so you do not need to worry about feature extraction)
  - These image features are stored in [handles.META\_DATA]
- Tasks
  - Implement the retrieval function for a set of query images (3.1)
  - Implement the relevance feedback function (3.2)

# Flow Chart

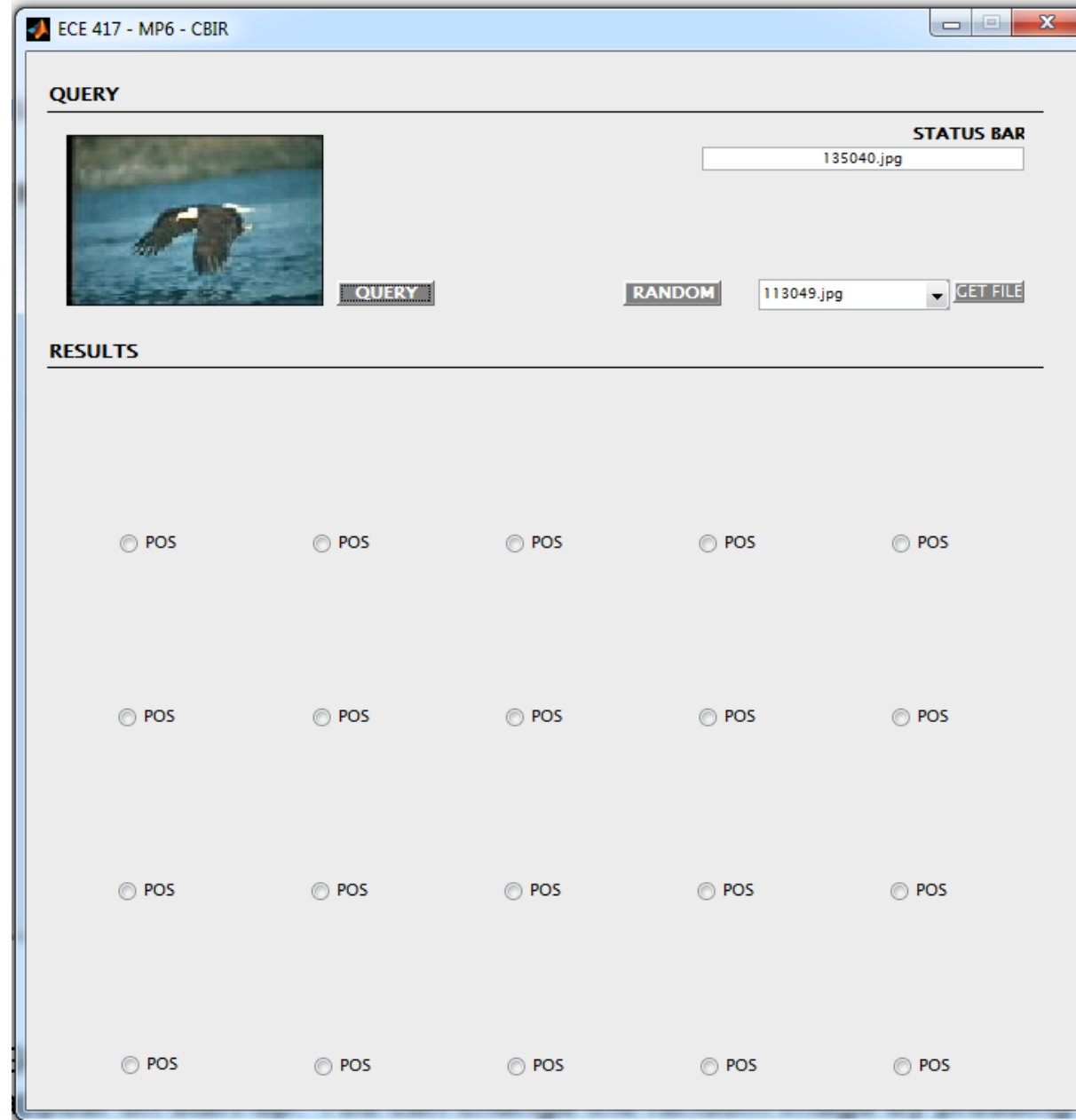


# Graphical User Interface (GUI)



# Initially

- Run cbirMP
- Buttons:
  - Random
  - Get File
  - **QUERY**
  - **POS**



# GUI Shell

- Each interactive component in the GUI (radio buttons, push-buttons, etc.) is controlled by a *call-back function*
- Every time a user-event is detected (mouse-click, hover, etc.) the appropriate call-back function is called
- For example, once the '**QUERY**' button is pushed, the functionality in the query-button's call-back function will be called
- Variables within a call-back function are **local**
- MATLAB uses the '**handles**' structure to share data globally

# handles

- Initialize to hold all the components in the GUI
- To get the positive indices, type:  
`>>handles.posInds;`
- Add a local variable K to be available for all the call-back functions, type:  
`>>handles.K = K;`  
`>>guidata(hObject,handles);` % **guidata** function saves the **'handles'** structure globally

# Important Global Variables

<b>name</b>	<b>type</b>	<b>description</b>
<b>handles.META_DATA</b>	[47 x 1400 double]	47-dimensional image feature for each of 1400 images in the database.
<b>handles.W</b>	[47 x 47 double]	Relevance feedback weighting matrix
<b>handles.posInds</b>	[1 x (variable) double]	Indices of images in the database that have thus far been labeled as relevant
<b>handles.currentTopInds</b>	[1 x (variable) double]	Indices of the Top 20 returned images given a particular query



# Task 1: Retrieval function (I)

- Given several query images (including those from user feedback), return top 20 images that are visually similar to the query images
- INPUT:
  - Image features for all images (`handles.META_DATA`),  
Indices of the query images (`handles.posInds`)
- OUTPUT:
  - Rank-sorted list of top 20 indices (`handles.currentTopInds`)

# Task 1: Retrieval function (II)

- 3 Subtasks:

- 1) Compute the mean image-feature vector  $q_c$  for the query images (given as the indices in **handles.posInds**)
- 2) Compute the Mahalanobis distance between the mean query images vector to each of the images in the database

$$D(j) = (q_c - x_j)^T W (q_c - x_j)$$

where  $q_c$  is the mean query-images feature,  $x_j$  is the  $j$ -th database image feature vector

- 3) Sort distances in  $D(j)$  to get the indices of the top 20 similar images and return the indices to **handles.currentTopInds**

- Note: To complete the 2<sup>nd</sup> subtask, you need the weighting matrix from task 2. You can first implement Task 2 by just returning a 47x47 identity matrix from the relevance feedback function

# Task 2: Relevance feedback function (I)

- Given a set of similar images (from user's feedback), find a diagonal weighting matrix such that the features with smaller variances have larger weights and vice versa
- INPUT:
  - Image features for all images (handles.META\_DATA)
  - Indices of the query images (handles.posInds)
- OUTPUT:
  - A diagonal weighting matrix ( $W$ )
- NOTE: if there is no image from user feedback, you have to return a 47x47 identity matrix

# Task 2: Relevance feedback function (II)

Retrieval function calls **RF(handles)** to analyze the features for images already labeled as positive in order to compute a weighting matrix  $W$

- 1) Determine whether the feedback is round 0 (initial seed image)
- 2) If round 0, return a 47x47 identity matrix
- 3) If not, for each feature dimension (47 in total)
  - Obtain the value of this feature for each of the positively-labeled images (**handles.posInds**)  
The end result should be a list of  $K$  values, where  $K$  is the length of **handles.posInds**
  - Compute the variance of this sample and add **0.0222** (for regularization)
  - Form and return a diagonal weighting matrix **W** such that:

$$W(i, i) = \frac{1}{\sigma_i^2 + 0.0222}$$

# Extra credit: Full weighting matrix

- Can you ‘find’ a full weighting matrix?
  - Inverse of covariance matrix
    - Any problem?
    - Regularized inverse of covariance matrix:  $\text{inv}(XX^T + aI)$
  - PCA
    - Given a set of images, find the projection matrix  $P$  such that principal components (those directions with high variance) are removed
    - $W = P^T P$  ( $P$  is a  $d$ -by-47 matrix)

# Extra credit 2: Using negative feedbacks

- Can you exploit negative feedbacks as well?
  - Implicit negative feedback: those images not being selected as positive
    - Assuming all the positive images are selected as positive
- How?
- LDA:
  - Given a few of positive and negative images, find the projection matrix  $P$  that best separate the two groups
  - Again,  $W = P^T P$  ( $P$  is a  $d$ -by-47 matrix)

# Experiments

- Conduct five interactive retrieval sessions (with different seed images as provided in the drop-down menu) to explore the effect of relevance feedback on retrieval performance
- Retrieval accuracy will be quantified by ***Precision***
- Precision is defined as the proportion of relevant images returned to total number of images returned (in this case 20). For each of the seed images, you will provide a plot showing precision as a function of feedback round for 3 rounds of search (1 initial + 2 relevance feedback). You may put all plots on the same figure.

# Deliverables

- MATLAB code for a fully-functional CBIR GUI. Both empty functions should be complete, and the program should work properly
- Quantitative results in plot form for your interactive retrieval experiments (Precision versus feedback round)
- Qualitative comments about your results. Specifically:
  - What is the general trend of precision versus interaction round? (Or, what should it be?)
  - Why do you think certain plots are different than others? (from an image representation perspective)



# MATLAB Hint

- Keep one GUI open at a time
- Use break-points to trace the codes, since the order-of-execution is more complex in GUI