

Master in Computer Vision Barcelona

Module: M1

Project: Content Based Image Retrieval

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M1 – Content Based Image Retrieval

Main goal

 To learn the basic concepts and techniques to build a simple query by example retrieval system for finding paintings in a museum image collection.



Can Framis museum dataset

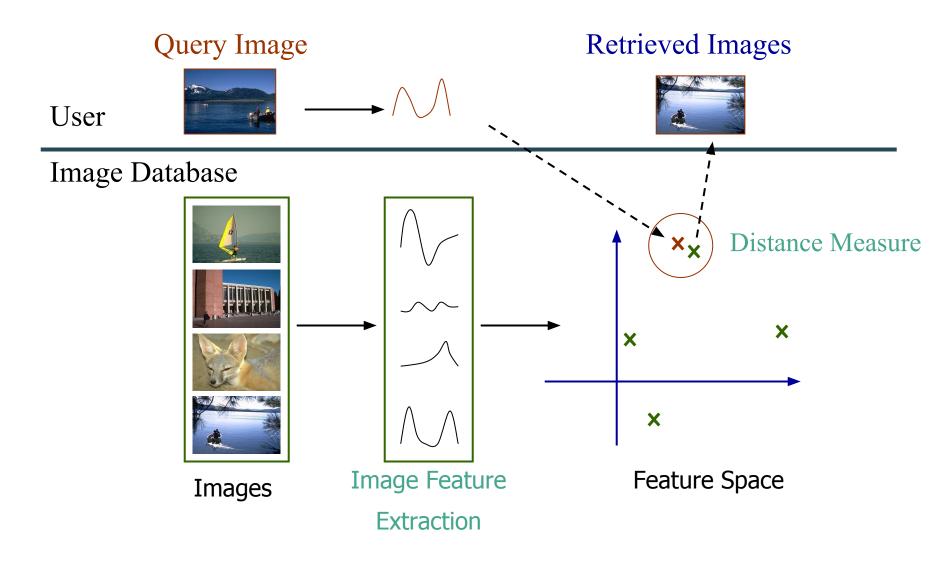
Scope

- Image retrieval based on color, texture, textual information
- Image retrieval based on keypoints and local descriptors
- Denoising, orientation correction and picture cropping
- Morphological filters to detect and remove overlying text from images
- Evaluation of system performance

Applicability

Almost any (small) query by example problem

M1 – Content Based Image Retrieval



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M1 – Content based image retrieval

- Goal: search images from a large image database (DB) based on visual contents
- Similarity: based on visual features: color, shape texture
 - e.g. color:
 - how to characterize 'color'?
 - how to measure 'color similarity'?

CBIR steps:

- Index the DB: generate descriptors (e.g. histograms) for all images; this can be done offline
- Extract features from the query image (e.g. compute histogram)
- Compute distances between descriptors from the query image and each
 DB image descriptor, and order the DB images according these distances

M1 – Dataset

Paintings dataset







• Query set 1 (QS1): cropped images, almost no rotation







Query set 2 (QS2): uncropped image (background)



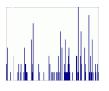
- Given the museum and the query dataset.
 For each image in the query dataset, retrieve the K most similar images in the Museum dataset, ordered by score
- Image descriptor: image gray-level / color histograms (1D)
- **Task 1:** Create Museum and query image descriptors (BBDD & QS1)
- Task 2: Implement / compute similarity measures to compare images
- **Task 3:** Implement retrieval system (retrieve top K results)
- Task 4: Evaluation using map@k
- **Task 5:** Background removal using color (QS2). Compute descriptor on foreground (painting)
- **Task 6:** Evaluation of picture masks and retrieval system (QS2)

Task1: Compute image descriptors (QS1)

- Color Histogram:
 - gray level / concatenate color component histograms
 - color space RGB, CieLab, YCbCr, HSV









Task 2: similarity measures

- Euclidean distance
- L1 distance
- χ² distance
- Histogram intersection (similarity)
- Hellinger kernel (similarity)

$$D(h_1, h_2) = \sqrt{\sum_{i=1}^{N} (h_1(i) - h_2(i))^2}$$

$$D(h_1, h_2) = \sum_{i=1}^{N} |h_1(i) - h_2(i)|$$

$$\sum_{i=1}^{N} (h_1(i) - h_2(i))^2$$

$$D(h_1, h_2) = \sum_{i=1}^{N} \frac{\left(h_1(i) - h_2(i)\right)^2}{h_1(i) + h_2(i)}$$

$$I(h_1, h_2) = \sum_{i=1}^{N} \min(h_1(i), h_2(i))$$

$$K(h_1, h_2) = \sum_{i=1}^{N} \sqrt{h_1(i) h_2(i)}$$

Task 3: For each image in QS1, compute similarities to museum images

Task 4: Evaluation

- Return the top k images (highest score, lowest distance)
- Evaluation with MAP@k (Mean Average precision at K)
 - Average precision computed for each query
 - Mean over all queries

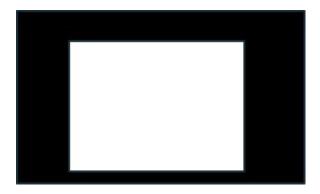
Metric available at: https://github.com/benhamner/Metrics

$$AP@K = \frac{\sum_{i=1}^{K} P@i}{K}$$

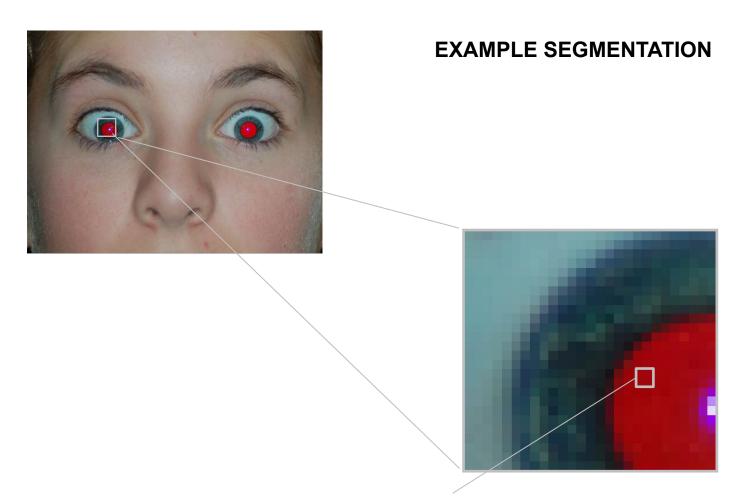
Task 5: For each image in the QS2, remove background using the background color (e.g. model background distribution, color thresholds)

- Create a binary mask to evaluate the method
- Compute the descriptors on the foreground pixels.
- Do not use contour detectors, object detectors, etc. Just color!









Pixel value can be between 0 and 255 (in a typical 8 bits/channel image) This pixel is (R,G,B)=(169,5,9)

##UPC

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EXAMPLE SEGMENTATION

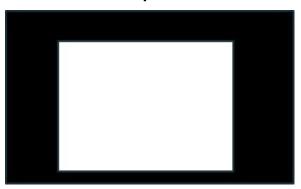
EXAMPLE SEGMENTATION





Task 6: Evaluation

- For background removal: precision, recall, F1

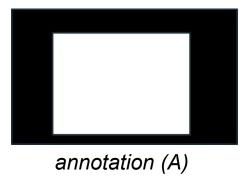


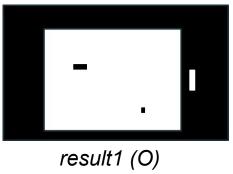
- For retrieval system: MAP@k (Mean Average precision at K)





Mask evaluation using precision, recall and F1-measure



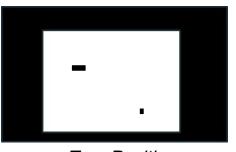


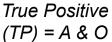


$$P = \frac{TP}{TP + FP}$$

$$R = \frac{TP}{TP + FN}$$

$$F_1 = 2 \frac{PR}{P+R}$$







False Negative $(FN) = A \& \neg O$



False Positive (FP) $= \neg A \& O$



True Negative (FP) $= \neg A \& \neg O$



	map@1	map@5
Method 1		
Method 2		
Method 3		

Provide map@k (k=1, k=5) results for the validation query sets 1&2 (QSD1 & QSD2)

	Precision	Recall	F1-mesure
Method 1			
Method 2			
Method 3			

Provide results for the validation query set 2 (QSD2)

Query sets:

There will be four query sets:

- Two for development (QSD1, QSD2), already available
 - Ground truth provided: correspondences & masks
 - Results for these query sets must appear on the slides
- Two for test (QST1, QST2), released on Friday (October 15) at 10h
 - Ground truth not provided
 - Teams submit 'blind' results for these query sets and we score them automatically

Submissions:

- For each test query in QST{1,2}, a list of the K=10 best results:
 - Create a python **list of lists**, with the image ids (integer numbers)
 - Example with 3 queries and k=2

Query: [[q1], [q2], [q3]]

Result: [[7,2], [76, 4], [43, 12]]

where 7 corresponds to image 00007.jpg, etc.

Note: Deliver pkl files to GDrive folder:

TeamX/week1/QST1/method1/result.pkl

TeamX/week1/QST1/method2/result.pkl

- For QST2, the binary masks must also be submitted TeamX/week1/QST2/method1/{result.pkl, *.png} TeamX/week1/QST2/method2/{result.pkl, *.png}
- Submit progress slides
- Deadline results and slides: Friday 15 Oct 2021 at 16:00
- Deadline questions to teams: Monday 18 Oct 2021 at 12:00