

Midterm for Multimedia Analysis and Indexing (MMAI) – Fall 2007

Name: _____ School ID: _____

Midterm (2:20 pm, Tuesday, Nov. 6, 2007)

Note:

- (1) Write down your NAME and School ID in the booklet and the test problem set (this paper). You have to submit them **BOTH** to TA after the midterm.
- (2) Best luck to you!

1. Co-occurrence texture: (20%)

Please show the three co-occurrence matrices $C_d(i, j)$, specified by different displacement vectors d ((c)-(e)), for a gray image I , which has the gray value in each pixel.

- 5 (1) What are the dimensions of the co-occurrence matrices? 4×4
- 5 (2) Please show the three co-occurrence matrices with $d=(0,1)$, $d=(1,0)$, and $d=(1,1)$.

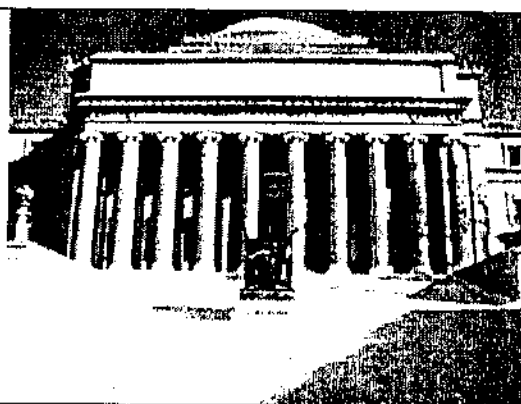
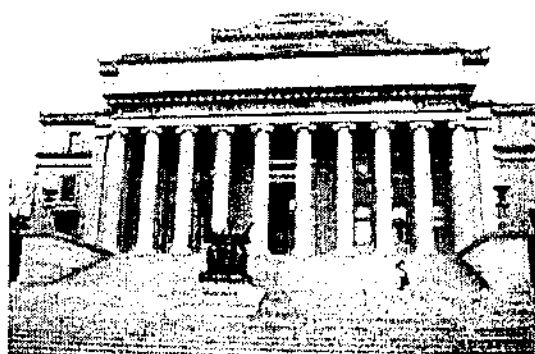
<table><tr><td>0</td><td>0</td><td>2</td><td>3</td></tr><tr><td>1</td><td>0</td><td>3</td><td>2</td></tr><tr><td>2</td><td>1</td><td>0</td><td>0</td></tr><tr><td>1</td><td>2</td><td>1</td><td>0</td></tr></table>	0	0	2	3	1	0	3	2	2	1	0	0	1	2	1	0	<div><div>... j ...</div><div><div>...</div><div></div></div></div>	<div><div>$d=(0,1)$<table><tr><td>i</td><td>j</td></tr></table></div><div>$d=(1,0)$<table><tr><td>i</td></tr><tr><td>j</td></tr></table></div><div>$d=(1,1)$<table><tr><td>i</td></tr><tr><td>j</td></tr></table></div></div>	i	j	i	j	i	j
0	0	2	3																					
1	0	3	2																					
2	1	0	0																					
1	2	1	0																					
i	j																							
i																								
j																								
i																								
j																								
(a) Image I	(b) $C_d(i, j)$	(c)-(e)																						

	0	1	2	3
0	2	0	1	1
1	3	0	1	0
2	0	2	0	1
3	0	0	1	0
	$d(0,1)$			
	0	1	2	3
0	2	3	0	0
1	0	0	2	0
2	1	1	0	1
3	1	0	1	0
	$d(1,0)$			
	0	1	2	3
0	3	0	0	1
1	0	2	0	0
2	0	0	2	0
3	1	0	0	0

II. Image Similarity and Near-Duplicate Detection (20%)

We are to design a system to *efficiently* and *effectively* discover the near-duplicates (picture pairs taken at the same site or with a very similar scene. See the following two example images.) from a very **LARGE** databases (i.e., Flickr) with around N user-contributed images ($N \sim 250$ millions). Given an image in the database, the system needs to show the top K near-duplicate images in the database and their corresponding near-duplicate scores. You are allowed to compute the near-duplicate similarities offline.

- (1) What might be good candidates for feature representations of images and their corresponding distance (or similarity) measures? Why?
- (2) What is the time complexity to compute all the pair-wise similarities in the image database? Any algorithms to speed up the near-duplicate detection (e.g., filtering out irrelevant images first with low-cost image similarity measures)? What might be the impact of your proposed approach in reducing the time complexity? Any tradeoff for accuracy?
- (3) With your proposed algorithm, what might be the estimated time complexity if a new image is inserted into the database (N images)?



Near-duplicates from the images taken by different persons in front of the Low Library of Columbia University, New York.

III. Precision and Recall (25%)

In the retrieved image list of depth 15, we had inspected the retrieved images and marked the hit (+) at different depths. There are totally **10** ground-truth images. Please

- (1) Complete the "recall" at different depths. Note that at depth i , we are to evaluate the retrieved results of the first i images (ranked 1 to i).
- (2) Complete the "precision" at different depths
- (3) Please draw the precision and recall curve of the retrieved results (at depth 15). Please use **ONLY** the recall and precision values where a "hit" occurs.

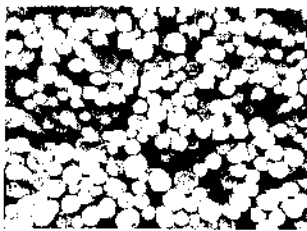
#	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
Hit	-	+	+	-	+	+	-	-	+	+	+	+	-	+	+
Recall	0	0.1	0.2	0.2	0.3	0.4	0.4	0.4	0.5	0.6	0.7	0.8	0.8	0.9	1
Precision	0	0.5	0.67	0.5	0.6	0.67	0.57	0.5	0.56	0.6	0.64	0.67	0.62	0.64	0.67

IV. Please briefly explain (10%)

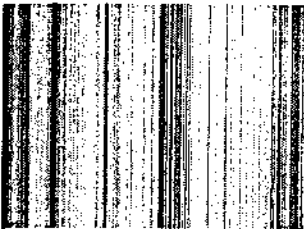
- (1) What's the *semantic gap*?
- (2) What's the *curse of dimensionality*?

V. Gabor Textures (25%)

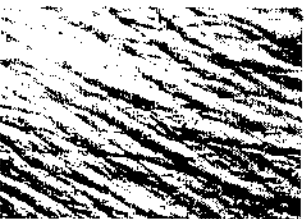
The following are the images and their corresponding plots of Gabor features in terms of the **mean** over the whole response map after being convoluted by a Gabor filter, specified by a scale and an orientation factor. In the texture plot, the Y-axis (\uparrow) represents (4) *scales* and the X-axis (\rightarrow) represents (6) *orientations*. The light color represents high intensity. Please find the best match between the images and features. Please shortly justify your answer.



(1) *d*



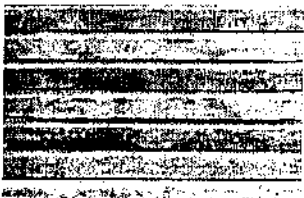
(2) *e*



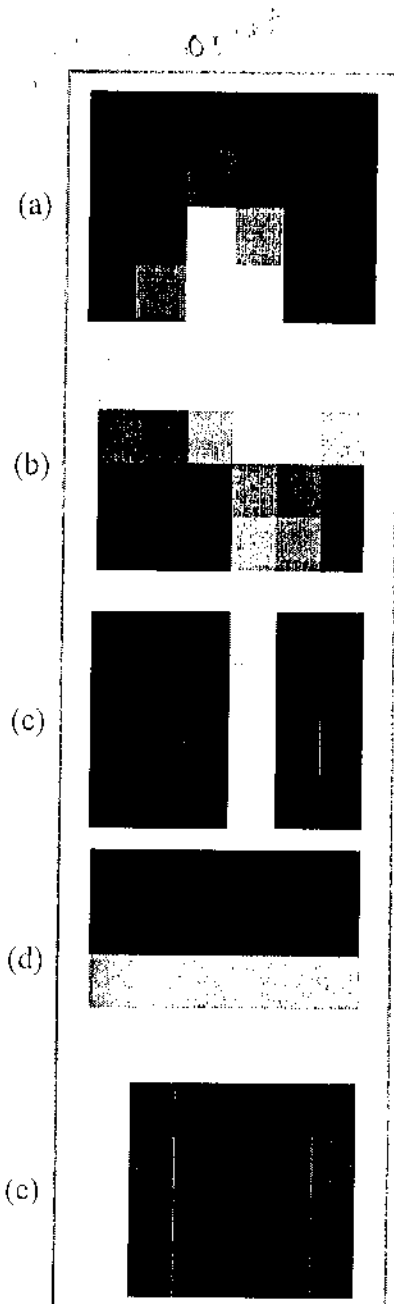
(3) *a*



(4) *b*



(5) *c*



Multimedia Analysis and Indexing – Spring 2007

Name: _____ School ID: _____

Midterm (2:30 pm, Wednesday, May 16, 2007)

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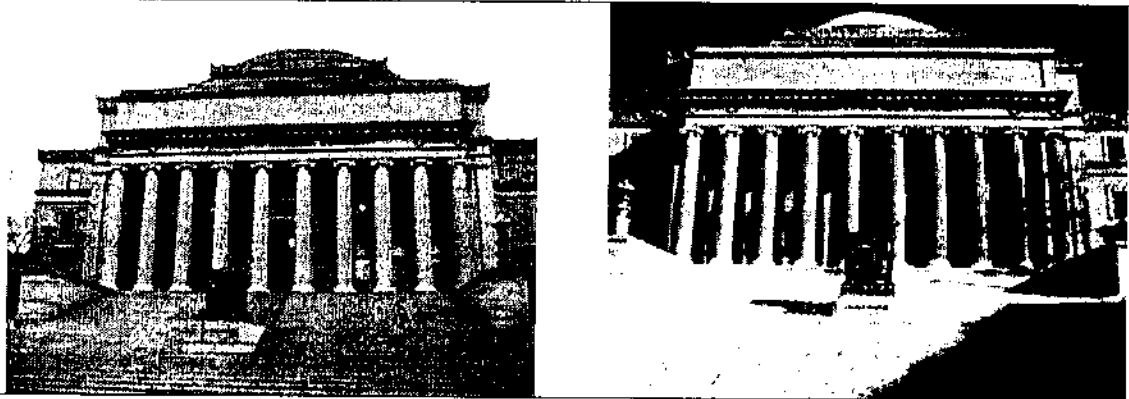
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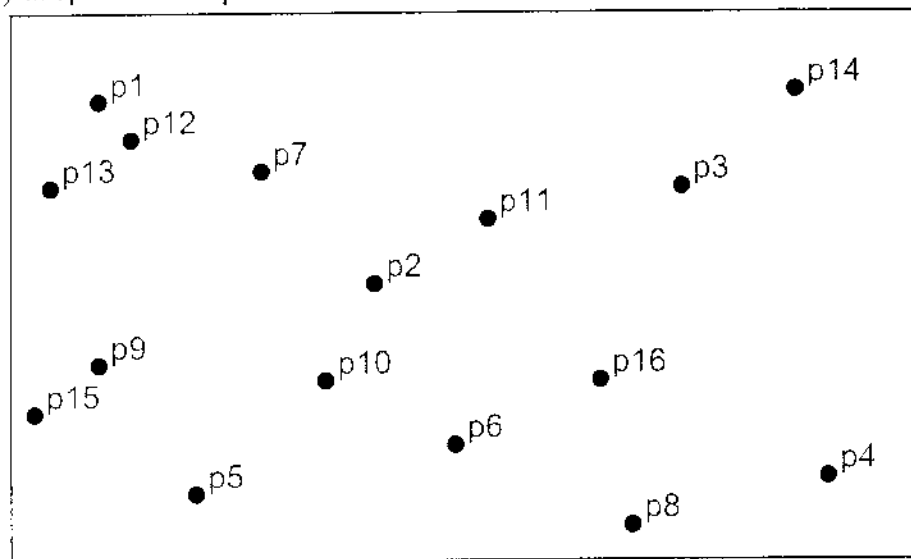
[illegible]

IV. Quad-tree and KD-tree (25%)

We have 16 points in a 2-dimensional space. Please organize these points by **space decomposition** and their corresponding **tree structures** in the following two methods:

- (1) *Quad-tree*, which splits the space into 4 equal subspaces and continues until each leaf node has a single point.
- (2) *KD-tree*, which recursively subdivides points into two halves (median values are used) using vertical and horizontal lines till one point left in each leaf.

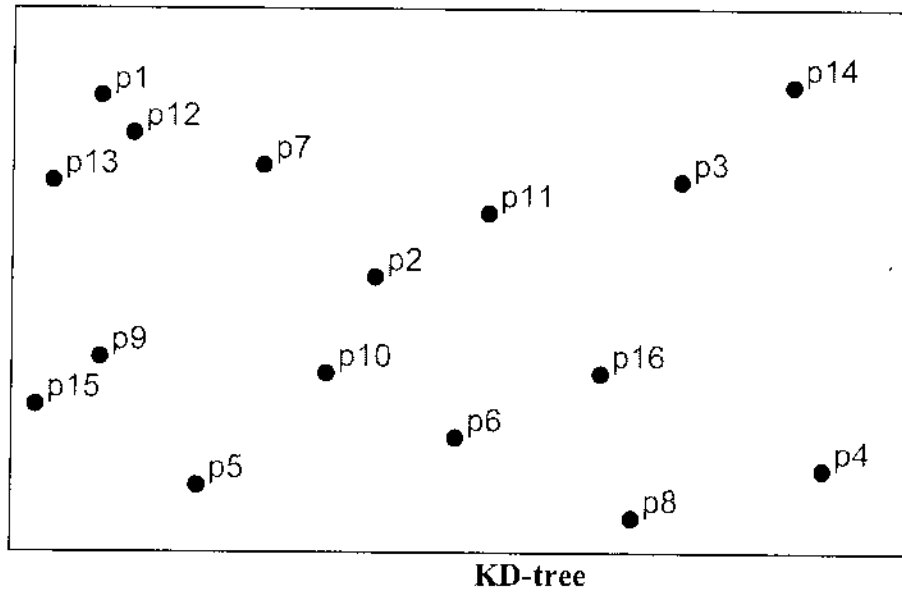
(1)-a: space decomposition



Quad-tree

(1)-b: tree representation in Quad-tree

(2)-a: space decomposition



(2)-b: tree representation in KD-tree