

Indexing spatially sensitive distance measures
using multi-resolution lower bounds

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Photoreceptors
(labeled by anti-rhodopsin)

Macrophages
(labeled by isolectin B4)

Confocal micrograph of cat retina
(by Geoff Lewis, Fisher lab, UCSB)

Microglia and blood vessels
(labeled by isolectin B4)

Müller cells
(labeled by anti-gliial fibrillary acidic protein)

Light

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Which image is most similar to image A—is it B or C?

Database images

Query image

B

C

L₂: d = 7.5
EMD: d = 37

L₂: d = 8.7
EMD: d = 23

L₂ thinks A and C are more similar.
(Biologists disagree!)

Earth mover's distance

The earth mover's distance (EMD)
[Werman et al., 1985; Peleg et al., 1989; Rubner et al. 2000]

Flow f_{ij} from every region of one image to every region of the other

Ground distance c_{ij} captures how far the "mass" moves

EMD = $\min_f \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} c_{ij} w^T f_{ij}$

LP-problem: subject to $f_{ij} \geq [0 \dots 0]^T$ (element-wise), w : weights

$\forall i, \sum_{j=0}^{n-1} f_{ij} = a_i, \quad \forall j, \sum_{i=0}^{n-1} f_{ij} = b_j$

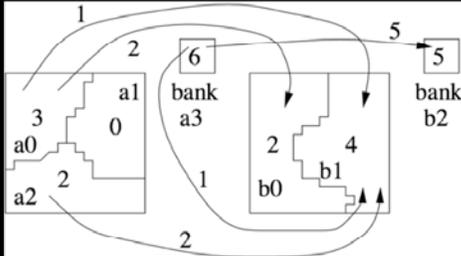
... but it's slow!

Outline

- EMD on feature vectors
 - Formulation with special "bank" region
 - Decompose EMD on feature vectors into many smaller LP-problems
 - Faster, uses less memory
- Lower bound for EMD
 - Spatially motivated
 - Faster to compute, using summary of image
 - Multiple resolutions
 - Range and k-NN query algorithms that use the lower bounds
 - Sequential scan and M-tree index structure
- Experimental results

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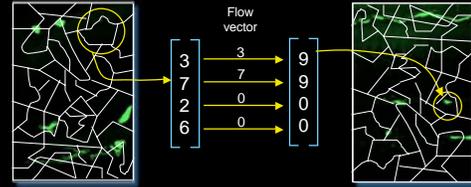
Introducing the "bank" region



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Decomposing the EMD

Flows are between the same dimensions of the feature vectors.



In other words, there is no crosstalk.

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Reducing crosstalk

- No crosstalk between independent dimensions
 - Color Layout Descriptor: DCT coefficients
 - Orthogonal bases found by PCA
- Sometimes crosstalk only between some dimensions
 - Concatenated feature vectors for two independent proteins
 - Cluster dimensions, no crosstalk between dimensions in different clusters

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Decomposition makes the LP-problem smaller

$$\min_F \sum_{i=0}^n \sum_{j=0}^n c_{ij} \sum_{k=1}^m w_k f_{ijk} = \sum_{k=1}^m \min_{F_k} \sum_{i=0}^n \sum_{j=0}^n c_{ij} w_k f_{ijk}$$

97 regions (left and right) and 12 dimensions (bottom center) are indicated by arrows.

Large LP problem
41 s
37 MB main memory

12 small LP problems
2.9 s in total
5 kB main memory

97 x 97 x 12 variables

97 x 97 variables for each problem

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Decomposition helps, but is not enough

$$\rho_{AB} = \sum_{k=1}^m \min_{F_k} \sum_{i=0}^n \sum_{j=0}^n c_{ij} w_k f_{ijk}$$

12 dimensions

97 regions (8 x 12 tiles + bank)

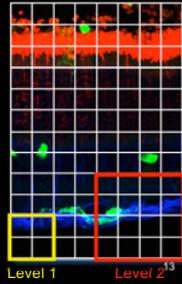
Decomposed:
97 x 97 variables, 12 dimensions

Number of variables quadratic in the number of regions

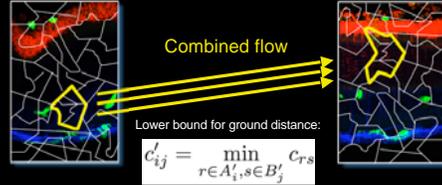
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Spatially motivated lower bound

- Cannot just use larger regions
 - Regions must be small enough to fit in layers of tissue
- Idea
 - Compute distance using larger regions
 - Modify the distance function so this distance is a lower bound for the EMD using smaller regions
 - Multiple resolutions



Lower bound by assuming best-case ground distance



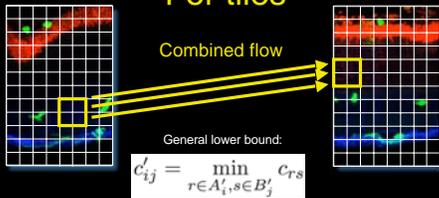
Lower bound for ground distance:

$$c'_{ij} = \min_{r \in A'_i, s \in B'_j} c_{rs}$$

$$\min_F \sum_{i=0}^n \sum_{j=0}^m c_{ij} w^T f_{ij} \geq \min_{F'} \sum_{i=0}^{n'} \sum_{j=0}^{m'} c'_{ij} w^T f'_{ij}$$

Proof is by decomposing flows and using $c' \leq c$ for each. 14

For tiles



General lower bound:

$$c'_{ij} = \min_{r \in A'_i, s \in B'_j} c_{rs}$$

For tiles, the ground distance has a simple formula:

$$d_{ij} = \begin{cases} [\max\{0, 2|l'_i - l'_j| - 1\}^2 + \max\{0, 2|l'_i - l'_j| - 1\}^2] & \text{if } i', j' \neq n' \\ 0 & \text{if } i' = j' = n' \\ \alpha & \text{otherwise} \end{cases} \quad (4)$$

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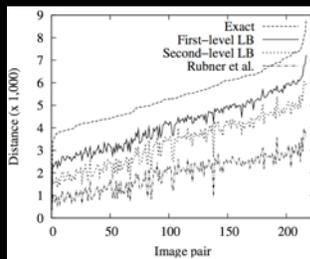
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Properties of the lower bound

Distance from image to each of 217 other images



Fast

- Full EMD: 2.9 s
- 1st level: 60 ms
- 2nd level: 4 ms

Tight

- 1st level: 25% lower
- 2nd level: 44% lower
- Rubner: 68% lower

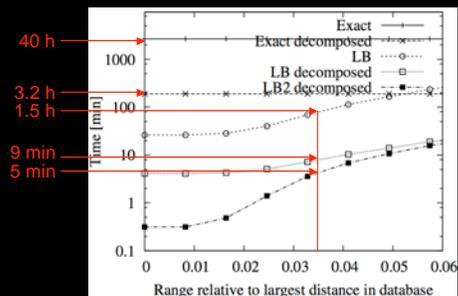
[Rubner et al., 2000]

8 x 12 tiles, Color Layout Descriptor (12-D)

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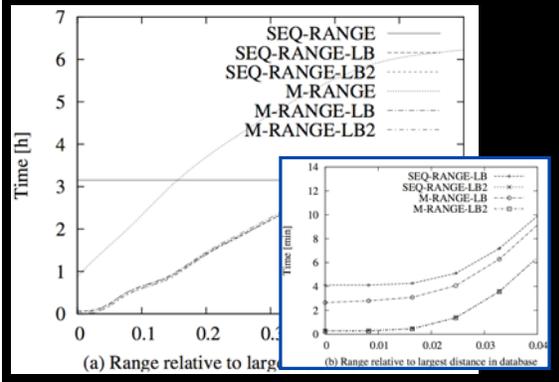
Effect of decomposition and lower bound

Range search on 3,932 retinal images

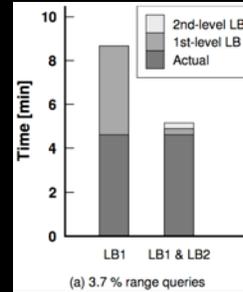


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Effect of lower bound and index structure — Range queries



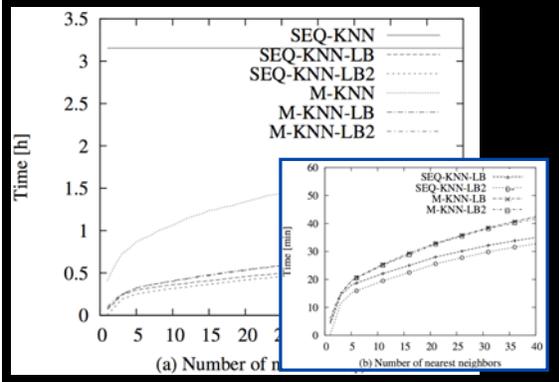
Breakdown of range query time



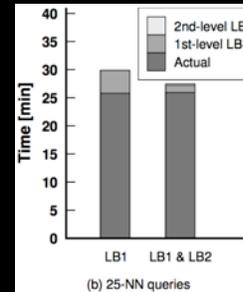
Search without lower bounds takes 3.2 h (not shown).

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Effect of lower bound and index structure — k-NN queries



Breakdown of 25-NN query time



Search without lower bounds takes 3.2 h (not shown).

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Summary

- EMD is a useful distance measure
 - Combines feature distance and spatial distance
- Techniques for speeding up EMD computation
 - Reduce search times up to 500 times
 - From 40 hours to 5 minutes
 - Make EMD viable, even when many regions are necessary
- Future
 - Integrate into BISQUE, the database infrastructure of the Bioimage project at UCSB (www.bioimage.ucsb.edu)
 - Other datasets
 - Crosstalk
 - Compare with new lower bounds (Assent et al., ICDE 2006)

Images by Geoff P. Lewis (Lab. of S.K. Fisher, UCSB)
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