# Scale-invariant Region-based Hierarchical Image Matching

## Sinisa Todorovic and Narendra Ahuja

**ICPR 2008** 



# PROBLEM STATEMENT

#### **Problem Statement**

Given a set of images

#### containing objects from an unknown category

where the images are captured

at varying distances from the objects

discover and segment all objects of the category

### Illustrative Example



input

output

Faces should be identified and segmented as the frequently occurring category in the input images Note that this is not a face-detection talk

## Challenges: Scale Variations



- Geometric and photometric properties of objects change
- Details visible in the high zoom-in, disappear at coarse scales

# MOTIVATION

## Motivation: Image = Segmentation Tree



multiscale segmentation

N. Ahuja 96, Tobb & Ahuja 97, Arora&Ahuja 06

## Motivation: Image = Segmentation Tree



multiscale segmentation

segmentation tree

N. Ahuja 96, Tobb & Ahuja 97, Arora&Ahuja 06

## Motivation: Learning Objects = Tree Matching

training



#### **Objects = Similar subtrees**

## Effect of Scale Changes



#### Width and depth of object subtrees varies

# PRIOR WORK

## Region Properties Associated with Each Node



 $\psi$  vector of region properties:

- Contrast with surround
- Area
- Displacement of centroids
- Orientation of principal axes
- Perimeter

## Region Properties Associated with Each Node



## $\psi$ vector of region properties:

- Contrast with surround
- Area
- Displacement of centroids
- Orientation of principal axes
- Perimeter

#### Area defined relative to parent's area ↓ Invariance to small scale variations

Todorovic & Ahuja 06, Ahuja & Todorovic 07





#### Match two regions



#### Match two regions

• If their immediate properties are similar



#### Match two regions

- If their immediate properties are similar
- AND the same holds for their subregions

## Addressing Instability of Image Segmentation



#### Many-to-many matching = Augmenting trees with mergers

Todorovic & Ahuja 06, Ahuja & Todorovic 07

## Addressing Instability of Image Segmentation



#### Many-to-many matching = Augmenting trees with mergers

Todorovic & Ahuja 06, Ahuja & Todorovic 07

## Addressing Instability of Image Segmentation



# Matching all descendants under a node $\Downarrow$ $\Downarrow$ $\Rightarrow$ tree flatteningMatching transitive closures of trees

Torsello & Hancock 03, Pelillo et al. 99, Glantz et al. 04

# **OUR APPROACH**



- Represent images as segmentation trees
- Down-weight fine details closer to leaf nodes Find weighted transitive closure of the trees
- Match by separating the scales of the objects and scene --Normalization of region properties

## Weighted Transitive Closures



Weights  $\rho$  associated with all edges in the tree

$$\rho(v, u) = rac{\operatorname{area}(u)}{\operatorname{area}(v)}$$









## Separation of Scene Scale from Object Scale



## Separation of Scene Scale from Object Scale



#### **Example:**

 $\delta_{\text{area}} = \operatorname{area}(v)/\operatorname{area}(v') \rightarrow \widetilde{\operatorname{area}}(v') = \delta_{\text{area}} * \operatorname{area}(v')$ 

 $\psi(v), \ \psi(v') =$  vectors of region properties

$$\delta_i = \psi_i(v) \oslash \psi_i(v') \to \widetilde{\psi}_i(v') = \delta_i \otimes \psi_i(v')$$

 $\Rightarrow \Delta = \{\delta_1, \ldots, \delta_d\}$  normalization factors

## Normalization



Use  $\Delta$  to normalize all descendants u' of v' $\widetilde{\psi}_i(u') = \delta_i \otimes \psi_i(u')$   $i = 1, \dots, d$ 

## Normalization



Use  $\Delta$  to normalize all descendants u' of v' $\widetilde{\psi}_i(u') = \delta_i \otimes \psi_i(u')$   $i = 1, \dots, d$ 

↓

Properties of all nodes are normalized to those of root v $\downarrow$ Separation of the scale of the object from the scale of the scene

## Tree Matching: Formulation

Given two weighted trees:  $t = (V, E, \psi, \rho)$  and  $t' = (V', E', \widetilde{\psi'}, \rho')$ 

For each pair of nodes:  $(v, v') \in V imes V'$ 

Find bijection between the descendants of v and v'

$$f = \{(u, u')\} \subset V \times V'$$

which minimizes the cost of matching:

$$C_{vv'} = \min_{f} \left[ \sum_{(u,u') \in f} A_{vv'} + \sum_{(w,w',u,u') \in f \times f} B_{vv'uu'} \right]$$

where A and B are defined in terms of region properties and edge weights

## Tree Matching: Formulation

#### **Relaxation of the discrete problem**

$$C_{vv'} = \min_{X} \left[ A^T X + \frac{1}{2} X^T B X \right]$$

s.t.
$$x_{uu'} \in [0,1], \quad \sum_u x_{uu'} = 1 \quad \sum_{u'} x_{uu'} = 1$$





#### prior work

#### our results







Caltech-101: Faces





#### matching the down-sampled textures







• Scale-invariant object matching achieved by:



- Scale-invariant object matching achieved by:
- Down-weighting the effect of missing fine details at coarser scales



- Scale-invariant object matching achieved by:
- Down-weighting the effect of missing fine details at coarser scales
- Separating the scale of the object from the scale of the scene