

Recognizing Instances of Deforming Models

Robert B. Fisher
School of Informatics
University of Edinburgh

PDM Based Classification/Recognition

Given:

- Unknown sample \vec{x}
- Structural model: mean \vec{m} + M variation axes \vec{a}_j
- Statistical model: class means $\{\vec{t}_i\}$ and associated covariance matrices $\{C_i\}$ for $i = 1..K$ classes

For each class i :

1. Project \vec{x} onto \vec{a}_j to get weights \vec{w} (M dim vector)

2. Compute distances:

$$d_i(\vec{w}) = ((\vec{w} - \vec{t}_i)'(C_i)^{-1}(\vec{w} - \vec{t}_i))^{\frac{1}{2}}$$

Select class i with smallest distance $d_i(\vec{w})$

Reject if smallest distance is too large

Model Learning Summary

1. Load image, threshold, get boundary and find corners (c.f. TASK 1, except with a better corner finding threshold)
2. Point Distribution Model learning method:
 - (a) Rotate TEEs to standard position
 - (b) Get vertices into in a standard order
 - (c) Construct structural model using PCA
 - (d) Project examples into PCA weight space
 - (e) Estimate statistical model of projections

Recognition Summary

1. Load image, threshold, get boundary and find corners (c.f. TASK 1, except with a better corner finding threshold)
2. Point Distribution Model classification method:
 - (a) Constraint: reject if not 8 vertices
 - (b) Rotate TEE to standard position for PDM (Constraint: reject if not 2 sets of 4 nearly parallel lines)
 - (c) Get vertices into vector in a standard order
 - (d) Project vector into PCA (structural model)
 - (e) Evaluate statistical likelihood of projection (statistical model) using Mahalanobis distance

?

 Code

```
function train

% training phase
datacount = 0;
maximages = 31;
imagestem=input('Training image file stem\n?','s');
for imagenum = 1 : maximages
    currentimagergb = imread([imagestem, ...
        int2str(imagenum),'.jpg'],'jpg');
    currentimage = rgb2gray(currentimagergb);

% get corners using TASK2's method
Image = getbinary(currentimage,0,0,0,0,0); %threshold
[H,W] = size(Image);
```

```
[r,c] = find( bwperim(Image,4) == 1 );    %perimeter
[sr,sc] = removespurs(r,c,H,W,0);        %clean
[tr,tc] = boundarytrack(sr,sc,H,W,0);    %track
datalines = zeros(100,4);                % space for results
numlines = 0;
findcorners(tr,tc,H,W,9,16);             %segment

% process boundary, assuming it is a TEE
if numlines == 8
    % rotate datalines to standard position
    [newlines,flag] = standard_position( ...
        datalines(1:numlines,:),numlines,11);

% get vertices
if flag
    % sort vertices into a standard order
```

```
sortvertices=sortvert(newlines(1:numlines,1:2));

% add to scatter matrix
[Vnum,Vcoord] = size(sortvertices);
if Vnum == 8
    % turn points into long array
    datacount = datacount + 1;
    allvertices(datacount,:) = ...
        reshape(sortvertices,1,Vnum*Vcoord);
end
end
end
end

% Create model
meanvertex = mean(allvertices);
```

```
vertexdev = allvertices - ones(datacount,1)*meanvertex;
scatter = vertexdev'*vertexdev;
[U,D,V]= svd(scatter);
modeldev = V(:,1:5)' % use only first 5 components

% get projections onto data
vecs = vertexdev*modeldev';

% get class mean vector and covariance matrix
[Mean,Invcor] = buildmodel(vecs,maximages);

% save training data
save modelfile modeldev meanvertex Mean Invcor
```

? Distance

Vector distance measure that takes account of different range of values for different positions in vector and correlation between dimensions

Given vectors \vec{a} , \vec{b} from a set with covariance \mathbf{C} , the Euclidean Distance between the vectors is:

$$\|\vec{a} - \vec{b}\| = [(\vec{a} - \vec{b})'(\vec{a} - \vec{b})]^{\frac{1}{2}}$$

The Mahalanobis Distance is:

$$[(\vec{a} - \vec{b})'\mathbf{C}^{-1}(\vec{a} - \vec{b})]^{\frac{1}{2}}$$

IE, scaled and de-correlated differences

? Code

Same as Training code up to where 8×2 sorted point list reshaped into 16-vector

```
% turn 8 2D points into 16 vector
tmp = reshape(sortvertices,1,2*Vnum);

% project onto eigenvectors
vec = (tmp - meanvertex)*modeldev';

% get class distance
dist = mahalnobis(vec',Mean,Invcor);
```

What We Have Learned

1. Use statistical classifier once deformation projected into space of expected values