Recognizing Instances of Deforming Models

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PDM Based Classification/Recognition

Given:

- Unknown sample \vec{x}
- Structural model: mean $\vec{m} + M$ variation axes \vec{a}_i
- 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}_i to get weights \vec{w} (M dim vector)
- 2. Compute Mahalanobis 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

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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 vector 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 weight space (structural model)
 - (e) Evaluate statistical likelihood of projection (statistical model) using Mahalanobis distance

Training Code

```
[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));
```

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```
% 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
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
```

Mahalanobis 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:

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

IE, scaled and de-correlated differences

Recognition 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 = mahalanobis(vec',Mean,Invcor);
```

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What We Have Learned

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