

Fall 2016 Homework Set 5

Ans: 1

0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	1	1	1	0	0	0	0	0	0	0	0	0	0	0
0	0	1	1	1	1	1	0	0	0	0	0	1	1	1	0	0
0	0	1	1	1	1	1	X	0	0	0	1	1	1	1	0	0
0	0	1	1	1	0	0	0	0	1	1	1	1	1	1	1	0
0	0	0	1	1	0	0	0	0	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	0	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	0	0	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	0	0	0	1	1	1	1	1	0

The above is the binary matrix of dimension 16×16 ,
the following structure element is

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

The above is the 3×3 structuring element.

- (a) If we apply dilation in the above given 16×16 image
& this 3×3 structuring element, we get the following
image mask matrix.

(a)

0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	1	1	1	1	1	0	0	0	0	0	0	0	0	0	0
0	1	1	1	1	1	1	1	0	0	0	1	1	1	1	0	0
0	1	1	1	1	1	1	1	0	1	1	1	1	1	1	1	0
0	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	0
0	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	1	1	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	0	0	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	0	0	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	0	0	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	0	0	1	1	1	1	1	1	1	1	1	1	1	1	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

Dilation Operation

The above matrix is the result of the dilation operation on 16×16 image. $\star 8E$

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

(b)

Erosion Operation

The above matrix is a result of erosion operation on a 16×16 image given image with SE $\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$

(c)

0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	1	0	1	0	0	0	0	0	0	0	0	0	0	0
0	0	1	0	0	0	1	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	0	1	0	0	0	0	0	0	0	0
0	0	0	0	1	0	0	0	0	1	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1

Hit miss operation.

(c) In the above image matrix, 4 structuring elements are used

$$\begin{bmatrix} 1 & & \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 0 \\ 0 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$$

on a given 16×16 image.

Ans: (a) Active Shape Models are the statistical based shaped model which captures variability in non-uniformness of the shape from the train a set of images. It involves manual marking landmark points on the boundaries or the edge segments of the facial edges, hence it is based on edges and edge patterns. During alignment of the test image, ASM deals with various operation like rotation, translation ~~and~~, scaling, PCA etc and these landmark points eventually converges to get best fit on the test image.

- b) Active Shape Model is widely used for face detection due to its robustness and variability, since it is geometrically constraint, it is more suited for matching object boundaries rather than appearances such as emotions. Active Appearance Model can capture textures in order to capture emotions as well as location which may not be as accurate as ASM.
- c) for capturing emotions of the training set of images containing different emotions such as happy, happiness, anger, disgust, surprise and neutral., the following algorithm describes to construct mean face of a person,
- 1) Compute ASM of each shape sample as the linear combination of eigen vectors, $b_s = f_s^T(x - \bar{u})$
 - 2) Through linear & non-linear interpolation, wrap each image to the mean shape
 - 3) Normalize each image to average intensity and unit variance \bar{g} .
 - 4) Perform PCA on normalized images.
 - 5) Express each intensity sample as a linear combination of eigen vectors $b_g = f_g^T(g - \bar{g})$ represent sample gray-level parameters

6) Concatenate shape coefficient b_s and gray level coefficient b_g

$$\begin{bmatrix} b_s \\ b_g \end{bmatrix} = \begin{bmatrix} W P_s^T (n - \bar{x}) \\ P_g^T (g - \bar{g}) \end{bmatrix}$$

W is weighting diagonal matrix

7) Apply PCA to the sample set of all b vectors, yielding ~~it~~

$$b = Q c$$

where Q containing eigenvectors and c are model coefficients
signifies how model is different from mean shape and
mean appearance. If $c=0$, model characterized as mean
shape & appearance.

~~if~~ Emotions in the given problem can be taken as training set of images. If we take $c=0$ from the last step of the above algorithm, we can get mean shape & appearance of all the emotions of the face.

(d) Active Appearance model represents object shape and its appearance via low dimensional subspace. It is widely used for fitting model especially in video sequence. For detecting emotions on a video sequence, one can take emotions as a set of image sequence (given 1000) in which each image consists of different class of emotion. For every given every class of emotion (anger, happiness, disgust, surprise & neutral) has equal number of sequences, we can calculate parameter c i.e. mean shape & mean appearance by setting it to 0. for each set of emotion. Hence, we can train the model on the given sets of image emotions & test it on different images for different emotions.