## Pattern Recognition and Machine Learning : Assignment 4

- The assignment is **due** on **April 4**.
- Submit a soft copy of the code and report highlighting the observations and inferences before the deadline.
- Let me inform you that this assignment is very challenging and exciting. So start early! Do ask for any clarifications, if necessary.

## Task 1

Implement a Gaussian Mixture Model (GMM) based clustering scheme on the image 'ski\_image.jpg'. To aid you through this task, you may consider incorporating the following suggestions:

- Assume that the mixture comprises 3 Gaussian components (clusters).
- The RGB values of pixel intensities (after appropriate normalization to the range [0,1]) can be used as features.
- You may consider starting the iterations using the means

$$\mu_1 = \begin{bmatrix} 0.47 \\ 0.47 \\ 0.47 \end{bmatrix}$$
 $\mu_2 = \begin{bmatrix} 0.05 \\ 0.05 \\ 0.05 \end{bmatrix}$ 
 $\mu_3 = \begin{bmatrix} 0.7 \\ 0.7 \\ 0.7 \end{bmatrix}.$ 

• The covariance matrix corresponding to each Gaussian may be initialized as

$$\Sigma_{i} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \qquad i = 1, 2, 3$$

• The weighting of each Gaussian component (prior weights) in the GMM may be initialized to  $\frac{1}{3}$ .

$$\pi_i = \frac{1}{3}$$
  $i = 1, 2, 3$ 

• It is very important that you iterate through the algorithm several times , so that the likelihood function converges.

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- 1. Display the segmented output.
- 2. Display a graph depicting the convergence of the log likelihood values.
- 3. What are the final values of the means, prior weights and covariance matrices

## Task 2

In this task, you will apply the Gaussian Mixture Model as a density estimation technique to the problem of classification. The description of the data to be used for training and testing is as follows:

- You have been provided training features, corresponding to two classes  $\omega_1$  and  $\omega_2$  in the files Pattern1.mat and Pattern2.mat respectively. Each file contains 200 instances (training examples), of 120 feature dimensions.
- The features corresponding to 100 testing samples of  $\omega_1$  and  $\omega_2$  are contained in Test1.mat and Test2.mat respectively.
- 1. Utilize the features contained in Pattern1.mat and Pattern2.mat to build separate Gaussian Mixture Models for the classes  $\omega_1$  and  $\omega_2$ . You may choose 2 Gaussian components in each mixture. The initial means may be randomly chosen from the training data by running the k-means algorithm. The covariance matrix for each Gaussian component can be initialized to the identity matrix.
- 2. Use the trained models from part (1) to test the performance of the features contained in the files Test1.mat and Test2.mat. Report the recognition accuracy for each class.