

Skin Detection based on Classification Algorithms: Exercise Set 6

Remark: Please note that the script load_data (providing the required files) and the auxiliary functions s = distance(p, q) and classify_nearestmean(A, M, k) (described below) are available on olat.

Task 1 Consider the function s = distance(p, q), which computes the euclidean distance between two (not necessarily 2-dimensional) points p and q.

Task 2 The goal is to evaluate various strategies for classifying a given point.

- (a) Have a look at the function $[\alpha] = \text{classify_nearestmean}(A, M, k)$, which takes as input:
 - \bullet an integer k denoting the number of classes,
 - \bullet a matrix M each row of which represents a not yet classified point, and
 - a matrix A whose rows can be partitioned into k equally sized sets of points belonging to the same class: With N denoting the number of elements per class, the points of class i are contained in the rows $N \cdot (i-1) + 1 \dots N \cdot i$, for each $i \in \{1, \dots, k\}$ (see the illustration to the right),

N	
N	
N	

and returns a vector α with $\alpha(i)$ denoting the class assigned to the point of the *i*th row of M according to the nearest mean strategy.

(b) $[\alpha] = \text{classify_gmm}(A, M, k)$ which is obtained from modifying the function from part (a) in such a way that α is determined using Gaussian mixture models.

Test Cases: For the input (A1, M1, 3) you should get [3;3;2;2]. Furthermore, also check whether in your internal computation, you get

means = $[19.7997 \ 60.5564]$, $[81.0068 \ 60.4333]$, $[39.5510 \ 19.7952]$, and

$$\text{covariance matrices} = \begin{pmatrix} 29.1874 & 0.9561 \\ 0.9561 & 27.9604 \end{pmatrix}, \begin{pmatrix} 33.2761 & -1.6249 \\ -1.6249 & 26.8107 \end{pmatrix}, \begin{pmatrix} 30.5662 & -0.2978 \\ -0.2978 & 25.2951 \end{pmatrix}.$$

For the input (A2, M2, 2) you should get [1;2;2;2;2].

Goal: In the subsequent tasks you will analyze various algorithms for skin detection. The below illustration shows a 'perfect' skin detection.

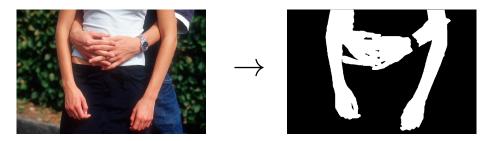


image credit: Prof. Thomas Vetter, University of Basel

Convention: Throughout this exercise sheet, intensity values of pixels are in the range of [0, 1].

Task 3 The goal is to evaluate the algorithms from Task 2 for the datasets skindata and nonskindata and the image test.¹

- (a) First, implement the following auxiliary functions:
 - M = rgbImage2Matrix(I) which takes as input an rgb image I and constructs a matrix by concatenating the columns of I.

(Example: The image
$$I^2 = \begin{pmatrix} [0.1 \ 0.2 \ 0.3] & [0.4 \ 0.6 \ 0.8] \\ [0.8 \ 0.2 \ 0.5] & [0.6 \ 0.3 \ 0.9] \end{pmatrix}$$
 becomes $M = \begin{pmatrix} 0.1 \ 0.2 \ 0.3 \\ 0.8 \ 0.2 \ 0.5 \\ 0.4 \ 0.6 \ 0.8 \\ 0.6 \ 0.3 \ 0.9 \end{pmatrix}$)

- I = Vector2grayImage(S, r, s) which takes as input a vector S and constructs a grayscale image I of size $r \times s$ by copying the elements of S column by column. (Example: S = [0.1; 0.5; 0.4; 0.6], r = 2 and s = 2 gives $\begin{pmatrix} 0.1 & 0.4 \\ 0.5 & 0.6 \end{pmatrix}$.)
- (b) Write a script that applies each strategy of Task 2 to the image test (using the previously developed auxiliary functions). Represent the output as a binary image (1 = 'skin', 0 = 'non-skin')
 - Evaluate the results visually (i.e. by looking at the result).
 - For each result, determine the corresponding error rate (the 'true classification' can be found in maskTest). ³
- (c) **Optional:** Apply your algorithm based on the Gaussian mixed model to some other images (e.g. from Google) and determine visually whether the result is good. **Remark:** This part does not have to be handed in.

¹These files are from Prof. Thomas Vetter, University of Basel.

²provided by load_data

³We will refer to this value as test error.

Task 4 The goal is to determine a skin/non-skin classification using Support Vector Machines. Write a script (using appropriate Matlab commands) that first trains a classifier for skindata and nonskindata and then applies the result to test.png.

Determine for the error rate, the false positive rate and the false negative rate and compare them with the respective values for the Gaussian mixture model.

Remark: It is recommended to tailor your script to the given situation (i.e., 2 groups of data, d = 3 dimensions and N = 10'000 samples for each training set). No generalizations are required!