ICT for Health
Laboratory # 4
Moles

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Table of Contents

Chronic kidney disease [1]

Use the agglomerative clustering of SciPy on the data of Chronic kidney disease and comment the results

Table of Contents

The data [1]

- Download file images.zip from the folder materiale of the class
- Unzip the file, you'll get several jpeg images of moles, the name of the files are
  1. low_risk_n.jpg (where n is an integer) for moles that have a low probability of being melanoma (i.e. tumors)
  2. medium_risk_n.jpg (where n is an integer) for moles that have a low probability of being melanoma
  3. melanoma_n.jpg (where n is an integer) for moles that have a high probability of being melanoma
- Use a tool of your operating system to view all the pictures, so that you have an idea of what you have to work with (for example use Gwenview in Linux)

Goal of the lab

- We want to help medical doctors in the analysis of the moles
- 5 features are considered by the doctor to diagnose melanoma: ABCDE

    A asymmetry
    B border
    C color
    D diameter
    E evolution

- We want to analyze borders

Main idea

- We use K-means in scikit-learn to find three clusters (quantization of the image with three levels)
- We find the contour of the cluster corresponding to the mole
- We evaluate the area of the cluster corresponding to the mole and the length of the contour (perimeter of the mole)
- We evaluate the perimeter of a perfect circle with area equal to that of the mole
- We evaluate the ratio between the perimeter of the mole and the perimeter of the corresponding circle: the higher is this value the more indented is the border.

The jpeg image

- The jpeg image is an image that has been compressed according to the jpeg standard
- To read the image in Python:
  ```
  import matplotlib.image as mpimg
  filein=...
  im = mpimg.imread(filein)
  ```
- `im` is an Ndarray with shape 583 x 583 x 3 and elements of type unint8 (unsigned integer with 8 bits); the image is made of 583 x 583 pixels
- `im[:,:,0]` is the amount of red color, from 0 to 255
- `im[:,:,1]` is the amount of green color, from 0 to 255
- `im[:,:,2]` is the amount of blue color, from 0 to 255
- Value `[0,0,0]` corresponds to black, value `[255,255,255]` corresponds to white
- To show the image in Python: `plt.figure()`
  ```
  plt.imshow(im)
  plt.title('original image')
  plt.show()
  ```

K-means [1]

- Import K-means by writing:
  ```
  from sklearn.cluster import KMeans
  ```
- To instantiate the K-means object write:
  ```
  kmeans = KMeans(n_clusters=3, random_state=0)
  ```
- To find the clusters you should write:
  ```
  kmeans.fit(im)
  ```
  but Python gives you an error, because it requires a 2D Ndarray, not a 3D Ndarray
- What can we do? the k-means algorithm does not take into consideration the order of the data, therefore we can reshape the 3D Ndarray into a 2D Ndarray:
  ```
  [N1,N2,N3]=im.shape
  im_2D=im_or.reshape((N1*N2,N3))# N1*N2 rows and N3
  columns
  ```

K-means [2]

- Then, use k-means as:
  kmeans.fit(im_2D)
- Note that class KMeans takes time to find the clustering: it actually tests some hundreds of different initial vectors and gives you the best clustering that minimizes the moment of inertia
- The attributes of class KMeans are:
  - kmeans.cluster_centers_, the centroids of the clusters
  - kmeans.labels_, the N1*N2 classes/clusters each pixel belongs to

  Note that the centroids are 3 NDarrays of float numbers, but we need uint8 numbers to show the image; therefore the centroids become:
  centroids=kmeans.cluster_centers_.astype('uint8')
  These three centroids represent the three colors that k-means found as representatives of all the image colors (the original image has potentially $2^{24} \simeq 16 \times 10^6$ different colors, but we want only 3 different colors)
- **Generate the image with only 3 colors, plot it and look at the difference between the original and the quantized image.**

The details

The contour [1]

- The classical algorithm to find the contour is the "snake" algorithm (or active contour), available both in Matlab and Python. You are NOT allowed to use the snake algorithm in this lab (otherwise you have nothing to do....)
- The idea is the following:
    1. Among the centroids, find the darkest color, it corresponds to the mole
    2. Find the median (not the mean) of the region where the quantized image is equal to this darkest color: this median is probably the center of the mole or it is close to be the center of the mole; note that other (small?) portions of the image might have the same darkest color
    3. Implement an algorithm that, starting from the center of the mole, finds a square or rectangular region that includes all the mole, but not the other areas (on the borders of the original image, typically) with the same color
    4. Consider this sub-image in the following
    5. Consider each column of the sub-image, find the index of the first pixel having the darkest color and the index of the last pixel having the darkest color: these two pixels are part of the contour

The contour [2]

6. Repeat the above point, considering now the rows
7. Plot the contour you found. Matplotlib class `matshow` might be better than `imshow`; note that the input of `matshow` must be an NDarray of dimension 2, not an image; you can specify the color map giving `matshow` the optional parameter `cmap='Blues'` (or other mappings); you can add the colorbar by writing `plt.colorbar()` right after `plt.matshow()`. See the help of `matshow`
8. **On your own: find a way to improve the quality of the contour; you are not allowed to discuss this specific topic with the other students, I expect that each student will find a different solution**
9. Once you have your contour find the ratio between the perimeter of the mole and the perimeter of the circle with the same area, analyze all the images and write a table with these ratios; include this table in the report.