Correction: Histogram-based segmentation

1 Matlab correction

1.1 Manual thresholding

Choose manually a value, by observing the histogram.

```
A=imread('cells.bmp');
B=(A>80);
sigure;
subplot(1,2,1);imshow(A);title('Original image');
subplot(1,2,2);imshow(B);title('Manual threshold');
```

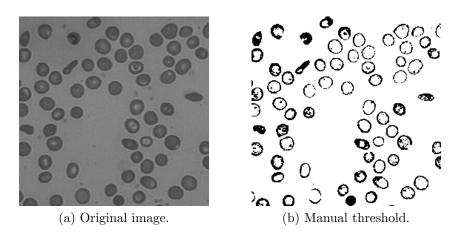


Figure 1: Simple thresholding.

1.2 Grayscale image, k = 2 in 1D

The automatic threshold selection proposed can be coded as follows:

```
function [s,B]=autothresh(A)
% Automatic threshold of image A
% return values:
% s: threshold value
5 % B: thresholded (binary) image

7 % initialization of s
s=0.5*(min(A(:)) + max(A(:)));
9 done = false;

11 % iterate until convergence of s
while ~done
13 B=(A>=s);
sNext=0.5*(mean(A(B))+mean(A(~B)));
```

```
done=abs(s-sNext) <0.5; % convergence ?
s=sNext;
r end
```

Then, display the different results with:

```
A=imread('cells.bmp');
% threshold determination
3 [s1,B]=autothresh(A);
s1

5
% Otsu (matlab function)
7 s2=graythresh(A);
s2=255*s2
9 C=(A>=s2);

11 % display results
figure;
13 subplot(2,2,1); imshow(A); title('Original image');
subplot(2,2,3); imshow(B); title('Automatic threshold');
15 subplot(2,2,4); imshow(C); title('Otsu threshold');
```

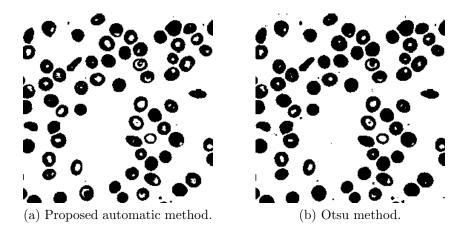


Figure 2

1.3 Simulation example, k = 3 in 2D

The first function generates random points around a center. The objective is to retrieve the different clusters (aka classes). Depending on the distance and the distribution of the points, this could not yield to the expected result.

```
function Y=generation(n, x, y)
2 % Generates n random points (normal law) around point
% of coordinates (x,y)
4 Y = randn(n,2)+ones(n,2)*[x 0; 0 y];
```

```
% Generate 3 point clouds
2 n=100;
X=[generation(n,3,4); generation(n,0,0); generation(n,-5,-3)];

% Classification
6 [idx, ctrs] = kmeans(X, 3, 'replicates', 5);

8 % Display results
figure();
10 plot(X(idx==1,1),X(idx==1,2),'r.','MarkerSize',12)
hold on
12 plot(X(idx==2,1),X(idx==2,2),'b+','MarkerSize',12)
plot(X(idx==3,1),X(idx==3,2),'g*','MarkerSize',12)

14 legend('Cluster 1','Cluster 2','Cluster 3')
```

1.4 Color image segmentation by K-means: k = 3 in 3D

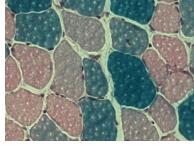
Each pixel of the color image is represented as a vector (3D point, with the RGB color values of the pixels). Then, the same method is performed. The result is represented in Fig.3.

```
1 % Load image
    I=imread('Tv16.png');
3    I=double(I);
    figure
5    imshow(I/255);

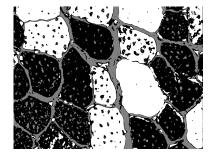
7 % Segmentation
    nCouleurs = 3; % number of clusters
9    nLignes = size(I,1);
    nCols = size(I,2);

11    X = reshape(I, nLignes*nCols, 3);
```

```
[index centres] = kmeans(X, nCouleurs, 'distance', 'sqEuclidean', '
     \hookrightarrow replicates', 3);
 \% 3D histogram (can be difficult to display, depending on machine)
id1=index==1;
  id2=index==2;
id3=index==3;
21 figure
  plot3(X(id1,1), X(id1,2), X(id1,3), 'r.')
23 hold on
  plot3 (X(id2,1), X(id2,2), X(id2,3), 'g.')
25 plot3 (X(id3,1), X(id3,2), X(id3,3), 'b.')
27 % Label each pixel
  labels = uint8(reshape(index, nLignes, nCols));
29 labels = imadjust(labels);
31 figure
  subplot (121)
33 imshow(I/255);
  subplot (122)
35 imshow(labels, []);
```



(a) Original image.



(b) Segmented image.

Figure 3: K-means segmentation applied on a color image. The segmentation is applied in the color space. The spatial informations of the image structures is lost.