Report for

Digital Image Processing

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version 1.1

$$(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

$$\zeta_k = |a|^{1/n} e^{i(\arg(a) + 2k\pi)/n}$$

$$e^{i\pi} + 1 = 0$$

$$\gamma = (-p) \wedge (-q)$$

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

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3

- (a) Write a computer program for computing the histogram of an image.
- (b) Implement the histogram equalization technique.
- (c) Your program must be general to allow any gray-level image as its input. As a minimum, your report should include the original image, a plot of its histogram, a plot of the transformation function, the enhanced image, and a plot of its histogram.

1.2 Generate the Histogram

1.2.1 Function

Generating the histogram of an image using following function:

H(i) = the number of pixel whose value enquals to i

1.2.2 Histogram

The histogram pictures of Fig1.jpg and Fig2.jpg are listed as follows:

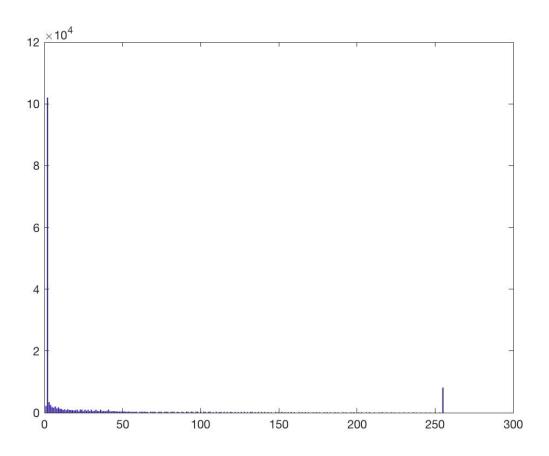


Figure 1.1: Histogram of fig1.jpg

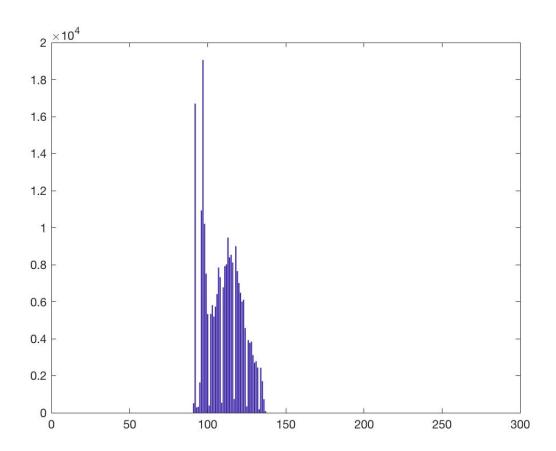


Figure 1.2: Histogram of fig2.jpg

1.3 Transfer Function

1.3.1 Implement the Histogram Equalization Technique

We use those functions to calculate the histogram equalization:

$$L = Max(image(r,c)) \ \forall r \in [1,rows] \ and \ \forall c \in [1,cols] \\ s(r_k) = L*T(r_k) = L*\sum_{j=0}^k P_r(r_j) = L*\sum_{j=0}^k \frac{n_j}{n}$$

1.3.2 Transfer Function

The transfer function of Fig1.jpg and Fig2.jpg are listed as follows:

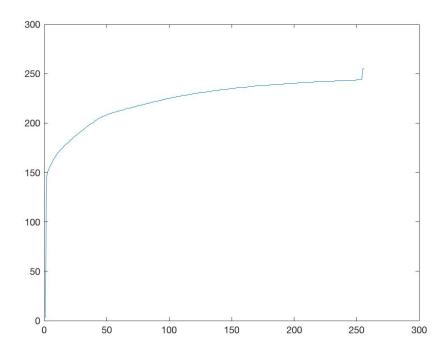


Figure 1.3: Transfer Function of fig1.jpg

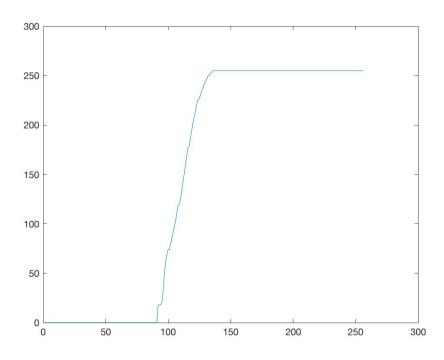


Figure 1.4: Transfer Function of fig2.jpg

1.4 Enhanced Images

1.4.1 Enhanced Function

We use the function

New $Image(r,c) = Transfer\ Function(image(r,c))\ \forall r \in [1,rows]\ and\ \forall c \in [1,cols]$ to enhance the original images.

1.4.2 Enhanced Images

Original Image

The original images and enhanced images and histogram comparation are listed as follows.

Image(Histogram Equalization)

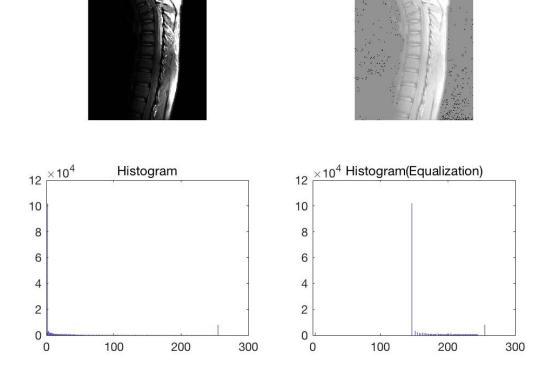
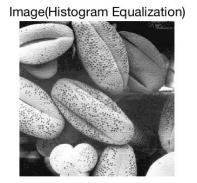
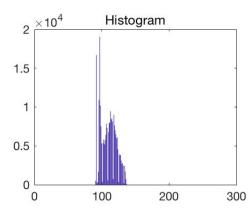


Figure 1.5: original image and enhanced image and histogram comparation of fig1.jpg







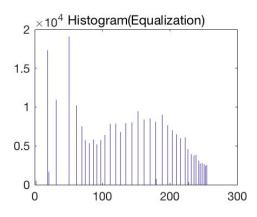


Figure 1.6: original image and enhanced image and histogram comparation of fig2.jpg

Implement the image enhancement task of Section 3.7 (Fig 3.43, page 171). The image to be enhanced is skeleton_orig.tif. You should implement all steps in Figure 3.43. (You are encouraged to implement all functions by yourself, not to directly use Matlab functions such as imfilter or fspecial.)

2.2 Image b

2.2.1 Laplacian Transform Filter

We apply Laplacian Transform on the original image to get the image (b) using the following filter.

$$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$$

2.2.2 Laplacian Transform

Image b:

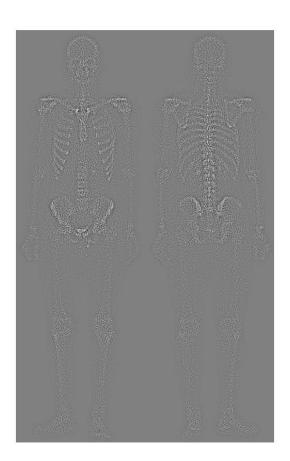


Figure 2.1: Laplacian Transform of skeleton_orig.tif

2.3 Image c

The we add the Laplacian of the original image to the original image, we will get the new image c. The new image c is a rather noisy sharpened image. Image c:

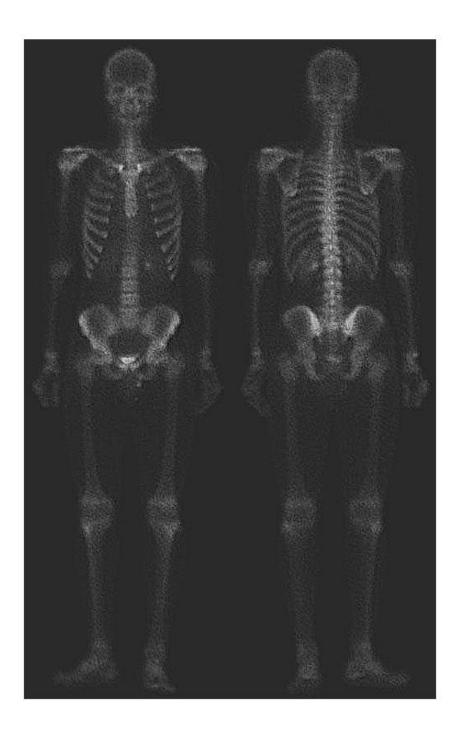


Figure 2.2: Laplacian Transform of skeleton_orig.tif

2.4 Image d

2.4.1 Sobel Gradient Masks

We will use two mask to separately get the components g_x and g_y . Then add the two components together, we will get the the sober gradient of the original image. The new image is as follows. As we can see, edges are much more dominant in this image than in the Laplacian image.

 g_x :

$$\begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

 g_y :

$$\begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

2.4.2 Image d

Image d:



Figure 2.3: Laplacian Transform of skeleton_orig.tif

2.5 Image e

Image e is formed by smoothing image d by 5*5 mean filter. Image e:

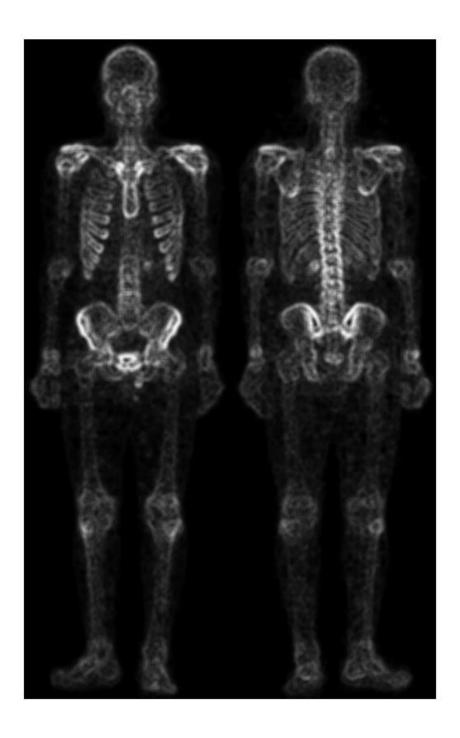


Figure 2.4: Laplacian Transform of skeleton_orig.tif

2.6 Image f

Image f is formed by the product of Laplacian and smoothed-gradient image. The dominance of the strong edges and the relative lack of visible noise, which is the key objective behind masking the Laplacian with a smoothed gradient image. Image f:



Figure 2.5: Laplacian Transform of skeleton_orig.tif

2.7 Image g

Adding the image f to the original image and then we get image g. image g:

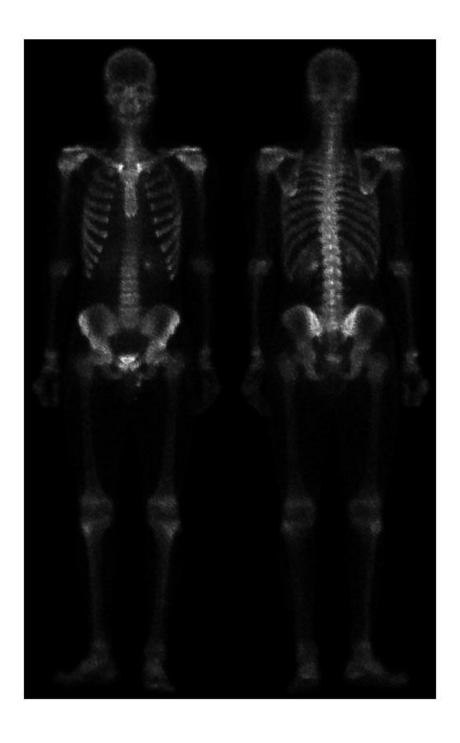


Figure 2.6: Laplacian Transform of skeleton_orig.tif

2.8 Image h

2.8.1 Power-Law Transformation

We use the following function to perform Power-Law Transformation on image g.

$$s = cr^{\gamma} \ (c = 1 \ and \ \gamma = 1)$$

2.8.2 Image h

Image h:



Figure 2.7: Laplacian Transform of skeleton_orig.tif

Implement the ideal, Butterworth and Gaussian lowpass and highpass filters and compare the results under different parameters using the image characters_test_pattern.tif (this image file can be found at the ftp server ftp://ftp.cs.sjtu.edu.cn:990/lu-ht/DIP/images) as the test pattern.

- 3.2 This is a section
- 3.2.1 This is a subsection
- 3.2.2 This is a subsection

In this problem, you are required to write a program to generate different types of random noise (Uniform, Gaussian, Rayleigh, Gamma, Exponential and Impulse, first started from the uniform noise and then use some functions to convert the uniform noise to Gaussian, Rayleigh, Gamma and Exponential; Impulse noise is generated in a different way, consulting the textbook and some other references) and then add these noises to the test patter image Fig0503(original_pattern).tif to compare the visual results of the noisy images.

Add some of these noises to the circuit image Circuit.tif (images can be found at ftp://ftp.cs.sjtu.edu.cn:990/lu-ht/DIP/images) and investigate the noise reduction results using different mean filters and order statistics filters as the textbook did at pages 344-352 (Pages 322-329 in the electronic version of the textbook).

- 4.2 This is a section
- 4.2.1 This is a subsection
- 4.2.2 This is a subsection

- 5.1 OverView
- 5.2 This is a section
- 5.2.1 This is a subsection
- 5.2.2 This is a subsection



A.1 Problem1

```
% Problem 1
  % by Xue Fanyong
2
  % Student ID:515030910443
  % Histogram Equalizatio
5
  1998 Main Part
6
  image1 = imread(' Image Path/Fig1.jpg' );
7
  image2 = imread( Image Path/Fig2.jpg' );
8
   [histogram1, histogram el, transfer fl, image el] =
10
   histogram equalization (image1);
11
   [histogram2, histogram_e2, transfer_f2, image_e2] =
12
   histogram_equalization(image2);
13
14
   plot_data(image1,image_e1,histogram1,histogram_e1,transfer_f1);
15
   plot_data(image2,image_e2,histogram2,histogram_e2,transfer_f2);
16
17
  %% Functions Part
18
19
  % get histogram of image
20
  % image: get histogram of it
21
  % histogram: the histogram of image
22
  function histogram = get_histogram (image)
23
       histogram = zeros(256,1);
24
       [row, col]=size(image);
25
       for r = 1:row
26
           for c = 1:col
27
                gray = image(r,c);
28
                histogram (gray+1)=histogram (gray+1)+1;
29
           end
30
       end
31
  end
32
33
  % do the histogram_equalization for image
  % image: do histogram_equalization for it
35
  % histogram: original histogram; histogram_e:
36
  % histogram after histogram
37
  % equalizatio; transfer_f: transfer function;
38
  % image_e: image after histogram
  % equalizatio
40
   function [histogram, histogram_e, transfer_f, image_e] =
41
   histogram_equalization (image)
42
       [row, col]=size(image);
43
       transfer_f = zeros(256,1);
44
       histogram = get_histogram(image);
45
       transfer_f(1) = 256*histogram(1)/(row*col);
46
47
       for i = 2:256
48
           transfer_f(i) = transfer_f(i-1)+255*histogram(i)/(row*col);
49
```

```
end
50
       transfer f = round(transfer f);
51
52
       image_e = image;
53
       for r = 1:row
54
            for c = 1:col
55
                image_e(r,c)=transfer_f(image(r,c)+1);
56
            end
57
       end
58
       histogram_e = get_histogram(image_e);
59
   end
60
61
  % plot data
62
  % image: original image; image_e:
63
  % image after histogram equalizatio;
  % histogram: original histogram;
65
  % histogram_e: histogram after histogram equalizatio;
66
  % transfer_f: transfer function
67
  function plot_data(image,image_e,histogram,histogram_e,transfer_f)
68
       figure();
69
       subplot (2,3,1);
70
       imshow(image);
71
       title(" Original Image");
72
       subplot (2, 3, 2);
73
       imshow(image e);
74
       title("Image(Histogram Equalization)");
75
       subplot (2,3,3);
76
       bar(histogram);
77
       title (" Histogram" );
78
       subplot(2,3,4);
79
       bar(histogram e);
80
       title(" Histogram(Equalization)");
81
       subplot(2,3,5);
82
       plot(transfer_f);
83
       title("Transfer Funciton");
84
  end
85
```

A.2 Problem 2

```
% Problem 2
% by Xue Fanyong
% Student ID:515030910443
% Combining spatial enhancement methods

% Main Part
image = imread(' Image Path/skeleton_orig.tif');
[row,col] = size(image);
mask = [-1 -1 -1;-1 8 -1;-1 -1 -1];
mask = double(mask);
```

20

```
b_image = laplace_transformations(image, mask);
11
  c image = b image+im2double(image);
12
  d_image = sobel_gradient(image);
13
  e_image = smooth(d_image);
  f_image = im2double(e_image).*c_image;
15
   g_image = abs(f_image)+im2double(image);
16
   h_image = sqrt(g_image);
17
   plot_data(image, b_image, c_image, d_image,
18
              e_image, f_image, g_image, h_image);
19
20
  %% Function Part
21
22
  % Laplace Transfromation for image using mask
23
  % Input:
24
       image:image you want to perform
  %
  %
       mask: Laplace mask you want to use
26
  % Output:
27
  %
       image_I:image after laplace transformation
28
29
   function image_I = laplace_transformations(image, mask)
30
       [row, col] = size(image);
31
       mask = double(mask);
32
       %append image
33
       image I = im2double(image);
34
       image = [zeros(row,2) image zeros(row,2)];
35
       image = [zeros(2,col+4);image;zeros(2,col+4)];
36
       image append = im2double(image);
37
38
       for r = 1:row
39
            for c = 1:col
40
                image_l(r,c) = sum(sum(image_append(r:r+2,c:c+2).*mask));
41
            end
42
       end
43
   end
44
   %{
45
       sobel gradient for image
46
   %}
47
   function image_s = sobel_gradient(image)
48
       [row, col] = size(image);
49
       x \text{ mask} = [-1 \ -2 \ -1; 0 \ 0 \ 0; 1 \ 2 \ 1];
50
       y_mask = [-1 \ 0 \ 1; -2 \ 0 \ 2; -1 \ 0 \ 1];
51
       image s = image;
52
       image = double(image);
53
54
       for r = 2:row-1
55
            for c = 2:col-1
56
                image_s(r,c) =
57
                abs(sum(sum(image(r-1:r+1,c-1:c+1).*x_mask)))+
58
                abs(sum(sum(image(r-1:r+1,c-1:c+1).*y_mask)));
59
            end
60
       end
```

21

```
end
62
   %{
63
        smooth image using 5*5 mean filter
64
   %}
65
   function image_s = smooth(image)
66
        [row, col] = size(image);
67
        image_s = image;
68
        for r = 3:row-2
69
             for c = 3:col-2
70
                 image_s(r,c) = mean(mean(image(r-2:r+2,c-2:c+2)));
71
             end
72
        end
73
   end
74
   %{
75
        plot data
76
   %}
77
    function plot_data(a,b,c,d,e,f,g,h)
78
        figure();
79
80
        subplot (241);
81
        imshow(a);
82
        title (' (a) Oringinal Image');
83
84
        subplot (242);
85
        imshow(b,[]);
86
        title(' (b) Laplacian of (a)');
87
88
        subplot (245);
89
        imshow(c,[]);
90
        title ('(c) Sharpened image');
91
92
        subplot (246);
93
        imshow(d);
        title (' (d) Sobel gradient');
95
96
        subplot (243);
97
        imshow(e);
98
        title (' (e) Smoothed sobel image');
99
100
        subplot (244);
101
        imshow(f,[]);
102
        title ('(f) Product of (c) and (e)');
103
104
        subplot (247);
105
        imshow(g,[]);
106
        title ('(g) Sharpened image');
107
108
        subplot (248);
109
        imshow(h,[]);
110
        title ('(h) Final result');
111
   end
112
```