

数字图像处理 Homework1

1552746 崔鹤洁

Date: 2017-

10-10

Problem1

(1)PPT43

a. 求期望

当 $X \sim N(\mu, \sigma^2)$, 换元, 令 $t = \frac{x-\mu}{\sigma}$

$$\begin{aligned} E(X) &= \int_{-\infty}^{+\infty} x \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\} dx \\ &= \int_{-\infty}^{+\infty} (\sigma t + \mu) \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{t^2}{2}\right\} dt \\ &= \sigma \int_{-\infty}^{+\infty} \frac{t}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt + \mu \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \end{aligned}$$

分析:

由于 $\frac{t}{\sqrt{2\pi}} e^{-\frac{t^2}{2}}$ 是奇函数, 在对称区间上积分值为0;

同时, $\frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}}$ 为标准正态分布, 即 $\mu = 0, \sigma = 1$ 时的正态分布概率函数的结果, 所以其在 $-\infty$ 到 $+\infty$ 上积分值为1。

故上式最终结果为 μ , 即 $E(X) = \mu$

b. 求方差

由数学期望的性质, 可以得到

$$\begin{aligned} D(x) &= E[x - E(x)]^2 \\ &= E\{X^2 - 2XE(X) + [E(x)]^2\} \\ &= E(X^2) - 2[E(X)]^2 + [E(x)]^2 \end{aligned}$$

$$= E(X^2) - (E(X))^2$$

由 (A) 内的推导过程, 可以知道, 当 $X \sim N(\mu, \sigma^2)$ 时, $E(X) = \mu$

又因为

$$\begin{aligned} E(X^2) &= \int_{-\infty}^{+\infty} x^2 \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{(x-\mu)^2}{2\sigma^2}\right\} dx \\ &= \int_{-\infty}^{+\infty} (\sigma t + \mu)^2 \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{t^2}{2}\right\} dt \\ &= \sigma^2 \int_{-\infty}^{+\infty} \frac{t^2}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt + \mu^2 \\ &= \sigma^2 + \mu^2 \end{aligned}$$

因此, 正态分布的方差为

$$D(X) = (\sigma^2 + \mu^2) - \mu^2 = \sigma^2$$

(2) PPT120

❖result image:

Analysis:

The algorithm of function **adaptiveMedianFilter** is:

- Get the padding-width according to the cornel window size
- Get the size of the noise picture
- Fill pixels at the edge of the original image, which is assured by padding-width
- Get the temp window area Sxy, and calculate Zemd, Zmin, Zmax of this area
- For each pixel in the original figure, do the following operations:

Stage A:

A1 = Zmed - Zmin

A2 = Zmed - Zmax

If A1 > 0 and A2 < 0

Go to stage B

Else

increase the window size

If window size ≤ Smax

repeat stage A

Else output Zmed

Stage B:

$B1 = Z_{xy} - Z_{min}$

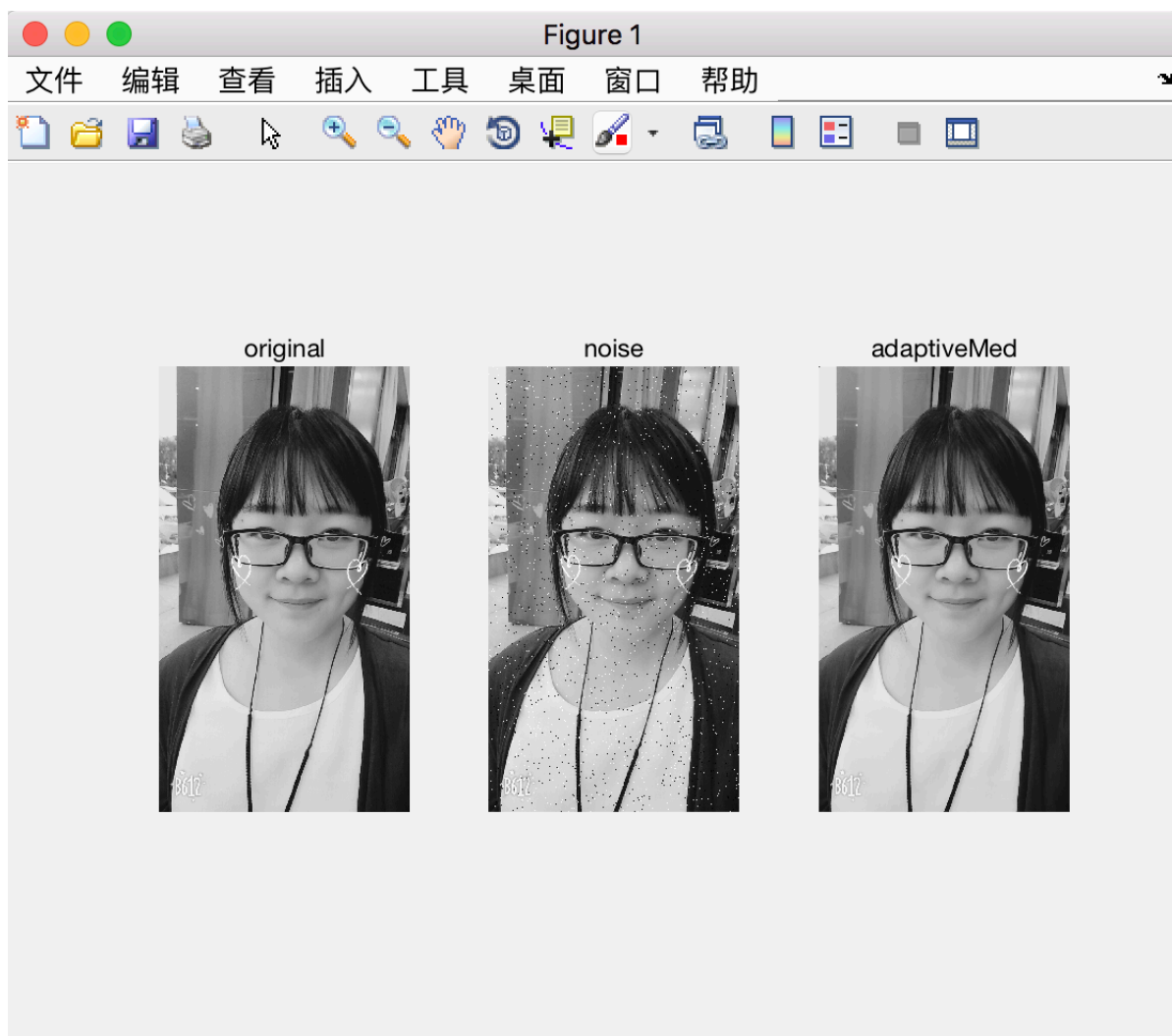
$B2 = Z_{xy} - Z_{max}$

If $B1 > 0$ and $B2 < 0$

output Z_{xy}

Else output Zmed

- If noise pixels in the window were more than half of the useful pixels, then it means you need to increase the size of the filter window S_{xy} , the increasing method is to make the window size + 2. For example, if the original filter window size is $3 * 3$, so after increasing the value of the window, the filter window size changed into $5 * 5$, if $S_{xy} < S_{max}$, continue this operation, otherwise the gray levels of the pixel is Zmed. By analogy, ***the adaptive control of filter window size is formed.***
- ***The maximum size of filter window S_{max}*** should adjust with the spacial density of noise. Generally speaking, if the spacial density of the noise is big, we should choose a bigger S_{max} ; if spacial density of the noise is small, S_{max} can be set a little smaller. Here we choose a fixed value, 11.



code:

```

clear
img = imread('cui.jpg');
img_gray = rgb2gray(img);
img_noi = imnoise(img_gray,'salt & pepper',0.02);
img_adaptive = adaptiveMedianFilter(img_noi,11);

figure;
% original image
subplot(1,3,1);
imshow(img_gray);
title('original')

% noised image
subplot(1,3,2);
imshow(img_noi);
title('noise')

% adaptive median filter image
subplot(1,3,3);
imshow(img_adaptive);
title('adaptiveMed')

function[img_result] = adaptiveMedianFilter(img_noi,max_window)

padding_width = (max_window-1)/2;
[m ,n] = size(img_noi);
% Initialize a image equal size with img_noi to receive pixels after
Adaptive Median Filter
img_result = img_noi;
% Fill pixels at the edge of the original image
img_noi = padarray(img_noi,[padding_width padding_width],0);

for i = padding_width+1:padding_width+m
    for j = padding_width+1:padding_width+n
        Zxy = img_noi(i,j);
        window = 3;
        while(window <= max_window)
            %Take the window of the filter plate
            tmp = img_noi(i-(window-1)/2:i+(window-1)/2,...
                j-(window-1)/2:j+(window-1)/2);
            Zmin = min(min(tmp));
            Zmax = max(max(tmp));
            Zmed = median(tmp(:));
            if((Zmed > Zmin) && (Zmed < Zmax))
                if((Zxy > Zmin) && (Zxy < Zmax))
                    img_result(i-padding_width,j-padding_width) = Zxy;
                else
                    img_result(i-padding_width,j-padding_width) = Zmed;
                end
            end
            window = window + 2;
        end
    end
end

```

```

        break;
    end
    window = window+2;
end
img_result(i-padding_width,j-padding_width) = Zmed;
end
end
end

```

Problem2

✦result image from the output .gif:

Analysis:

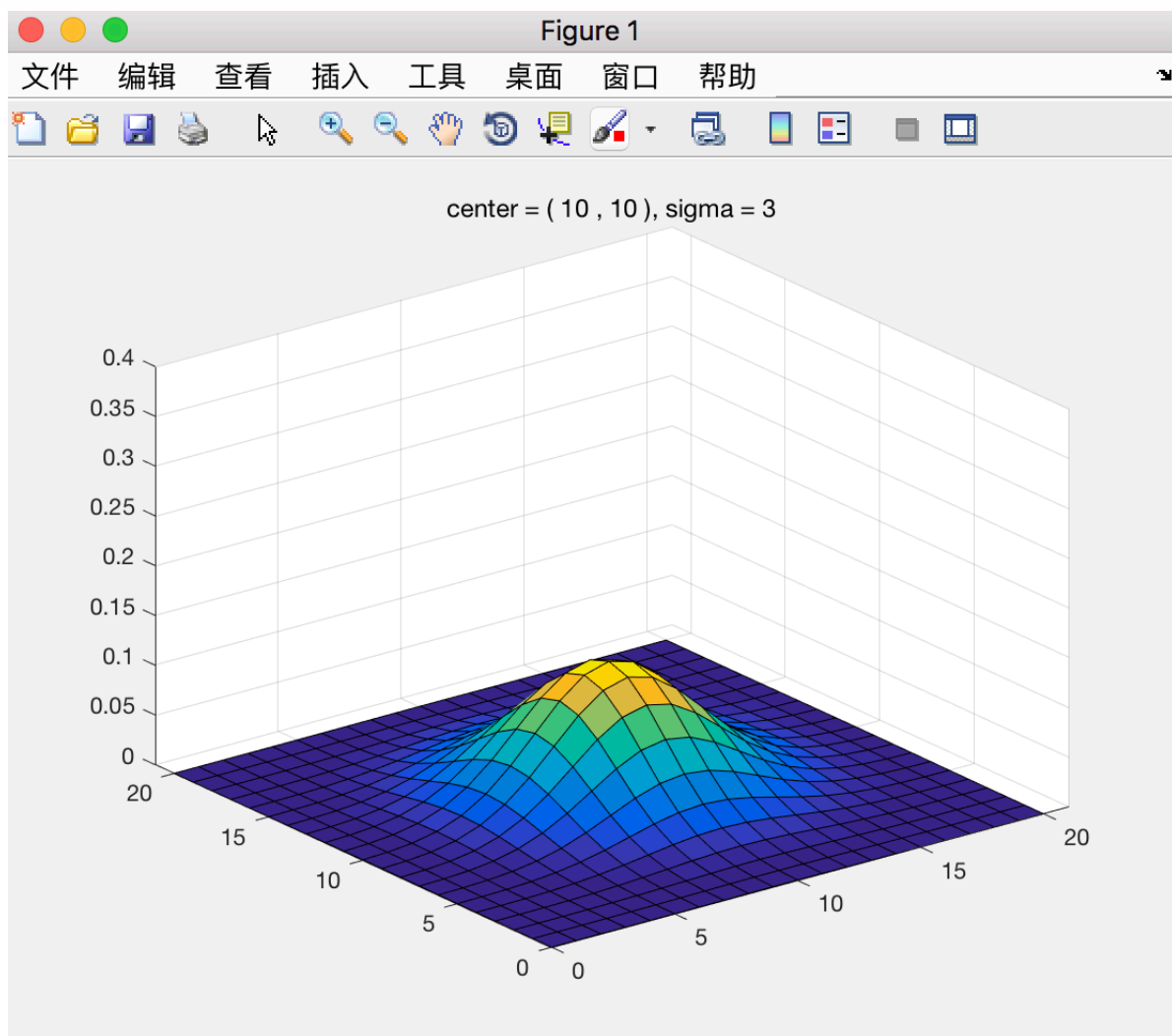
The sigma is changed from 1 to 9; the center is (10,10);

The gaussian expression is

$$f(x,y) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{((x-x_0)^2 + (y-y_0)^2)}{2\sigma^2}}$$

(x_0, y_0) is the center, σ is the sigma, which changed from 1 to 9 in this ***GaussianFunction.gif***

It can be seen from the .gif that as the sigma grows, the gaussian curve flattens out.



code:

```

clear all
clc
X = 0 : 1 : 20;
Y = 0 : 1 : 20;
figure
filename = 'GaussianFunction.gif';

for sigma = 1:9
    % calculate the value of Z
    Z = zeros(21, 21);
    for row = 1 : 1 : 21
        for col = 1 : 1 : 21
            Z(row, col) = (X(row) - 10) .* (X(row)-10) + (Y(col) - 10) .*
(Y(col) - 10);
        end
    end
    Z = -Z/(2*sigma*sigma);
    Z = exp(Z) / (sqrt(2*pi) * sqrt(sigma*sigma));

    % show the gaussian surface
    surf(X, Y, Z);
    title(sprintf(' center = ( 10 , 10 ), sigma = %d ',sigma));
    axis([0 21 0 21 0 0.1]);

    drawnow
    % get the frame
    frame = getframe(gcf);

    %%To make GIF files, images must be index images
    im = frame2im(frame);
    [A,map] = rgb2ind(im,256);

    %create it at the first time
    if sigma == 1
        imwrite(A,map,filename,'gif','LoopCount',Inf,'DelayTime',0.2);
    else
        %DelayTime is used to set up GIF files to play fast or slow
        imwrite(A,map,filename,'gif','WriteMode','append','DelayTime',0.2);
    end
end
end

```

Problem3

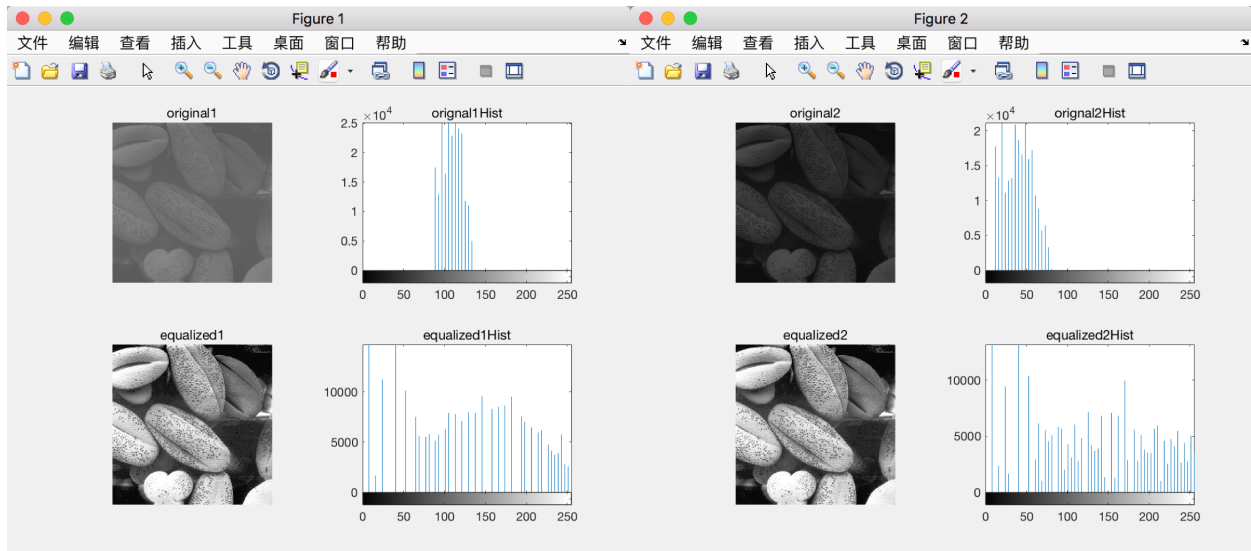
❖result image:

On the left is figure and Hist distribution for imageQ_1_1 before and after histogram equalization;

On the right is figure and Hist distribution for imageQ_1_2 before and after histogram equalization.

Analysis:

- Since it is discrete, equalized figures are not strictly uniform.
- Statistics from local image histograms can be used for local image enhancement; it can enhance dark areas while leaving the light area as unchanged as possible.



code:


```
img1=imread('Q_1_1.tif');
img2=imread('Q_1_2.tif');

%Implement the histogram equalization
im_histeq1=histeq(img1);
im_histeq2=histeq(img2);

%show figure1 before equalized
figure(1);
subplot(2,2,1);
imshow(img1);
title('original1')

%show figure1's Hist before equalized
subplot(2,2,2);
imhist(img1,64);
title('originalHist')

%show figure1 after equalized
subplot(2,2,3);
imshow(im_histeq1);
title('equalized1')

%show figure1's Hist after equalized
subplot(2,2,4);
imhist(im_histeq1,64);
title('equalized1Hist')

%show figure2 before equalized
figure(2);
subplot(2,2,1);
imshow(img2);
title('original2')

%show figure2's Hist before equalized
subplot(2,2,2);
imhist(img2,64);
title('original2Hist')

%show figure2 after equalized
subplot(2,2,3);
imshow(im_histeq2);
title('equalized2')

%show figure2's Hist after equalized
subplot(2,2,4);
imhist(im_histeq2,64);
title('equalized2Hist')
```

Problem4

❖result image:

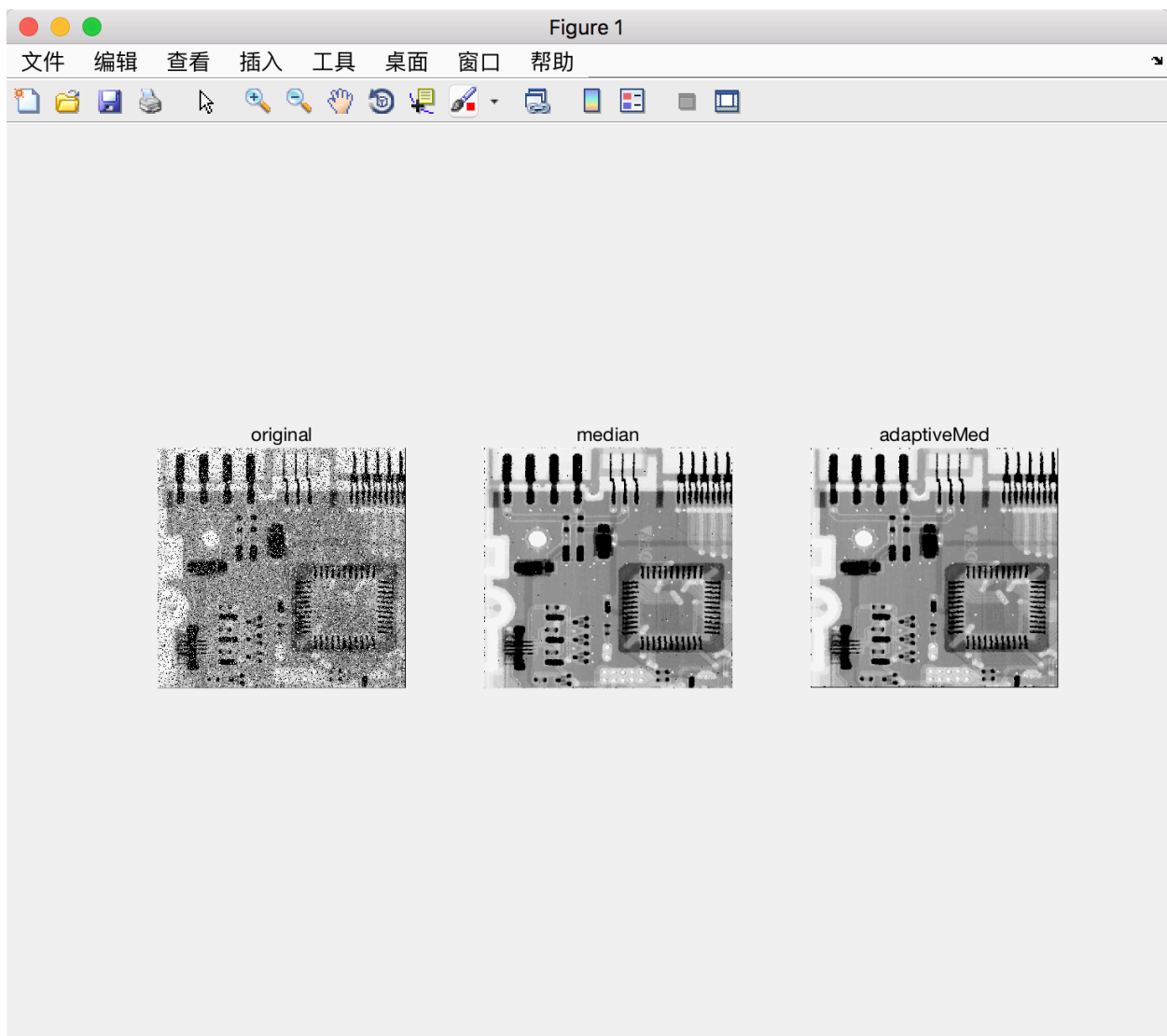
The picture in the middle uses median filter in Matlab;

the picture on the right uses adaptive median filter implemented by myself in **homework1_(2)**

Analysis:

we can see the result of adaptive median filter have a better result than median filter provided by Matlab.

- The adaptive median filter can handle much more spatially dense impulse noise, and also performs some smoothing for non-impulse noise
- The key insight in the adaptive median filter is that the filtersize changes depending on the characteristics of the image



❖code:

```
img=imread('Q_2.tif');  
figure;  
% Using median filtering  
img_med = medfilt2(img);
```

```

% Using adaptiveMedian filtering
img_adaptive = adaptiveMedianFilter(img,11);

% original image
subplot(1,3,1);
imshow(img);
title('original')

% noised image
subplot(1,3,2);
imshow(img_med);
title('median')

% adaptive median filter image
subplot(1,3,3);
imshow(img_adaptive);
title('adaptiveMed')

function[img_result] = adaptiveMedianFilter(img_noi,max_window)

padding_width = (max_window-1)/2;
[m ,n] = size(img_noi);
% Initialize a image equal size with img_noi to receive pixels after
Adaptive Median Filter
img_result = img_noi;
% Fill pixels at the edge of the original image
img_noi = padarray(img_noi,[padding_width padding_width],0);

for i = padding_width+1:padding_width+m
    for j = padding_width+1:padding_width+n
        Zxy = img_noi(i,j);
        window = 3;
        while(window <= max_window)
            %Take the window of the filter plate
            tmp = img_noi(i-(window-1)/2:i+(window-1)/2,...
                j-(window-1)/2:j+(window-1)/2);
            Zmin = min(min(tmp));
            Zmax = max(max(tmp));
            Zmed = median(tmp(:));
            if((Zmed > Zmin) && (Zmed < Zmax))
                if((Zxy > Zmin) && (Zxy < Zmax))
                    img_result(i-padding_width,j-padding_width) = Zxy;
                else
                    img_result(i-padding_width,j-padding_width) = Zmed;
                end
            end
            break;
        end
        window = window+2;
    end
end

```

```

        img_result(i-padding_width,j-padding_width) = Zmed;
    end
end
end

```

Problem5

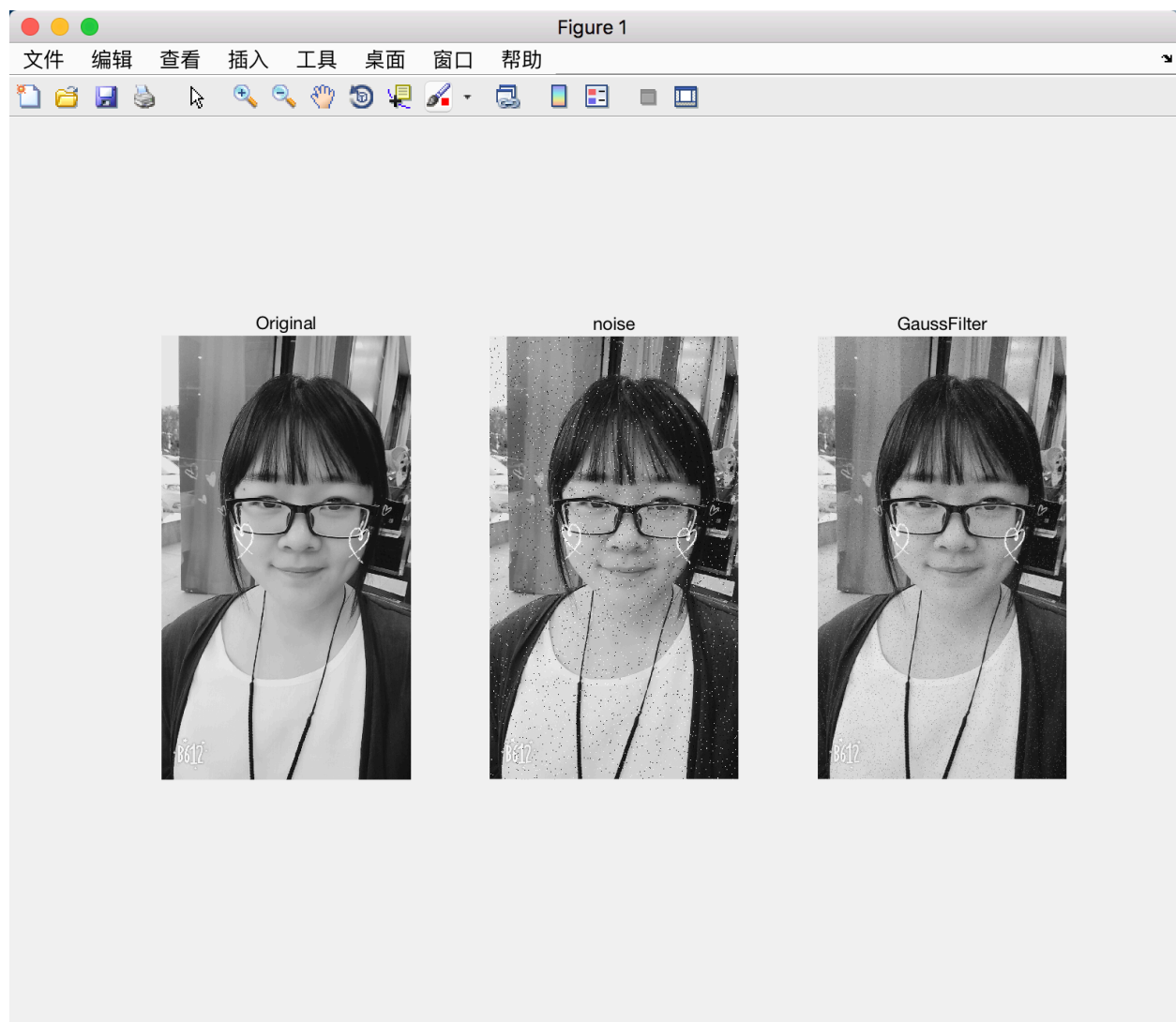
(1) Create a gauss filter of size 4*4, and apply the filter to the image with convolution, padding

✦result image:

Analysis:

According to **the rule of thumb**: set filter half-width to about 3σ , and the required size is 4, so I set the sigma to 2/3;

The picture on the right is the result image after GaussFilter.



✦code:

Here I provide 3 implements code using different methods:

- Impletment using **imfilter**:

```

img=imread('cui.jpg');
img_gray = rgb2gray(img);
figure;

% add noise
img_noi = imnoise(img_gray,'salt & pepper',0.02);

% Rule of thumb: set filter half-width to about 3σ
sigma = 0.5;
% create a Gauss filter of size 4x4
gausFilter = fspecial('gaussian', [3,3], sigma);

% Gauss filtering
gaus= imfilter(img_noi, gausFilter, 'replicate');

subplot(1,3,1);
imshow(img_gray);
title('Original')

subplot(1,3,2);
imshow(img_noi);
title('noise')

subplot(1,3,3);
imshow(gaus);
title('GaussFilter')

```

- Impletment using **conv2**:

```

img=imread('cui.jpg');
img_gray = rgb2gray(img);
figure;

% add noise
img_noi = imnoise(img_gray,'salt & pepper',0.02);

% Rule of thumb: set filter half-width to about 3σ
sigma = 0.5;
% create a Gauss filter of size 4x4
gausFilter = fspecial('gaussian', [3,3], sigma);
% convolution and padding
img_convo = conv2(double(img_noi),gausFilter,'same');
img8_convo = uint8(img_convo);

subplot(1,3,1);
imshow(img_gray);
title('Original')

subplot(1,3,2);
imshow(img_noi);
title('noise')

subplot(1,3,3);
imshow(img_convo);
title('GaussFilter')

```

- Implement convolution and padding by myself:

```

img=imread('cui.jpg');
img_gray = rgb2gray(img);
figure;

% add noise
img_noi = imnoise(img_gray,'salt & pepper',0.02);
img_gaus = img_noi;

% Rule of thumb: set filter half-width to about 3σ
sigma = 0.5;
% create a Gauss filter of size 4x4
gausFilter = fspecial('gaussian', [3,3], sigma);

[m ,n] = size(img_noi);
% padding
padding_width = 2;
img_noi = padarray(img_noi,[padding_width padding_width],0);
% rotate the gausFilter
gausFilter = rot90(gausFilter,2);
% convolution
for i = padding_width+1:padding_width+m
    for j = padding_width+1:padding_width+n
        temp = img_noi(i-1:i+1,j-1:j+1);
        img_gaus(i-padding_width,j-padding_width) =
sum(sum(double(temp).*gausFilter));
    end
end

subplot(1,3,1);
imshow(img_gray);
title('Original')

subplot(1,3,2);
imshow(img_noi);
title('noise')

subplot(1,3,3);
imshow(img_gaus);
title('GaussFilter')

```

(2) Give one intensity transformation function for spreading the intensities of an image such that the lowest is I_{min} and the highest is I_{max} , ($0 \leq I_{min} \leq I_{max} \leq 255$). Denote by F_{max} and F_{min} the maximum and minimum intensities values of the input image.

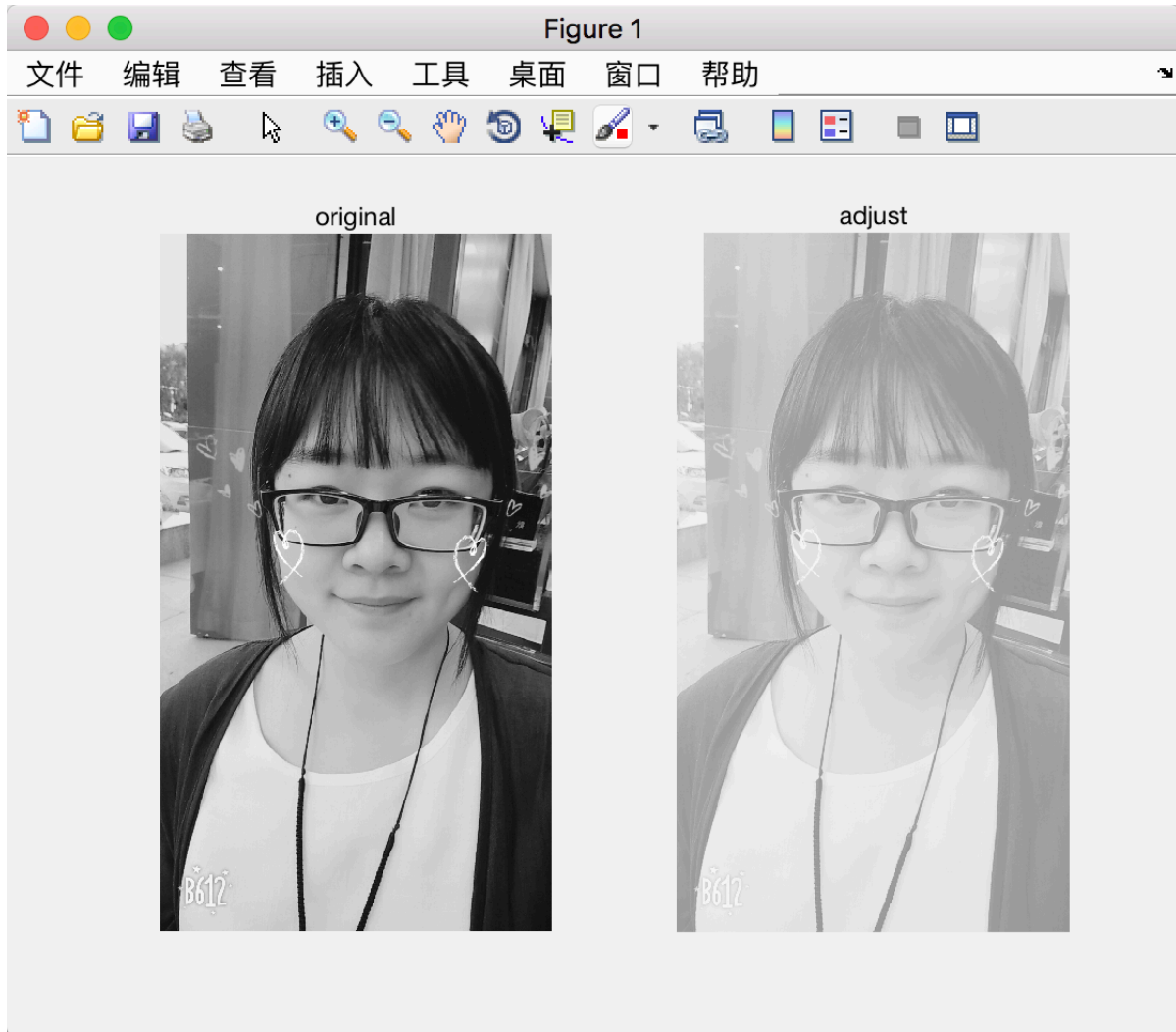
❖result image:

Analysis:

Running the program, it will first let the user input the target range of intensities values (0~255);

Then, it will cast the maximum and minimum intensities values of the input image to the target.

In the code part, I provide 2 implements, one is using Matlab function ***imadjust***, the other is a linear mapping ($\sigma=1$) without using Matlab Function.



✦code:

Here I provide two implement code using different methods:

- Implement Matlab function ***imadjust***:


```

img=imread('cui.jpg');
img_gray = rgb2gray(img);

% let the user input the target range of intensities values
targetMax = input('Please input target I_max(0~255): ');
targetMin = input('Please input target I_min(0~255): ');
% Change the target range from 0 to 1 to use the adjust function
targetMax = targetMax/256;
targetMin = targetMin/256;

% find the maximum and minimum intensities values of the input image
counts = imhist(img_gray);
minBinValue = find(counts>0, 1, 'first');
minBinValue = minBinValue/256;
maxBinValue = find(counts>0, 1, 'last');
maxBinValue = maxBinValue/256;

% spreading the intensities of the input image
img_adj = imadjust(img_gray,[minBinValue,maxBinValue],
[targetMin,targetMax]);

figure;

subplot(1,2,1);
imshow(img_gray);
title('original')

subplot(1,2,2);
imshow(img_adj);
title('adjust')

```

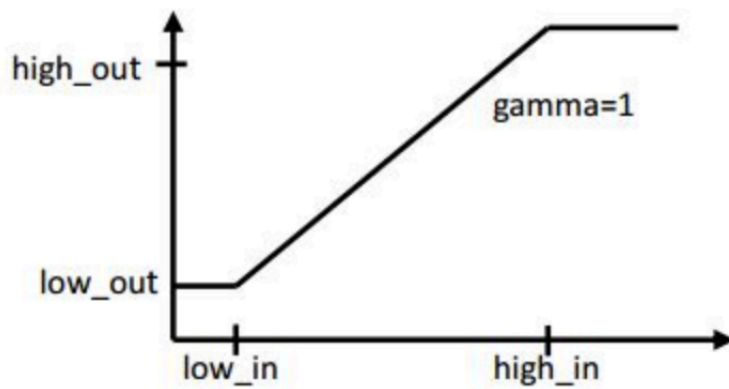
- Implement without Matlab function *imadjust*:

Analysis:

Suppose $*J = \text{imadjust}(I, [\text{low_in}; \text{high_in}], [\text{low_out}; \text{high_out}], \text{gamma})^*$

Map the brightness value in image I to the new value in J , and the value between low_in and high_in is mapped to the value between low_out and high_out . The values below low_in and above are cut off, which means that the values below low_in are mapped to low_out , and the values above high_in are mapped to high_out .

The gray mapping process is shown as the followed picture:



```

img=imread('cui.jpg');
img_gray = rgb2gray(img);
img_adj = img_gray;

% let the user input the target range of intensities values
targetMax = input('Please input target I_max(0~255): ');
targetMin = input('Please input target I_min(0~255): ');

% Linear projection imadjust
[m,n] = size(img_gray);
for i = 1:m
    for j = 1:n
        if img_gray(i,j) < targetMin
            img_adj(i,j) = targetMin;
        elseif img_gray(i,j) > targetMax
            img_adj(i,j) = targetMax;
        end
    end
end

figure;
subplot(1,2,1);
imshow(img_gray);
title('original')

subplot(1,2,2);
imshow(img_adj);
title('adjust')

```