Image Processing Programming Assignment #2

In this assignment, our goal is to implement some image segmentation techniques. There are two tasks in total, which are, trying various gradient filters on the images and choose one between the implementation of the Canny edge detector, Hough transforms, or watershed algorithm for segmentation. The code is done with Matlab programming language and is released on https://github.com/KennethYapWL/NCTU-IP-2020/tree/main/Project%202, and the references are listed on the last page of this report.

Those methods implemented are:

- 1. Gradient Filters
 - → Prewitt filter
 - → Sobel filter
 - \rightarrow LoG filter
- 2. Canny Edge Detection
- 3. Thresholding
- 4. Conversion between color models (RGB, YCbCr, HSI, HSV)
- 5. And some preprocessing techniques used in the previous project.

This report is organized as follow:

- → Section 1 (Experimental Results): This part will show all the results of six different images after doing edge detection.
- → Section 2 (Observations and Discussions): This part will give brief descriptions of the other experiment I have tried, also the further observations of section 1 will be discussed in this section.
- → Section 3 (Code Analysis): This part will show all the code used in this assignment.
- → Section 4 (References): This part will list all the articles I have read.

Section 1: Experimental Results

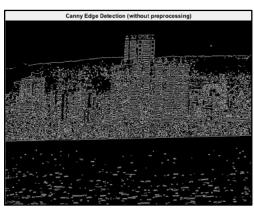
This section is focused on showing the results of the edge detection applied to six different images. 4 of the images are from the previous project, and the other two images are from Google.

Note that the value of the threshold in Prewitt, Sobel, and LoG is between 0 to 1. For example, 0.5 means threshold value of 255 * 0.5 = 127.5.

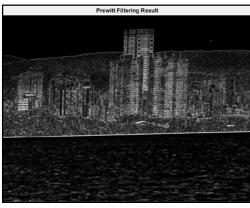
A) The first image (p1im1.png)



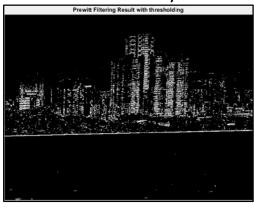
Original Image



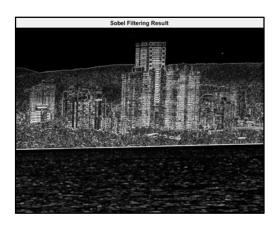
Canny (sigma=0.15, low threshold=50, high threshold=90)



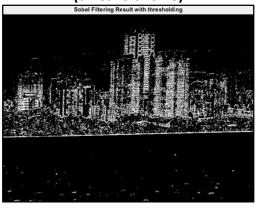
Prewitt



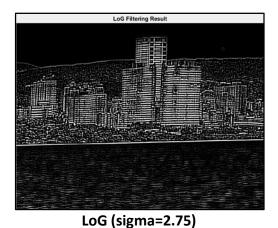
Prewitt with thresholding (threshold=0.48)

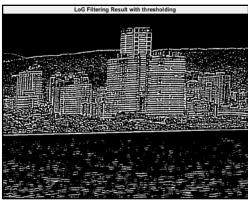


Sobel



Sobel with thresholding (threshold=0.49)





LoG with thresholding (0.25)

We could see that the result of Sobel and Prewitt are quite similar to each other, but Sobel is letting more pixels as edge points. LoG is more able to catch the edge information on the 'ocean' part and the 'mountain' part in the image.

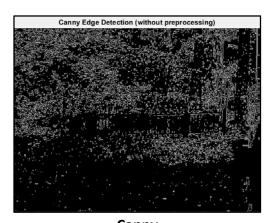
The result of Canny edge detection is quite different to the result of those gradient filters. Notice that some of the edge point of the Sobel, Prewitt and LoG doesn't connect to another edge point, while each Canny edge point seems to have connected to form edge lines.

More details can be found on the file "Sol_p1im1.m" in the provided GitHub.

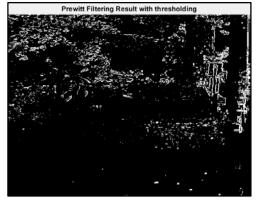
B) The second image (p1im2.png)



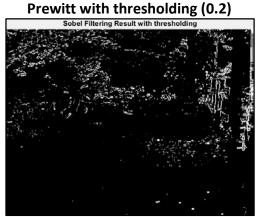




Canny (sigma=0.08, low threshold=30, high threshold=45)



Sobel Filtering Result Sobel Filtering Result





Sobel with thresholding (0.3)

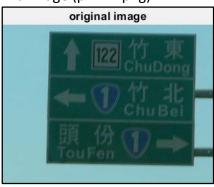
Log Filtering Result with thresholding

LoG(sigma=2.3)

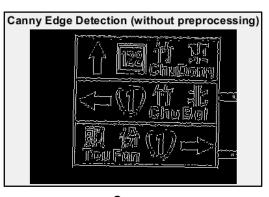
LoG with thresholding (0.4)

More details can be found on the file "Sol_p1im2.m" in the provided GitHub.

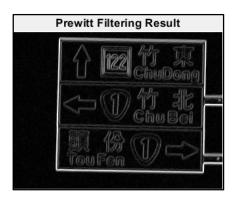
C) The third image (p1im4.png)

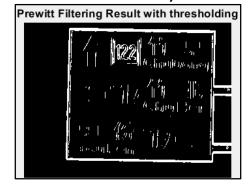


Original Image

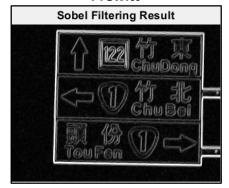


Canny (sigma=0.00001, low threshold=20, high threshold=50)

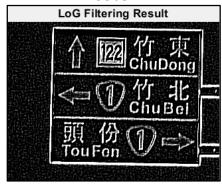




Prewitt

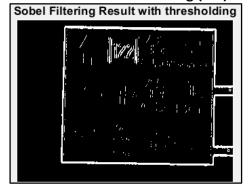


Sobel

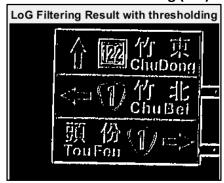


LoG(sigma=1.71)

Prewitt with thresholding (0.3)



Sobel with thresholding (0.5)



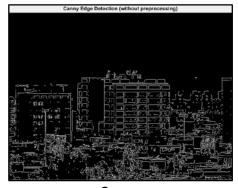
LoG with thresholding (0.75)

More details can be found on the file "Sol_p1im4.m" in the provided GitHub.

D) The forth image (p1im5.png)



Original Image



Canny (sigma=0.0005, low threshold=20, high threshold=50)



Prewitt



Prewitt with thresholding (0.2)



LoG(sigma=2.2)



Sobel with thresholding (0.2)



LoG with thresholding (0.35)

More details can be found on the file "Sol_p1im5.m" in the provided GitHub.

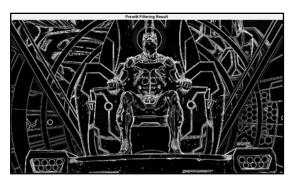
E) The fifth image [From Google] (batman_god_of_knowledge.png)



Original Image



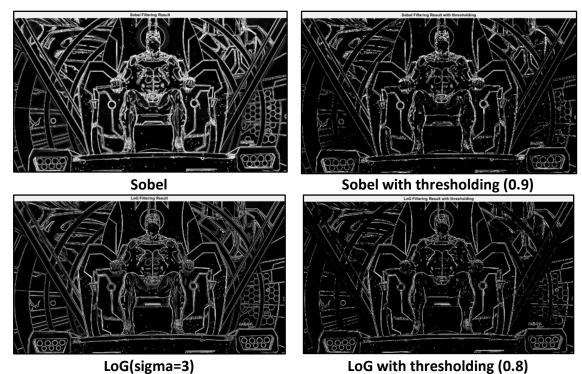
Canny (sigma=0.8, low threshold=40, high threshold=80)



Prewitt



Prewitt with thresholding (0.9)



More details can be found on the file "Sol_ batman_god_of_knowledge.m" in the provided GitHub.

F) The sixth image [From Google] (batman_three_jokers.png)

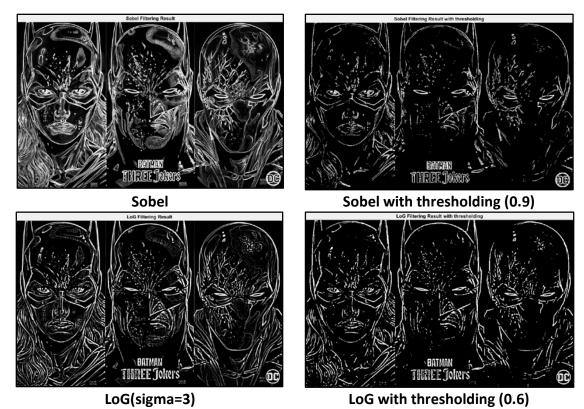


Prewitt



threshold=90)

Prewitt with thresholding (0.8)



More details can be found on the file "Sol_ batman_three_jokers.m" in the provided GitHub.

Section 2: Observations and Discussions

1. How pre-processing affects edge detection ???

As assignment asked, I have done some experiments about how pre-processing could affects edge detection. This section will only show one of them, like previous project, the image (p1im1.png) is first pre-processed by the following steps:

- 1) Piecewise Linear Stretching, with settings, r1=120, s1=70, r2=250, s2=190. Figure 1.A.3 shows the transform function of these settings.
- 2) Adaptive Local Noise Reduction filtering, with a filter size of 3.
- 3) Color Correction by adjusting the HSV components, with H and S remain unchanged, and V subtract 40 (V = V - 40).

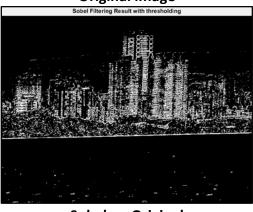


Original Image

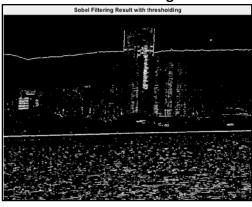


After Preprocessing

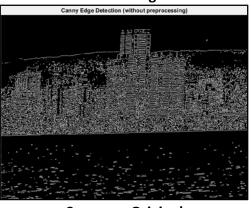
Enhanced Image



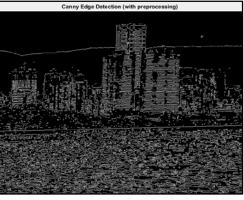
Sobel on Original



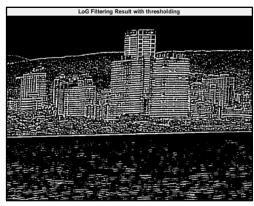
Sobel on Enhanced

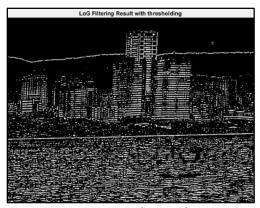


Canny on Original



Canny on Enhanced





LoG on Original

LoG on Enhanced

As the results shown, the experiment found that contrast adjustment or color correction usually will change the focus point of edge detection process. For example, LoG on original image had more edge candidate points in "mountain" and "building" while LoG on enhanced image is focus on the "ocean" part.

Besides, the experiment also found that, the smoothing prevent edge detection process affected from noises. For the four images supplied from previous project, similar preprocessing are done, and the results of edge detection on each image are achieved.

More details can be found on the files "Sol_p1im1.m", "Sol_p1im2.m", "Sol_p1im4.m", and "Sol_p1im5.m" in the provided GitHub.

2. Gradient filtering on different color model.

It is interesting to know what outcome will be produced by applying gradient filters on different color model. This experiment not just doing gradient filters on gray scale images, but also do that on other color model such as each component of RGB, Intensity component of HSI, Y component of YCbCr. The experiment below shows the result of using Sobel filter on different color models.



Original Image



Sobel on Gray Scale





Sobel on HSI

Sobel on RGB



Sobel on YCbCr

Notice that the result on YCbCr is closest to the original image. Edge detection can also produce a 3 channel output, one of the common color model choice they have used is YCbCr.

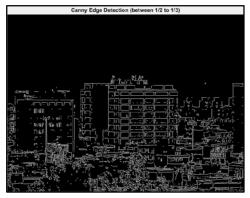
More details can be found on the file "sobel_on_different_color_model.m" in the provided GitHub.

3. What is the common ratio between higher threshold and lower threshold in canny edge detection.

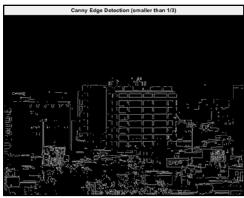
As stated in ref[1], the thresholds used in canny edge detection is usually on the ratio of $\frac{lower_threshold}{higher_threshold} \in [\frac{1}{3}, \frac{1}{2}] \text{ . A simple experiment had been done to see what happen if we}$ didn't follow this ratio.



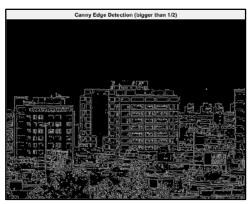
Original Image



Canny with ratio in [1/3,1/2]



Canny with ratio smaller than 1/3



Canny with ratio greater than 1/2

From the result above, by comparing with the Canny Edge with ratio in [1/3, 1/2], the result of Canny Edge with ratio smaller than 1/3 lost some edge candidate points, while the Canny Edge with ratio greater than 1/2 had too much edge candidates, which some of them might not suitable to form an edge.

More details can be found on the file "ratio_between_thresholds.m" in the provided GitHub.

Section 3: Code Analysis

Since the content in files "Sol_p1im1.m", "Sol_p1im2.m", "Sol_p1im4.m", "Sol_p1im5.m", "Sol_ batman_god_of_knowledge.m", and "Sol_ batman_three_jokers.m" are already described in Section 1, these files will not be shown in this section. (There are almost more than 200 lines of code in each of the files)

Note that the explanations of each part of the code are shown in the comment (font color in light green).

- 1. Gradient filter.
 - → Prewitt filtering

→ Sobel filtering

```
ction transImgMatr = sobel_filtering(imgMatr, direction)
% ref: https://www.geeksforgeeks.org/matlab-image-edge-detection-using-
imgMatr = double(imgMatr);
hg = size(imgMatr,1); % get the height of image
wd = size(imgMatr,2); % get the width of image
% zero padding on image matrix
filtSize = 3;
paddMatr = padarray(imgMatr,[shift shift],0);
sobely = [-1,-2,-1; 0,0,0; 1,2,1];
gx = zeros(hg,wd);
gy = zeros(hg,wd);
for row = 1 : hg
    for col = 1 : wd
        mul = paddMatr(row : row + (filtSize - 1),col : col + (filtSize - 1)).* sobelx;
        gx(row,col) = sum(mul(:));
        mul = paddMatr(row : row + (filtSize - 1),col : col + (filtSize - 1)).* sobely;
        gy(row,col) = sum(mul(:));
        switch direction
               gx(row,col) = 0; % remove gx
               gy(row,col) = 0; % remove gy
grad = sqrt((gx .^2) + (gy .^2));
transImgMatr = uint8(grad);
```

→ LoG filtering

2. Canny Edge Detection.

In this implementation, I divided steps into several function.

→ Canny Edge Detection Main Function

```
ransImgMatr = canny_edge_detection(imgMatr, sigma, low_thres, high
% This function perform the canny edge detection
    . https://towardsdatascience.com/canny-edge-detection-step-by-step-in-python-computer-vision-b49c3a2d8123
. https://www.mathworks.com/matlabcentral/fileexchange/41221-canny-edge-detector
% return transformed image matrix
imgMatr = double(imgMatr);
transImgMatr = smoothing_with_gaussian(imgMatr, sigma);
[transImgMatr, theta] = gradient_calculation(transImgMatr);
transImgMatr = non_maxima_suppression(transImgMatr, theta);
transImgMatr = double thresholding(transImgMatr, low thres, high thres)
transImgMatr = hysteresis edge tracking(transImgMatr);
transImgMatr = uint8(transImgMatr);
```

→ Smoothing with Gaussian

```
tion transImgMatr = smoothing_with_gaussian(imgMatr, sigma, <u>filtSz</u>)
hg = size(imgMatr,1); % get the height of image wd = size(imgMatr,2); % get the width of image
filtSz = ceil(sigma) * 5;
filtsz = (filtsz - 1) / 2;
paddMatr = padarray(imgMatr,[shift shift],'replicate');
Gauss1D = \exp(-(x.^2/(2*sigma^2)));
GaussX=reshape(Gauss1D,[length(Gauss1D) 1]);
GaussY=reshape(Gauss1D,[1 length(Gauss1D)]);
GaussMatr = zeros(length(GaussX),length(GaussY));
    for y = 1 : length(GaussY)
        GaussMatr(x,y) = GaussX(x) * GaussY(y);
% create the matrix with width and height
transImgMatr = zeros(hg,wd);
        for x = 1 : filtSz
for y = 1 : filtSz
                 tmp = tmp + GaussMatr(x,y) * paddMatr(x + (row - 1), y + (col - 1));
        transImgMatr(row,col) = tmp;
```

→ Gradient Calculation Function

```
ion [transImgMatr, theta] = gradient_calculation(imgMatr)
hg = size(imgMatr,1); % get the height of imag
wd = size(imgMatr,2); % get the width of image
filtSize = 3;
paddMatr = padarray(imgMatr,[shift shift],0);
sobelx = [-1,0,1; -2,0,2; -1,0,1];

sobely = [-1,-2,-1; 0,0,0; 1,2,1];
\$ create the matrix with width and height \$ exacly same as the input image to store the transformed pixel values
for row = 1 : hg
    for col = 1 : wd
         mul = paddMatr(row : row + (filtSize - 1),col : col + (filtSize - 1)).* sobelx;
         gx(row,col) = sum(mul(:));
         mul = paddMatr(row : row + (filtSize - 1),col : col + (filtSize - 1)).* sobely;
grad = sqrt((gx .^2) + (gy .^2));
%grad = grad / max(grad(:)) * 255;
transImgMatr = grad;
[gradHg, gradWd] = size(gx);
epsilon = 0.0000000001; % prevent infinity results (in case there could be 0 in gx) theta = atan(gy./(gx + epsilon)) * (180/3.1412);
 for row = 1 : gradHg
     for col = 1 : gradWd
    if (theta(row,col) < 0)</pre>
for row = 1 : gradHg
    elseif (22.5 <= theta(row,col)) && (theta(row,col) < 67.5)</pre>
       theta(row,col) = 45;
elseif (67.5 <= theta(row,col)) && (theta(row,col) < 112.5)</pre>
       theta(row,col) = 90;
elseif (112.5 <= theta(row,col)) && (theta(row,col) < 157.5)
            theta(row,col) = 135;
```

→ Non Maxima Suppression Function

```
function transImgMatr = non_maxima_suppression(Grad, theta)
    GradHg = size(Grad, 1); % get the height of gradient
    GradWd = size(Grad, 2); % get the width of gradient

paddMatr = padarray(Grad, (1 1),0); % zero padding

% create the matrix with width and height
 % exacly same as the input image to store the transformed pixel values
    transImgMatr = zeros(GradHq,GradWd);
```

→ Double Thresholding Function

→ Hysteresis Edge Tracking Function

3. Thresholding.

→ Thresholding and convert image to binary image

```
function imgMatr = convert_to_binary_image(imgMatr, threshold)

% This function perform the conversion from image to binary image
% note that threshold value is between 0 to 1
% ref: it's same to builtin function --> im2bw()
% return transformed image matrix

imgMatr = uint8(imgMatr);
th = 255 * threshold;

imgMatr(imgMatr < th) = 0;
imgMatr(imgMatr >= th) = 1;

imgMatr = double(imgMatr);
```

4. Preprocessing Functions

This actually already described on previous assignment, here will only show the preprocessing functions that are used in this assignment.

→ Gaussian Smoothing Filter

```
ion transImgMatr = gaussian_filtering(imgMatr,sigma,filtSize)
\mbox{\$} This function perform the noise removal with gaussian filtering \mbox{\$} return transformed image matrix
hg = size(imgMatr,1); % get the height of image
wd = size(imgMatr,2); % get the width of image
% replicate padding on image matrix
shift = double(floor(filtSize / 2));
paddMatr = padarray(imgMatr,[shift shift],'replicate');
% Make 2D Gaussian kernel x=-ceil(shift):ceil(shift);
\begin{aligned} & \text{Gauss1D} = \exp(-(x.^2/(2*\text{sigma}^2))); \\ & \text{Gauss1D} = & \text{Gauss1D/sum}(\text{Gauss1D}(:)); \end{aligned}
GaussX=reshape(Gauss1D, [length(Gauss1D) 1]);
GaussY=reshape(Gauss1D,[1 length(Gauss1D)]);
GaussMatr = zeros(length(GaussX),length(GaussY));
for x = 1 : length(GaussX)
    for y = 1 : length(GaussY)
            GaussMatr(x,y) = GaussX(x) * GaussY(y);
\mbox{\tt \$} create the matrix with width and height \mbox{\tt \$} exacly same as the input image to store the transformed pixel values
transImgMatr = zeros(hg,wd);
 % Do spatial filtering
for row = 1 : hg
    for col = 1 : wd
             tmp = 0;
            for x = 1 : filtSize
    for y = 1 : filtSize
                        tmp = tmp + GaussMatr(x,y) * paddMatr(x + (row - 1), y + (col - 1));
            transImgMatr(row,col) = tmp;
 transImgMatr = uint8(transImgMatr);
```

→ Adaptive Filter

```
nction transImgMatr = adaptive filtering(imgMatr,filtWd,filtHg)
 imgMatr = double(imgMatr);
 hg = size(imgMatr,1); % get the height of image
 wd = size(imgMatr,2); % get the width of image
 % zero padding on image matrix
 shiftHg = double(floor(filtHg / 2));
 shiftWd = double(floor(filtWd / 2));
 paddMatr = padarray(imgMatr,[shiftHg shiftWd],0);
 transImgMatr = zeros(hg,wd);
 local_var = zeros(hg,wd);
 local_mean = zeros(hg,wd);
     for col = 1 : wd
         filter = paddMatr(row : row + (filtHg - 1),col : col + (filtWd - 1));
         local_mean(row,col) = mean(filter(:));
         local_var(row,col) = mean(filter(:).^2) - mean(filter(:)).^2;
 noise var = mean(local var(:));
 local_var = max(local_var,noise_var);
 transImgMatr = imgMatr - (noise_var./local_var).*(imgMatr - local_mean);
  transImgMatr = uint8(transImgMatr);
```

→ Bilateral Filter

```
on transImgMatr = bilateral_filtering(imgMatr,filtSize,sigma_d,sigma_g
imgMatr = double(imgMatr);
hg = size(imgMatr,1); % get the height of image
wd = size(imgMatr,2); % get the width of image
% replicate padding on image matrix
shift = double(floor(filtSize / 2));
paddMatr = padarray(imgMatr,[shift shift],0);
Gauss1D = exp(-(x.^2 / (2 * sigma_d^2)));
Gauss1D = Gauss1D / sum(Gauss1D(:));
GaussX=reshape(Gauss1D, [length(Gauss1D) 1]);
GaussY=reshape(Gauss1D, [1 length(Gauss1D)]);
GaussMatr = zeros(length(GaussX),length(GaussY));
for x = 1 : length(GaussX)
        GaussMatr(x,y) = GaussX(x) * GaussY(y);
transImgMatr = zeros(hg,wd);
for row = 1 : hg
for col = 1 : wd
         L = paddMatr(row : row + (filtSize - 1),col : col + (filtSize - 1));
         H = \exp(-(L - imgMatr(row,col).^2) / (2 * sigma_g^2));
          Bilt = H.* GaussMatr;
          transImgMatr(row,col) = sum(Bilt(:).* L(:)) / sum(Bilt(:));
     end
transImgMatr = uint8(transImgMatr);
```

- 5. Sharpening Code.
 - → Laplacian Filter

- 6. Color Space Conversion Code.
 - → Conversion of RGB to HSI

→ Conversion of HSI to RGB

→ Conversion of RGB to HSV

```
for row = 1 : hg
    for col = 1 : wd
       R = imgMatr_R(row,col);
       G = imgMatr_G(row,col);
        B = imgMatr_B(row,col);
        Max = max(R, max(G, B));
       Min = min(R, min(G, B));
        elseif (Max == R)
            if (G < B) H = H + 360; end
        elseif (Max == G)
        H = 60 * (B - R) / delta + 120;
elseif (Max == B)
       H = H + 360;
       H = H / 360.0;
      if (Max == 0)
       s = 1 - Min / Max;
       V = Max;
       H_matr(row,col) = H;
       S_matr(row,col) = S;
       V_matr(row,col) = V;
end
HSV = cat(3,H_matr,S_matr,V_matr);
```

→ Conversion of HSV to RGB

```
for row = 1 : hg
       H = imgMatr_H(row,col);
        S = imgMatr_S(row,col);
       V = imgMatr_V(row,col);
       X = C * (1 - abs( mod(H/60,2) - 1 ));

m = V - C;
        if (H >= 0 && H < 60)
           R = C; G = X; B = 0;
        elseif (H >= 60 && H < 120)
           R = X; G = C; B = 0;
        elseif (H >= 120 && H < 180)
          R = 0; G = C; B = X;
        elseif (H >= 180 && H < 240)
            R = 0; G = X; B = C;
        elseif (H >= 240 && H < 300)
        elseif (H >= 300 && H < 306)
           R = C; G = 0; B = X;
       R_{matr(row,col)} = (R + m) * 255;
        G_{matr(row,col)} = (G + m) * 255;
        B \text{ matr(row,col)} = (B + m) * 255;
    end
end
RGB = uint8(cat(3,R_matr,G_matr,B_matr));
```

→ Conversion of RGB to Grayscale

→ Conversion of RGB to YCbCr

→ Conversion of YCbCr to RGB

Section 4: References

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