


```

0    0    0    255  255  255  255  0    0    0
0    0    0    255  255  255  255  0    0    0
0    0    0    255  255  255  255  0    0    0
0    0    0    255  255  255  255  0    0    0
0    0    0    255  255  255  255  0    0    0
0    0    0    0    0    0    0    0    0    0];

```

```

w1 = [1 -2 1]
w2 = [1;-2;1]
Iout1 = conv2(I,w1);
Iout2 = conv2(I,w2);

```

Iout1=

```

0    0    0    0    0    0    0    0    0    0    0    0
0  255 -255    0    0    0    0    0    0 -255  255    0
0  255 -255    0    0    0    0    0    0 -255  255    0
0  255 -255    0    0    0    0    0    0 -255  255    0
0    0    0  255 -255    0    0 -255  255    0    0    0
0    0    0  255 -255    0    0 -255  255    0    0    0
0    0    0  255 -255    0    0 -255  255    0    0    0
0    0    0  255 -255    0    0 -255  255    0    0    0
0    0    0  255 -255    0    0 -255  255    0    0    0
0    0    0    0    0    0    0    0    0    0    0    0

```

Iout2 =

```

0    0    0    0    0    0    0    0    0    0
0  255  255  255  255  255  255  255  255    0
0 -255 -255 -255 -255 -255 -255 -255 -255    0
0    0    0    0    0    0    0    0    0    0
0 -255 -255    0    0    0    0 -255 -255    0
0  255  255    0    0    0    0  255  255    0
0    0    0    0    0    0    0    0    0    0
0    0    0    0    0    0    0    0    0    0
0    0    0    0    0    0    0    0    0    0
0    0    0 -255 -255 -255 -255    0    0    0
0    0    0  255  255  255  255    0    0    0
0    0    0    0    0    0    0    0    0    0

```

first filter is used to detect vertical edges, and second one is used to detect horizontal edges

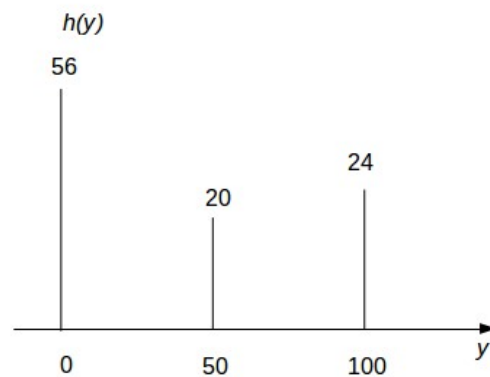
3) Apply the Sobel in Figure 3.41 (d),(e) to the *I* matrix above

4) Draw the histogram of the following matrix

$I_2 = I$

0	0	0	0	0	0	0	0
100	100	100	100	100	100	100	0
100	100	100	100	100	100	100	0
100	100	100	100	100	100	100	0
0	50	50	50	50	0	0	0
0	50	50	50	50	0	0	0
0	50	50	50	50	0	0	0
0	50	50	50	50	0	0	0
0	50	50	50	50	0	0	0
0	0	0	0	0	0	0	0

I



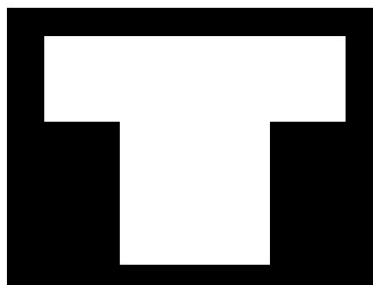
5) Suggest an intensity transformation function that can increase the contrast of the previous image

→ Original image:



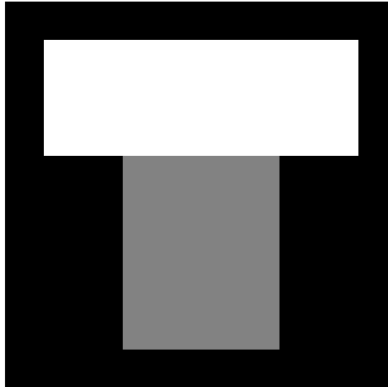
$y_2 = 0$ if $y < 10$, $y_2 = 255$ if $y \geq 10$

This way, the foreground will be all white, and the background will be all black



Another function

$y_2 = 0$ if $y < 10$, $y_2 = 128$ if $10 \leq y < 70$, $y_2 = 255$ if $y \geq 70$



6) In Gimp, in the Filters → blur. There is an option to (Pixelize). Write an octave/matlab program to do the same task

```
function I2 = pixelize2(I1,width,height)
% I1 is the input image, width and height are the widths and heights of different blocks
[tot_rows,tot_cols] = size(I1);
num_block_rows = ceil(tot_rows/height);
num_block_cols = ceil(tot_cols/width);
I2 = I1; % output image
for u=1:num_block_rows
    for v = 1:num_block_cols
        Iblock = I1(1+(u-1)*height:min(height+(u-1)*height,tot_rows),1+(v-1)*width:min(width+(v-1)*width,tot_cols));
        % I have chosen the min of the argument and the tot_rows or tot_cols to take care of the edges
        I2(1+(u-1)*height:min(height+(u-1)*height,tot_rows),1+(v-1)*width:min(width+(v-1)*width,tot_cols)) = round(mean(Iblock(:)));
    end
end
```

7) In Gimp, in the Filters → blur. There is an option to (Motion blur). Write an octave/matlab program to do the task of (Linear) blur.

Linear blurring is done by a convolution between a mask and the original image. Depending on the mask shape, blurring will be in a certain direction. For example, for a diagonal 45 degrees blurring and length of 4 pixels, the mask will be

```
¼ 0 0 0
0 ¼ 0 0
0 0 ¼ 0
0 0 0 ¼
```

```
function I2 = motion_blur(I,h_blur)
I2 = conv2(I,h_blur);
```

A more accurate and complicated implementation can be understood from here

<http://www.mathworks.com/matlabcentral/answers/15895-what-is-the-kernel-of-linear-motion-blur-in-fspecial-function>

and in octave, after loading the image processing package, writed (edit fspecial) and find (motion) to understand how it is written

This is a snapshot from a paper that describes motion blurring

II. MATHEMATICAL MODEL OF LINEAR MOTION BLURRING AND ITS ATTRIBUTES

Commonly used linear model for image blur is given by:

$$g(x, y) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} h(x, \alpha, y, \beta) f(\alpha, \beta) d\alpha d\beta \quad (1)$$

where $h(x, \alpha, y, \beta)$ is a linear *PSF*, $f(x, y)$ is the ideal image, $g(x, y)$ is the observed image. If we consider the Spatially Invariant case of uniform linear motion along horizontal direction, the *PSF* $h(x, y)$ is given by:

$$h(x, y) = \begin{cases} \frac{1}{L} & \text{if } \sqrt{x^2 + y^2} \leq \frac{L}{2} \text{ and } \frac{x}{y} = -\tan \varphi \\ 0 & \text{otherwise} \end{cases} \quad (2)$$

As seen in equation(2), motion blur depends on two parameters: Motion length(L) and Motion direction (φ). The