# TP1 Analyse d'image

# **Gradients**



Figure 1 Amplitude of gradient in X axis



Figure 2 Amplitude of gradient in Y axis



Figure 3 Amplitude of bi-directional gradient



Figure 4 Amplitude of multi-directional gradient

Gradients are computed using a convolution with the appropriate kernel. The bi-directional gradient is the gradient computed with the X and Y directions, the multi-directional gradient is computed with X, Y, North-West  $(3\pi/4)$ , North-East  $(\pi/4)$ .

## **Thresholding methods**

### **Global thresholding**

### **Explanation**

This method sets to 0 the value of the pixels below a user defined threshold, the pixels above are set to 255.

### <u>Result</u>



Figure 5 Global thresholding, with high threshold set to

### **Observations**

I noticed that this method is not accurate as it misses a lot of areas of where the change of intensity is subtle, resulting in an image with less contours, with the majority contours corresponding to areas of high change in intensity.

### Local thresholding

### **Explanation**

This method computes with a convolution the mean of the gradients around a pixel, sets to 0 the value of the pixels below the local mean and 255 for the pixels above. The default kernel size is 3x3, computer the mean considering 8 neighbors.

### <u>Result</u>



Figure 6 Local threshold, kernel size 11x11

#### Observations

I noticed that this method finds a lot more contours but fails to separate the background from the foreground as well as the global thresholding.

### Hysteresis thresholding

### Explanation

This method applies the same logic as a global threshold for the amplitude values strictly less than the low threshold or strictly higher than the high threshold. There is an extra step for the pixels noted P whose gradient amplitudes are in between the thresholds: it checks if at least one of their neighbors' amplitudes are strictly higher than the high threshold, and if it is, sets the value of P to 255, otherwise, sets it to 0.

The low and high threshold must be chosen carefully, in my implementation, I chose to make the low threshold  $L_t$  depend on the high threshold  $H_t$  based on a user chosen scalar s, such that  $L_t = s * H_t$ 

### Result



Figure 7 Hysteresis thresholding, Ht = 0.13, s = 0.5 (i.e Lt = 0.065)

I noticed that this method found more contours, by also finding contours in areas of subtle change in intensity in addition to areas of high change. I found that this method solved the issues with global thresholding regarding the areas of subtle changes in intensity, by adding a extra local thresholding step. However, the contours are thick and may not correspond to the reality.

#### Non-Maximal Suppression

This method removes pixels that have a gradient value lower than the neighboring pixels in the gradient direction. In my implementation, I chose the neighboring pixels based on the angle of the gradient for the current pixel.

# <u>Result</u>



Figure 8 Non-max suppression on image of bi-directional amplitude



Figure 9 Hysteresis thresholding applied

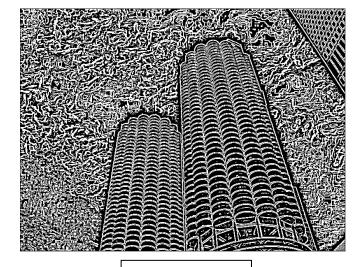
# **Observations**

I noticed that this method had the effect of thinning the contours, noticeable on Fig 9. with a hysteresis threshold. This method, combined with hysteresis thresholding produces an image where the contours are thin and accurate.

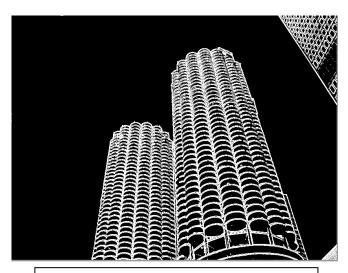




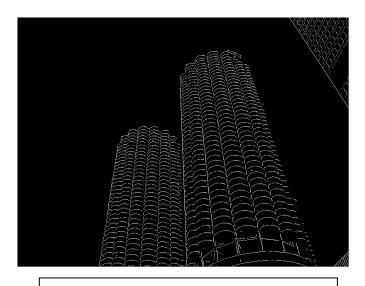
Global thresholding



Local thresholding



Hysteresis thresholding without non-Max suppression



Non-Max suppression, hysteresis threshold