



MIDTERM 1

ASSIGNMENT 6

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[HTTPS://GITHUB.COM/ELIAPICCOLI/ISPR-PROJECTS](https://github.com/eliapiccoli/ispr-projects)



EDGE DETECTION

EDGE DETECTION CODE

```
1 % read the image and take its gray representation
2 T = rgb2gray(imread('./dataset/2_13_s.bmp'));
3
4 % ----- Filtering
5 % Reduce possible noise in the image
6 % Trade-off between edge strength and noise reduction
7 GF = getfilter('gaussian');
8 TGF = convolution(T, GF);
9
10 % ----- Enhancement
11 % Determine changes in intensity in the neighbourhood of a point
12 % Compute the gradient magnitude, using the gradients along X-Y
13 SFX = getfilter('sobel', 'X');
14 SFY = getfilter('sobel', 'Y');
15 TSFX = convolution(T, SFX);
16 TSFY = convolution(T, SFY);
17 MTSF = uint8(sqrt(double((TSFX.^2)+(TSFY.^2))));
18
19 % ----- Detection
20 % Filter out only strong edges
21 RMTSF = rescale(MTSF,'InputMin',0,'InputMax',255);
22 threshold = 180/255;
23 TRMTSF = RMTSF;
24 TRMTSF(RMTSF < threshold) = 0;
```

GETFILTER CODE

```
1 function [F] = getfilter(varargin)
2 %GETFILTER Return the requested filter
3 % Give the name of the filter return the matrix of its representation.
4 % This version produces filters of default size [3x3].
5 % Certain parameters only apply to certain filters.
6 %Input:
7 % - FN : string, filter name
8 % - A : string, for gradient filters specify the axis (X, Y)
9 % - S : float, for gaussian filter specify standard deviation
10 % - N : integer, specify filter dimension
11 %Output:
12 % - F : NxN matrix representing the filter
13
14 % input checking
15 [FN, A, S, N] = parseparameters(varargin{:});
16 % Assign to F the correct filter
17 switch FN
18     case 'roberts'
19         if A == 'X'; F = [1 0 ; 0 -1]; else; F = [0 -1 ; 1 0]; end
20     case 'sobel'
21         if A == 'X'; F = [1 0 -1; 2 0 -2; 1 0 -1]; else; F = [1 2 1; 0 0 0; -1 -2 -1]; end
22     case 'prewitt'
23         if A == 'X'; F = [-1 0 1; -1 0 1; -1 0 1]; else; F = [1 1 1; 0 0 0; -1 -1 -1]; end
24     case 'average'
25         F = ones(N)/(N*N);
26     case 'gaussian'
27         r = N;c = N;
28         [row, col] = meshgrid(-(r-1)/2:(r-1)/2, -(c-1)/2:(c-1)/2);
29         F = exp(-(row.^2+col.^2)/(2*S^2));
30         F = F./sum(F(:));
31     case 'log' % Laplacian of Gaussian
32         % gaussian filter
33         r = N;c = N;
34         [row, col] = meshgrid(-(r-1)/2:(r-1)/2, -(c-1)/2:(c-1)/2);
35         F = exp(-(row.^2+col.^2)/(2*S^2));
36         F = F./sum(F(:));
37         % laplacian filter
38         F1 = F.*((row.*row + col.*col - 2*S^2)/(S^4));
39         F = F1 - sum(F1(:))/prod(r*c);
40 end
41 end
```

CONVOLUTION CODE

```
1 function [C] = convolution(I, K, P)
2 %CONVOLUTION Compute 2D-Convolution between an image I and kernel K
3 % Computes convolution between the Image and the Kernel. If provided the
4 % algorithm uses Matlab ParPool to compute the result.
5 %Input:
6 % - I : matrix [ N x M ], Image
7 % - K : matrix [ N x N ], Kernel
8 % - P : boolean, parallel computation
9 %Output:
10 % - C : Convolution Matrix
11
12 % cast I to int32 to avoid data type problems during computation
13 I = cast(I, 'int32');
14
15 % Create result matrix
16 [IR,IC] = size(I);
17 [KR,KC] = size(K);
18 C = zeros(1, IR*IC);
19
20 % Compute padding of original image (replicate)
21 Xpad = floor(KR/2);
22 Ypad = floor(KC/2);
23 upad = repmat(I(1, :), Xpad, 1);
24 bpad = repmat(I(end, :), Xpad, 1);
25 PI = [upad ; I ; bpad];
26 lpad = repmat(PI(:, 1), 1, Ypad);
27 rpad = repmat(PI(:, end), 1, Ypad);
28 PI = [lpad PI rpad];
29 [~, PIC] = size(PI);
```

```
31 % linearize the matrixes for better parallel computation
32 LPI = reshape(PI', 1, []);
33 LK = reshape(K', 1, []);
34 % reverse the kernel
35 LK = flip(LK);
36
37 % Compute convolution
38 if P
39     parfor index=0:IR*IC-1
40         C(index+1) = computeconv(index, IC, PIC, LPI, KR, KC, LK, Xpad, Ypad);
41     end
42 else
43     for index=0:IR*IC-1
44         C(index+1) = computeconv(index, IC, PIC, LPI, KR, KC, LK, Xpad, Ypad);
45     end
46 end
47
48 % un-linearize C
49 C = reshape(C, [IC, IR]);
```

```
1 function[sum] = computeconv(index, IC, PIC, LPI, KR, KC, LK, XP, YP)
2     x = floor(index/IC);
3     y = mod(index, IC);
4     sum = 0;
5     for i=0:KR-1
6         for j=0:KC-1
7             ImX = x + i - floor(KR/2);
8             ImY = y + j - floor(KC/2);
9             sum = sum + LK(i*KR+j+1)*LPI((ImX+XP)*PIC+ImY+YP+1);
10        end
11    end
12 end
```

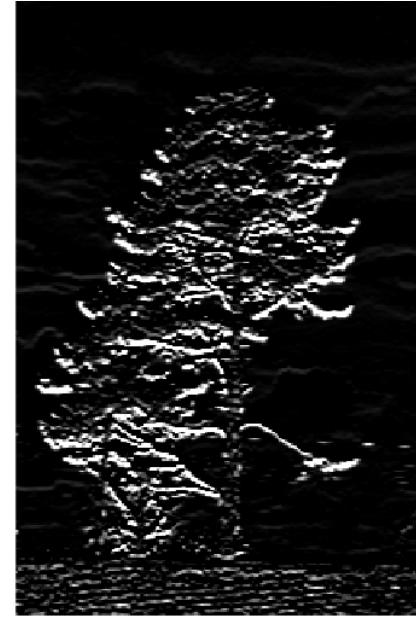
TREE



ORIGINAL



SOBEL GRADIENT X



SOBEL GRADIENT Y

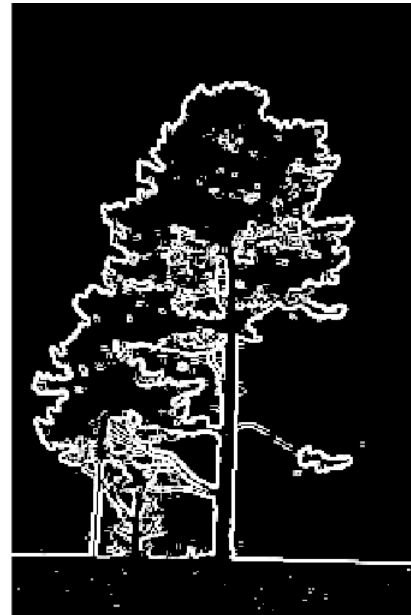


SOBEL GRADIENT MAGNITUDE

ROBERTS



PREWITT



SOBEL



FACE



ORIGINAL



SOBEL GRADIENT X



SOBEL GRADIENT Y



SOBEL GRADIENT MAGNITUDE



ROBERTS



PREWITT

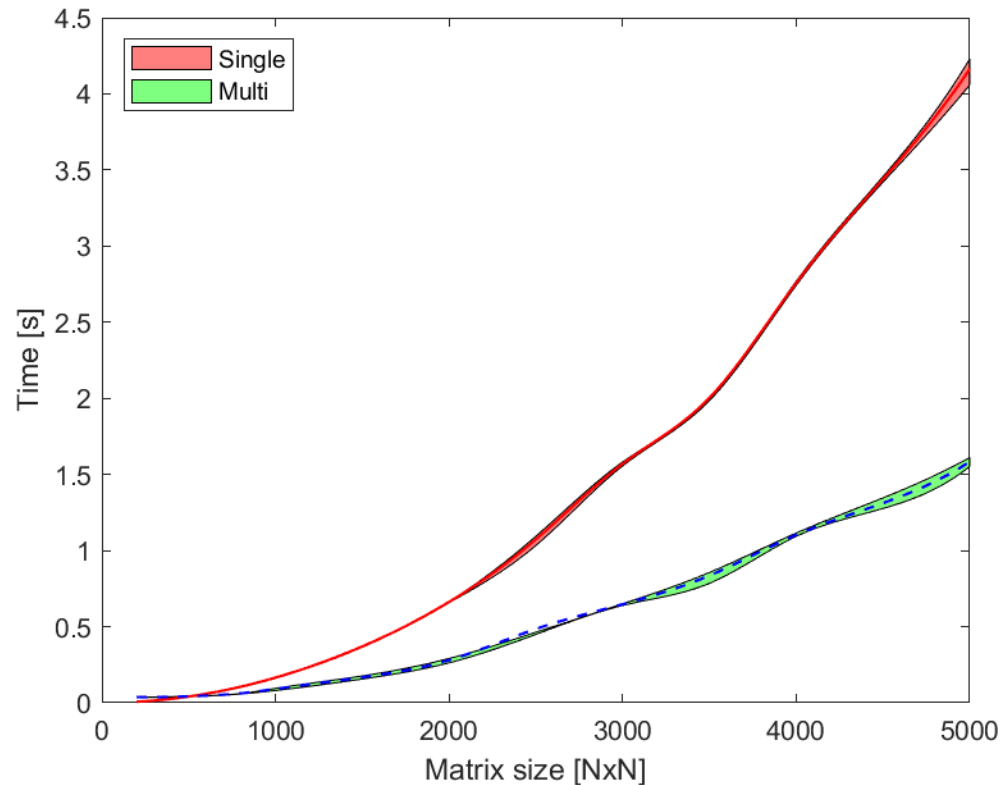


SOBEL

CAN WE IMPROVE?

COMPUTATIONAL POINT OF VIEW

Use parallel computation to compute convolution of different cells at the same time.



[Test on Intel® Core™ i9-9880H Processor – 8 workers]

MODEL POINT OF VIEW

We have seen only filters that compute the gradient, so they consider only the first derivative information.

What if, we try to exploit second derivative information?

Laplacian of Gaussian

