

Computer Vision I - Sheet 5

Group 2

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1 Gaussian Pyramids

The source code for this task can be found in the files `sh05ex01.m` and `gaussianPyramid.m`. To avoid opening Matlab only for viewing the code, it's included below. The results of this exercise are shown in the figures 1 and 2. Comparing the two pyramids, one can see a few differences, but in the overall comparison they look quite similar. Nevertheless there should be quite huge differences if one would compare the reconstructed images. Because of the missing gaussian filtering, the reconstructed image of the second pyramid should show a low of distortions.

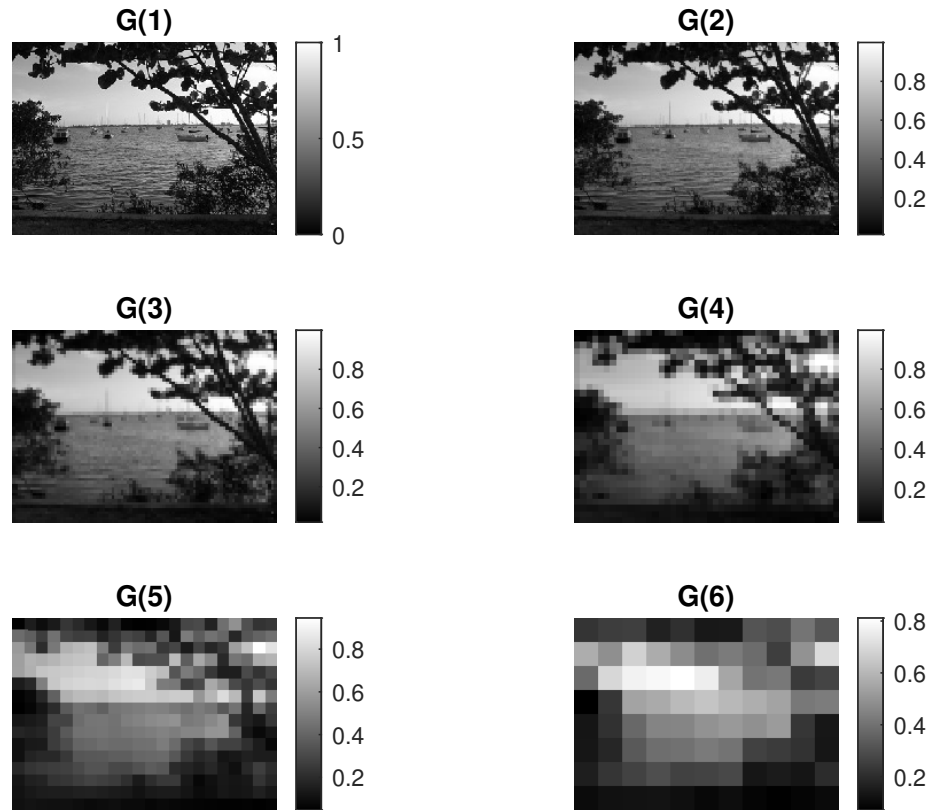


Figure 1: Gaussian pyramid with Gaussian filtering

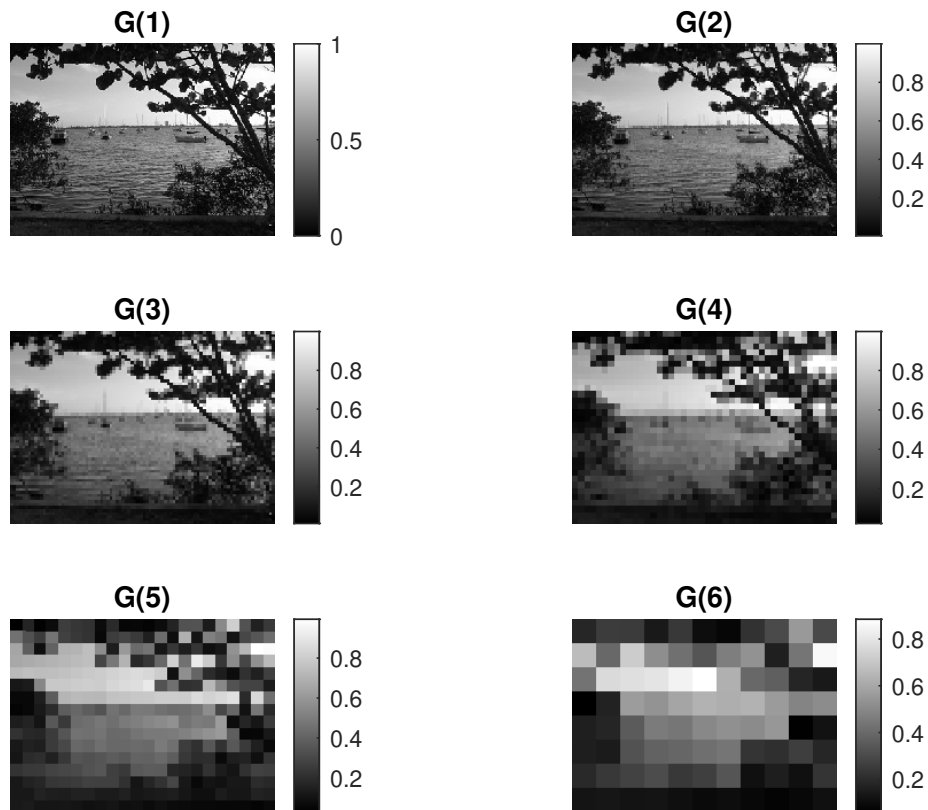


Figure 2: Gaussian pyramid without Gaussian filtering

```

1 image = im2double(imread("boats.tif"));
2
3 n=6;
4
5 images_gauss = gaussianPyramid(image, n, 1);
6 images = gaussianPyramid(image, n, 0);
7
8 %% Visualization
9 close all;
10
11 figure('Name','With gaussian filtering');
12 for i = 1:n
13     subplot(3,n/3,i);
14     imshow(images_gauss{i}, []);

```

```

15     colorbar();
16     title(['G(' num2str(i) ')']);
17 end
18 saveas(gcf, 'ex01_with.eps', 'eps')
19
20 figure('Name', 'Without gaussian filtering');
21 for i = 1:n
22     subplot(3,n/3,i);
23     imshow(images{i}, []);
24     colorbar();
25     title(['G(' num2str(i) ')']);
26 end
27 saveas(gcf, 'ex01_without.eps', 'eps')

1 function images = gaussianPyramid(image, n, useGaussian)
2     clear images;
3     images{1} = image;
4
5     % TODO: Parameter f r gauss bestimmen
6     gauss = fspecial('gaussian');
7
8     for i = 2:n
9         if useGaussian == 1
10             image = imfilter(image, gauss);
11         end
12         image = 0.5 * image(1:2:end, 1:2:end) + 0.5 * image
13             (2:2:end, 2:2:end);
14         images{i} = image;
15     end
16 end

```

2 Laplacian Pyramids

The source code for this task can be found in the files `sh05ex02.m` and `reconstruct.m` or below. Figure 3 shows the generated Laplacian pyramid, the corresponding reconstructed image is shown in figure 4. The figures 5 and 6 show the same Laplacian pyramid where compression was applied. Obviously the compression wasn't lossless, as the image now has quite huge artifacts in it. Figure 7 shows the resulting error of the compression for different threshold levels. As one would expect, the error raises with the threshold value, because more values are treated as zero.

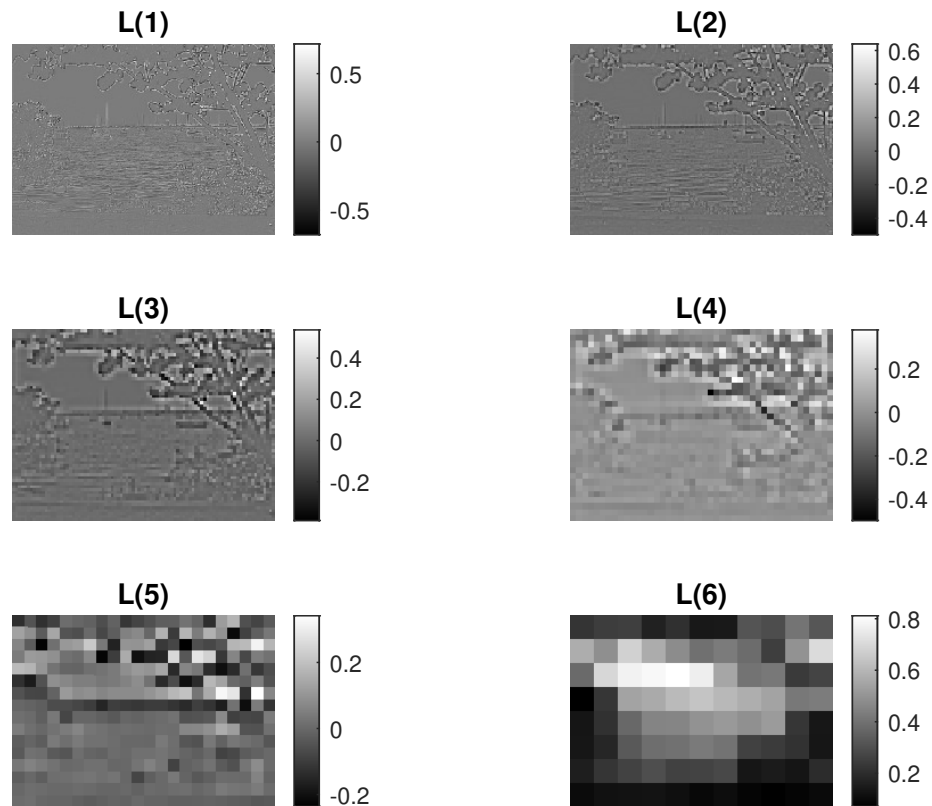


Figure 3: Laplacian pyramid



Figure 4: Reconstructed image

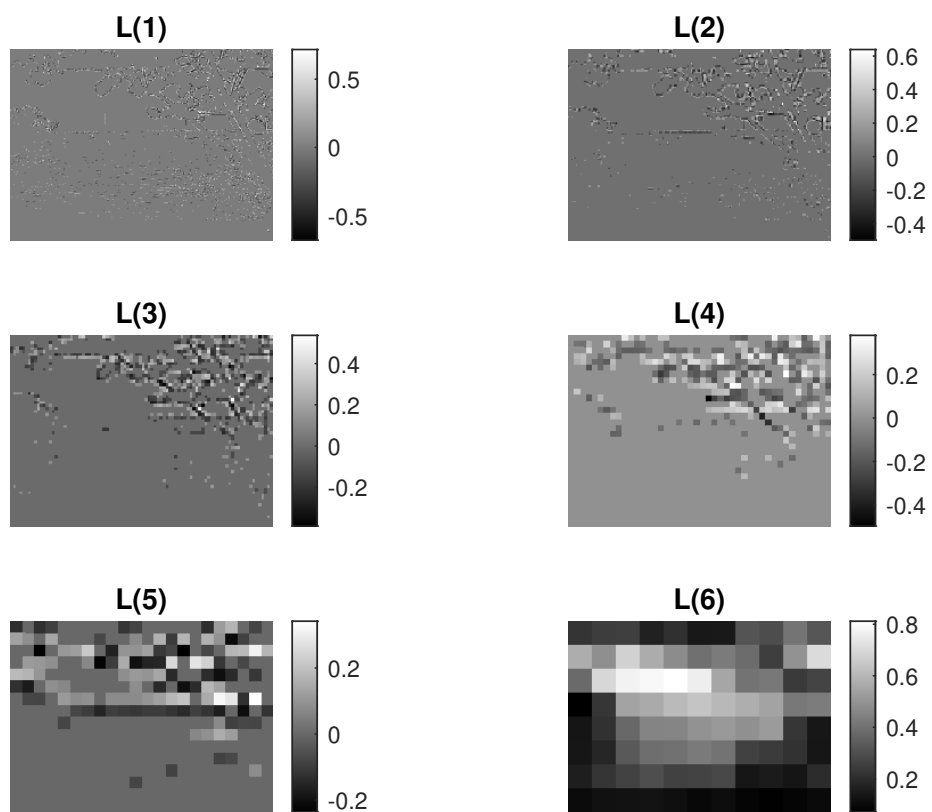


Figure 5: Laplacian pyramid with applied compression



Figure 6: Reconstructed image from the compressed pyramid

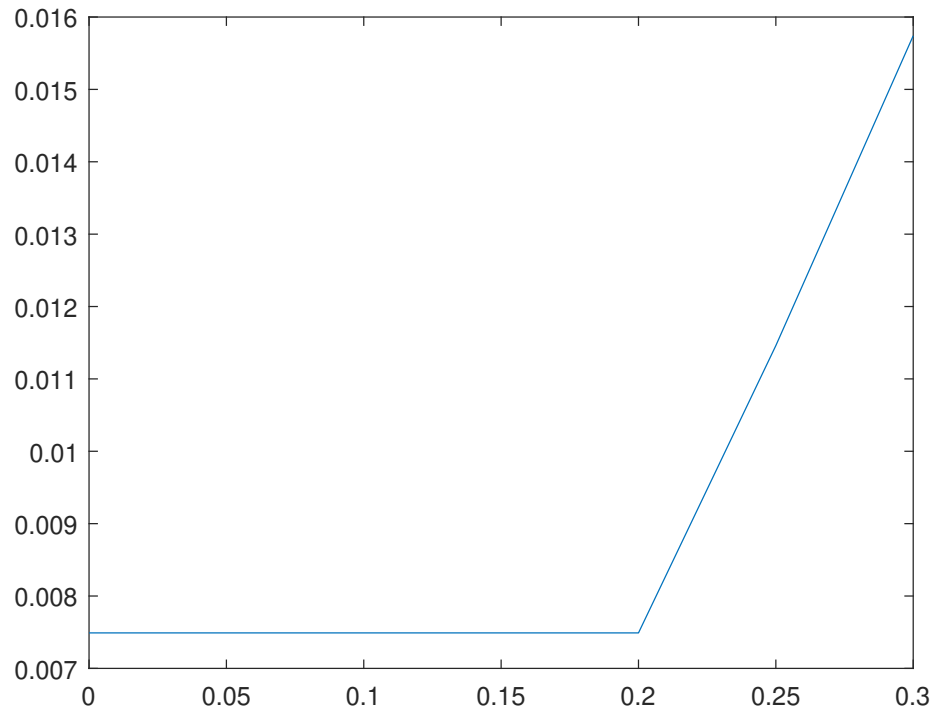


Figure 7: Error for different threshold values

```

1 image = im2double(imread('boats.tif'));
2 n = 6;
3 G = gaussianPyramid(image, n, 1);
4
5 % Calculate Laplacian Pyramid
6 for i = 1:n
7     % calc L{i}
8     if i ~= n
9         L{i} = G{i} - imresize(G{i+1}, 2);
10    else
11        L{i} = G{i};
12    end
13 end
14
15 %% Plot L
16 close all;
17 figure('Name', 'Laplacian Pyramid');

```

```

18 for i = 1:n
19     subplot(3,n/3,i);
20     imshow(L{i}, []);
21     colorbar();
22     title(['L(' num2str(i) ')']);
23 end
24 saveas(gcf, 'ex02_pyramid.eps', 'epsc')
25
26
27 % Reset G
28 clear G;
29
30 % Reconstruct image from L
31 figure('Name', 'Reconstructed Image from L');
32 imshow(reconstruct(L), []);
33 saveas(gcf, 'ex02_reconstructed.eps', 'epsc')
34
35 % Compression
36 lambda = 0.2;
37 for i = 1:n-1
38     L{i}(abs(L{i}) < lambda * max(max(abs(L{i})))) = 0;
39 end
40
41 %% Plot L
42 figure('Name', 'Laplacian Pyramid after Compression');
43 for i = 1:n
44     subplot(3,n/3,i);
45     imshow(L{i}, []);
46     colorbar();
47     title(['L(' num2str(i) ')']);
48 end
49 saveas(gcf, 'ex02_pyr_compress.eps', 'epsc')
50
51
52 % Reset G
53 clear G;
54
55 % Reconstruct image from L
56 figure('Name', 'Reconstructed Image from L after
57         Compression');
57 imshow(reconstruct(L), []);
58 saveas(gcf, 'ex02_rec_compress.eps', 'epsc')
59
60 lambda = 0:0.05:0.3
61 error = zeros(1, size(lambda,2));
62

```

```

63 for i = 1:size(lambda,2)
64     LT = L;
65     for k = 1:n-1
66         LT{k}(abs(LT{k}) < lambda(i) * max(max(abs(LT{k}))))
            = 0;
67     end
68     error(i) = immse(image, reconstruct(LT));
69 end
70
71 figure('Name', 'Error');
72 plot(lambda, error)
73 saveas(gcf, 'ex02_error.eps', 'epsc')

1 function image = reconstruct(L)
2     n = size(L, 2)
3
4     G{n} = L{n}
5     for i = n-1:-1:1
6         G{i} = imresize(G{i+1}, 2) + L{i};
7     end
8     image = G{1};
9 end

```

3 Gabor wavelets

The source code for this task can be found in the file `sh05ex03.m` or below. The Matlab script for the `getGabor` function was given in the exercise. Figure 8 shows the results of the energy calculations. One can easily stand, that the main orientations of the texture in the image are vertical and horizontal, because this orientations showed the biggest results.

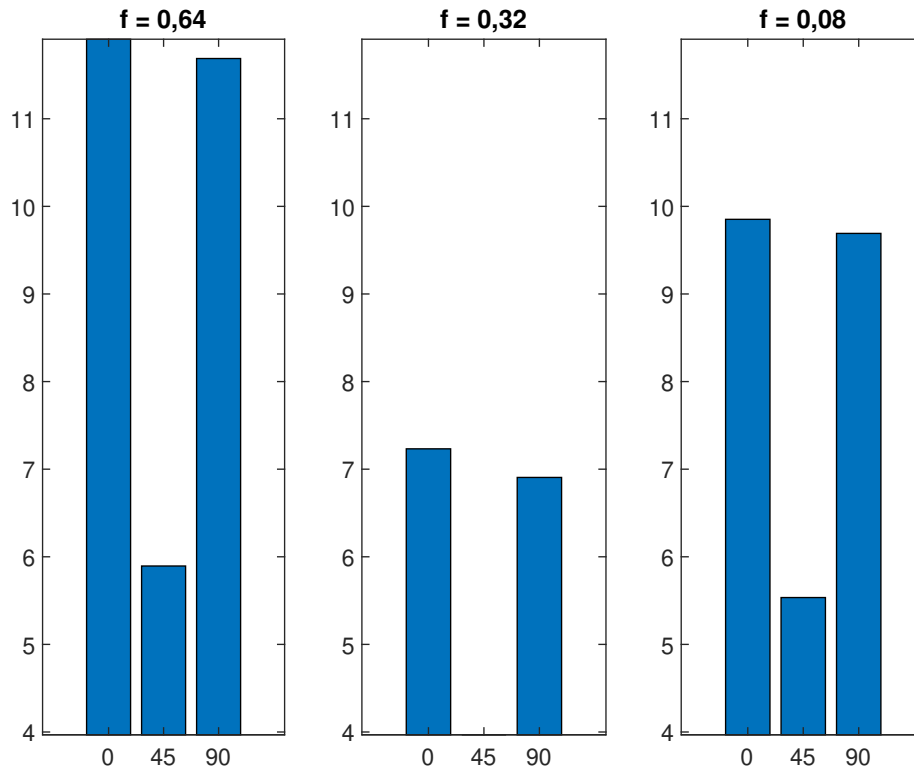


Figure 8: Filter energy of the different Gabor wavelets

```
1 clear; clc; close all;
2
3 image = imread('basket.jpg');
4
5 orientations = [0, pi/2, pi/4]*(180/pi);
6 frequencies = [0.64, 0.32, 0.08];
7
8 b = 2.32;
```

```

9
10 Gabors = {};
11
12 for f = frequencies
13     for theta = orientations
14         % b = f * sigma = const
15         Gabors{end+1} = getGabor(f, b/f, theta);
16     end
17 end
18
19 energy = zeros(1,length(Gabors));
20 for k = 1:length(Gabors)
21     energy(k) = sqrt(sum(sum(abs(imfilter(image, Gabors{k},
22         'conv').^2))));
23
24 close all;
25 figure();
26 axis auto;
27 subplot(1,3,1);
28 bar(orientations, energy(1:3));
29 title('f = 0,64');
30 ylim([min(energy), max(energy)]);
31
32 subplot(1,3,2);
33 bar(orientations, energy(4:6));
34 title('f = 0,32');
35 ylim([min(energy), max(energy)]);
36
37
38 subplot(1,3,3);
39 bar(orientations, energy(7:9));
40 title('f = 0,08');
41 ylim([min(energy), max(energy)]);
42 saveas(gcf, 'ex03.eps', 'epsc')

```