

EECS 442 Computer Vision Homework 04

Fan Bu
UMID: 68836435
Unique Name: fanbu

11/21/2016

Question 1, Harris Corner Detection

(a)

In this section, we choose $w = 3$ for the size of the window W .

Image result is shown in Figure 1:



(a) Result for I1.png



(b) Result for I3.png

Figure 1: Harris Corner Detection, $w = 3$

Key steps of the algorithm:

- Step1, Compute magnitude of the x and y gradients at each pixel.
- Step2, Due to Harris, construct a Gaussian window M around each pixel.
- Step3, Compute λ_s of second moment matrix M
- Step4, Compute $R = \det(M) - k(\text{tr}(M))^2$
- Step5, Set proper threshold T
- Step6, If $R > T$, a corner is detected; then, retain point of local maxima.
- Step7, Superpose corners on images.

More explanation is written in the attach code.

```

1 %% EECS 442 - HW 04 - Q1 Harris Corner Detection
2 %
3 % Date: 11 / 21 / 2016
4 % Author: Fan Bu
5 %
6 % ellipse(ra,rb,ang,x0,y0,C,Nb)
7 %
8 % Instructions(Latex code)
9 % Key steps of the algorithm:\\
10 % Step1, Compute magnitude of the x and y gradients at each pixel.\\
11 % Step2, Due to Harris, construct a Gaussian window M around each pixel.\\
12 % Step3, Compute  $\lambda_s$  of second moment matrix  $M$  \\
13 % Step4, Compute  $R = \det(M) - k(\text{tr}(M))^2$  \\
14 % Step5, Set proper threshold  $T$  \\
15 % Step6, If  $R > T$ , a corner is detected; then, retain point of local maxima. \\
16 % Step7, Superpose corners on images.\\
17
18 %% ----- Initialization -----
19 clear all;
20 close all;
21 clc;
22 %% ----- Load data -----
23 img1 = imread('I1.png');
24 img2 = imread('I3.png');
25
26 img = img2;
27 w = 3; %choose window size
28
29 %% ----- Compute M,R and ellipse parameter -----
30 % translate to gray image
31 gray_img = rgb2gray(img);
32 [height,width] = size(gray_img);
33 % initialize corner position
34 is_edge = zeros(height, width);
35 % initialize R score
36 R = zeros(height, width);
37
38 % applying sobel edge detector in the horizontal direction
39 fx = [-1 0 1;
40         -1 0 1;
41         -1 0 1];
42 Ix = filter2(fx,gray_img);
43 % applying sobel edge detector in the vertical direction
44 fy = [1 1 1;
45         0 0 0;
46         -1 -1 -1];
47 Iy = filter2(fy,gray_img);
48 % compute terms for second moment matrix M
49 Ix2 = Ix.^2;
50 Iy2 = Iy.^2;
51 Ixy = Ix.*Iy;
52
53 % perform gaussian filtering for eliminating noises
54 h= fspecial('gaussian',[w w],2);
55 Ix2 = filter2(h,Ix2);
56 Iy2 = filter2(h,Iy2);
57 Ixy = filter2(h,Ixy);
58
59 % Initialize Perem
60 Rmax = 0;
61 figure;
62 imshow(img);
63 hold on;
64 for i = 1:height

```

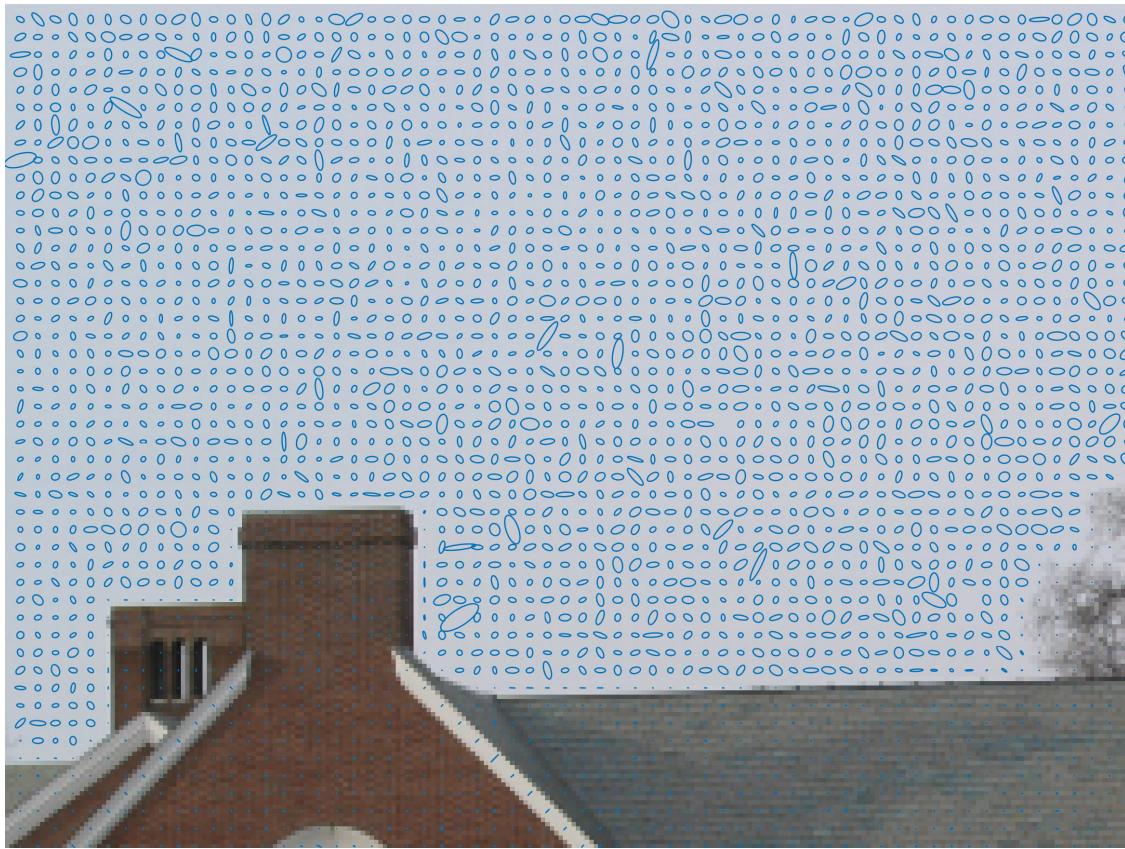
```

65     for j = 1:width
66         % compute second moment matrix M
67         M = [
68             Ix2(i,j), Ixy(i,j);
69             Ixy(i,j) Iy2(i,j)
70         ];
71         % compute R(i,j) = det(M)-0.04*(trace(M))^2
72         R(i,j) = det(M)-0.04*(trace(M))^2;
73         if R(i,j) > Rmax
74             Rmax = R(i,j);
75         end;
76
77     end;
78 end;
79 num_edge = 0;
80 Rmax
81 T=0.1*Rmax
82 % determine corner by comparing with threshold and all neighbor points
83 for i = 2:height-1
84     for j = 2:width-1
85         if R(i,j) > T ...
86             && R(i,j) > R(i-1,j-1) ...
87             && R(i,j) > R(i-1,j) ...
88             && R(i,j) > R(i-1,j+1) ...
89             && R(i,j) > R(i,j-1) ...
90             && R(i,j) > R(i,j+1) ...
91             && R(i,j) > R(i+1,j-1) ...
92             && R(i,j) > R(i+1,j) ...
93             && R(i,j) > R(i+1,j+1)
94
95             is_edge(i,j) = 1;
96             num_edge = num_edge+1;
97         end;
98     end;
99 end;
100 [edge_col, edge_row] = find(is_edge == 1);
101 num_edge
102 %% ----- Draw corners on image and save -----
103 plot(edge_row, edge_col,'r*','linewidth',1.1);
104 print(gcf,'-djpeg','HW4_q1_a_I3.jpeg','-r400')

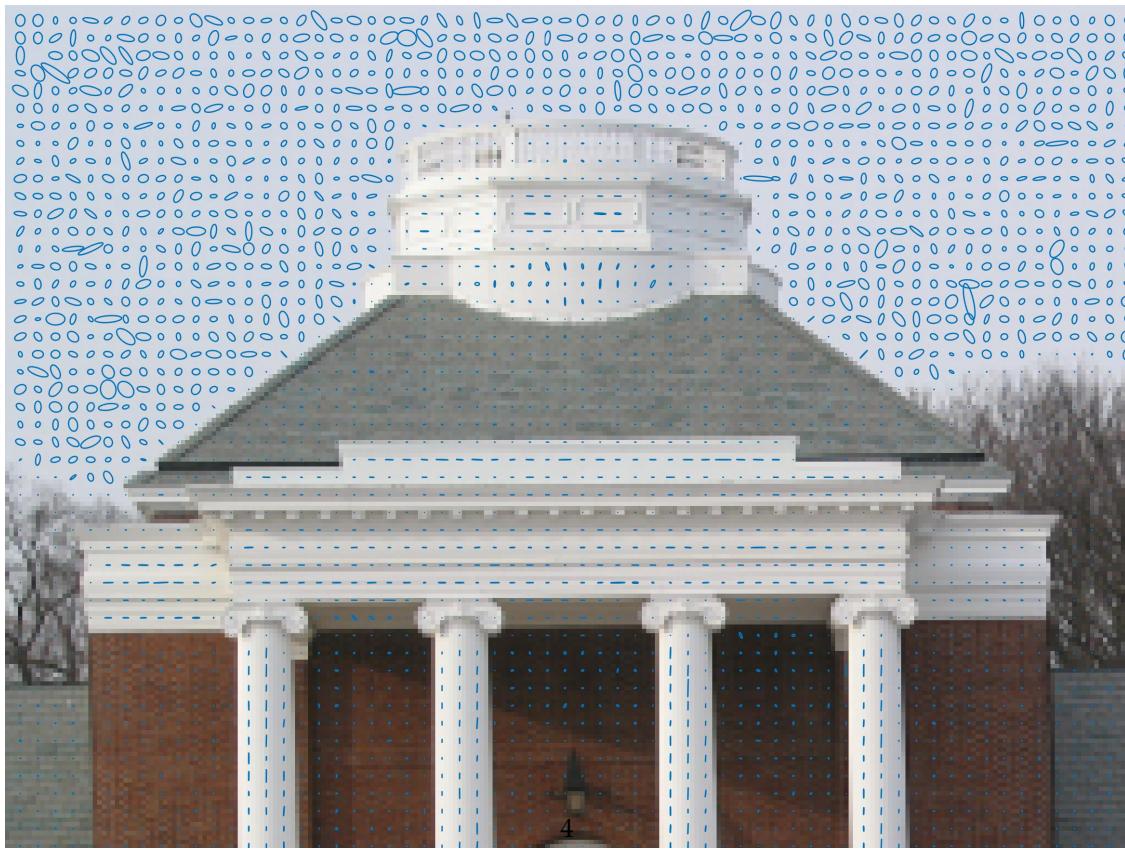
```

(b)

Repeat this step by using different sizes of the window $W = 3, 4, 6, 9$. Results are shown in the following figures (Note that pictures are zoomed in):

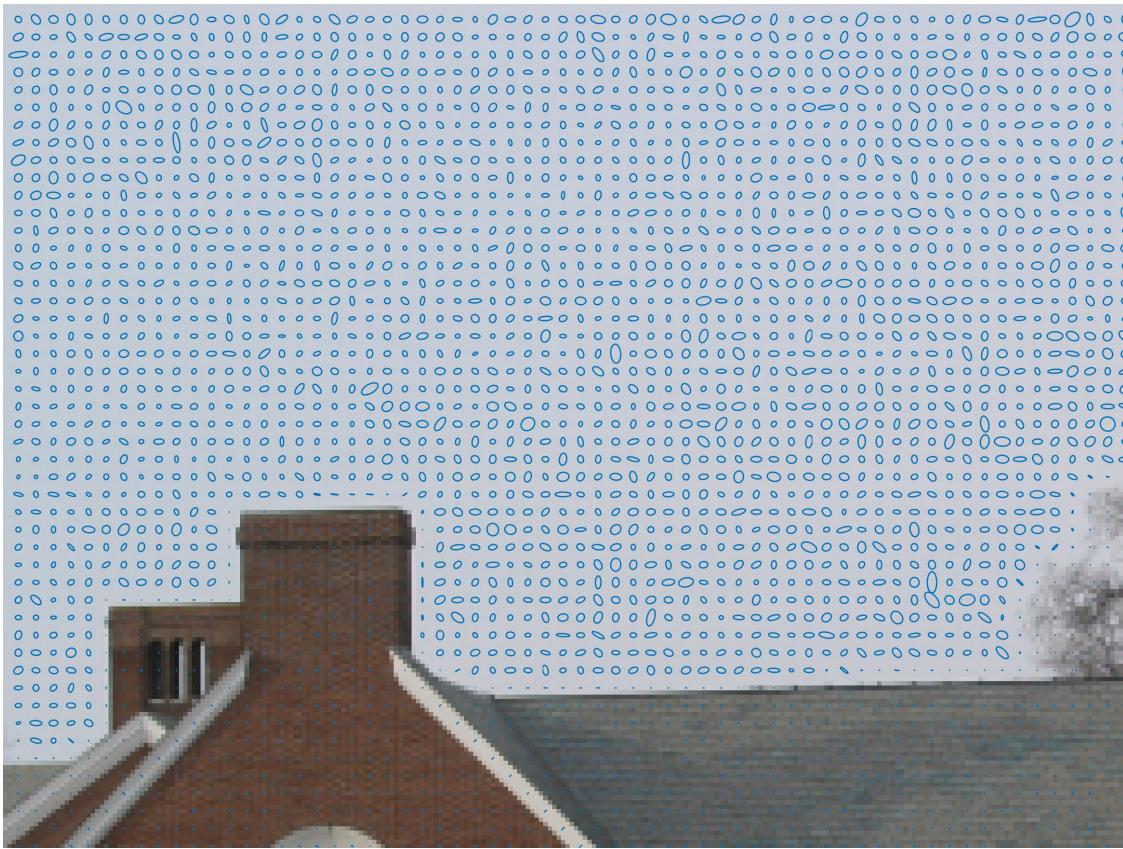


(a) Result for I1.png

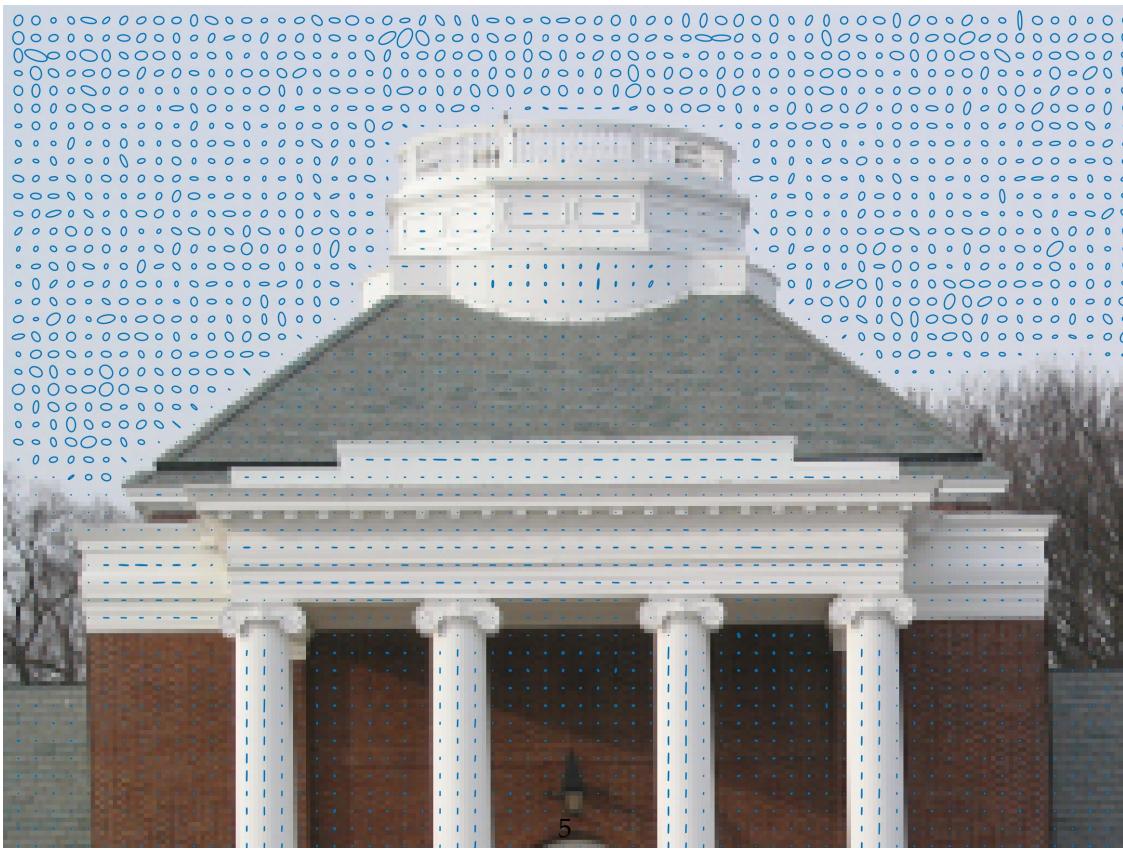


(b) Result for I3.png

Figure 2: Superimpose Ellipses, $w = 3$

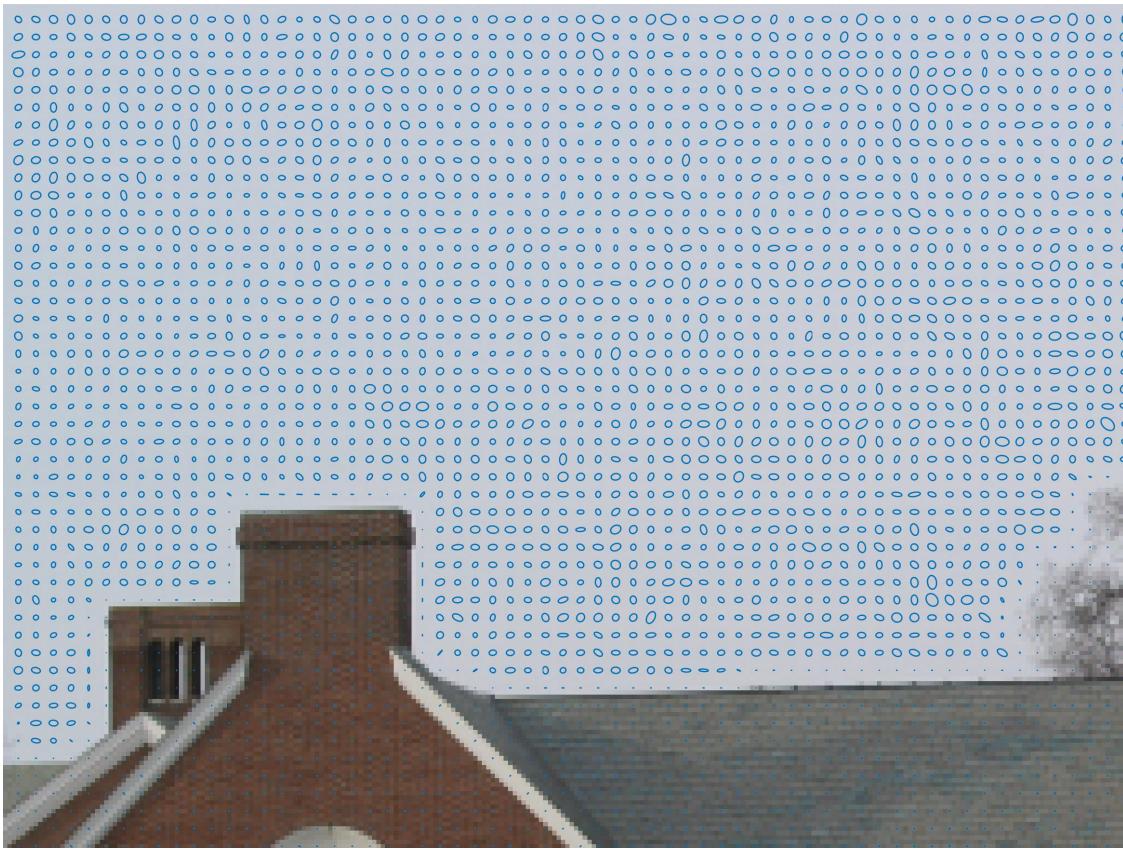


(a) Result for I1.png

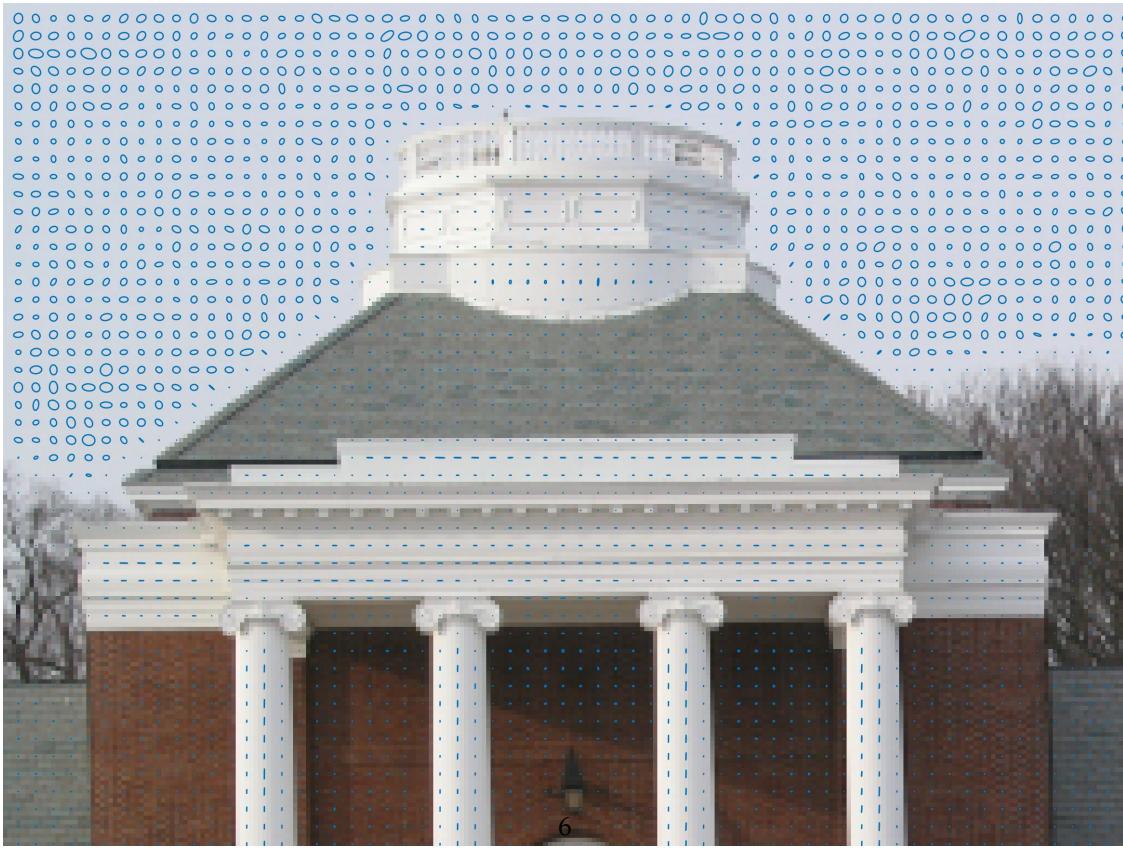


(b) Result for I3.png

Figure 3: Superimpose Ellipses, $w = 4$

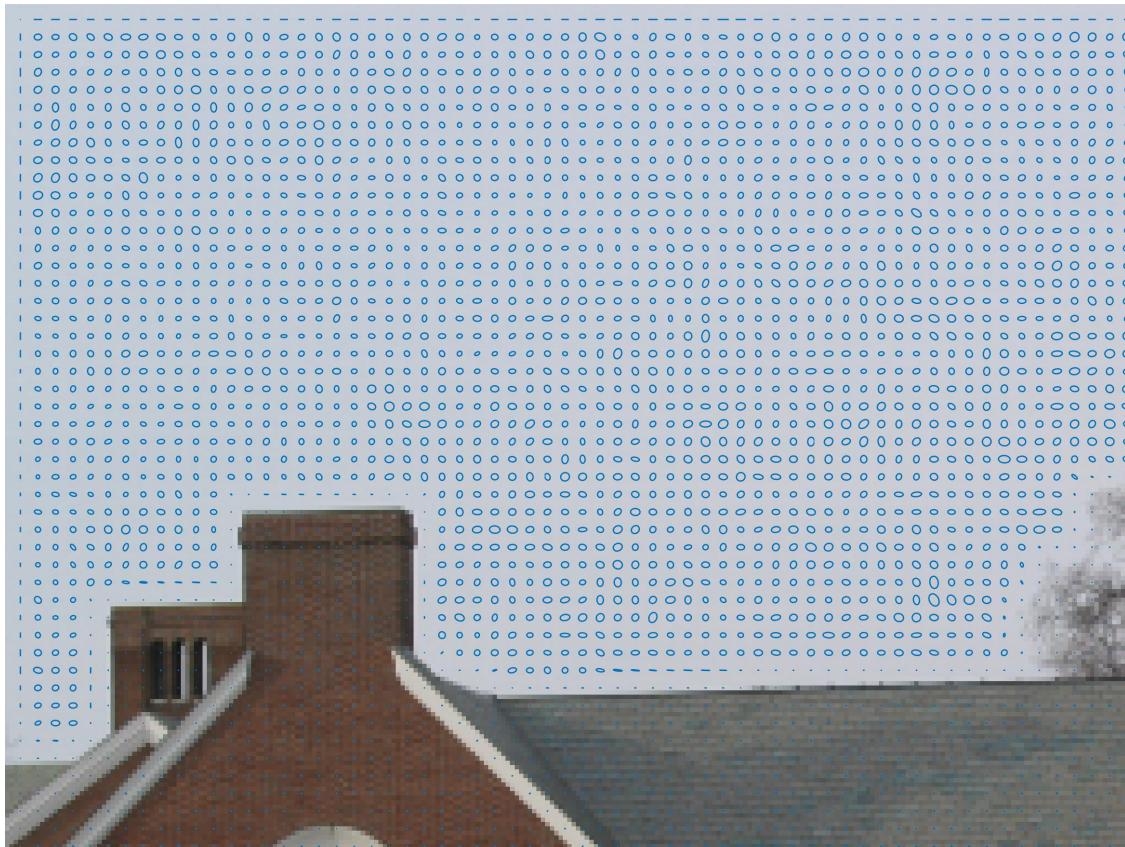


(a) Result for I1.png

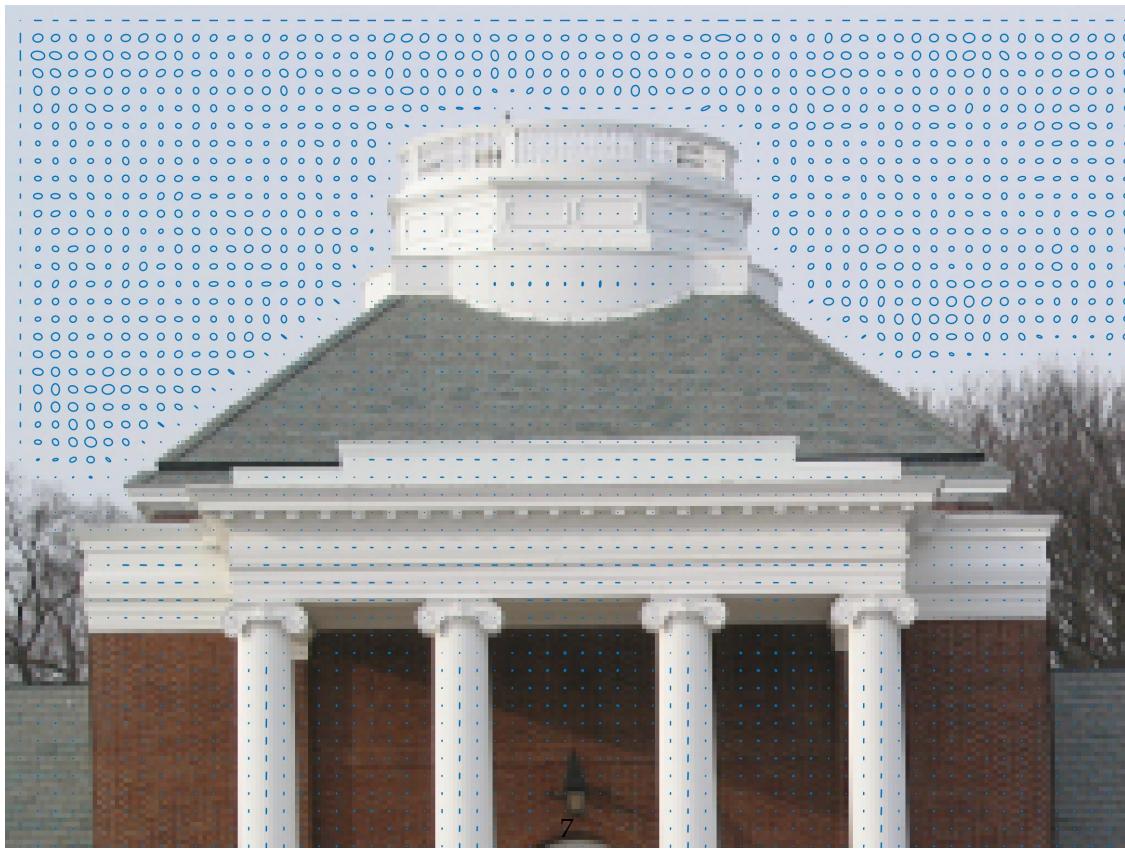


(b) Result for I3.png

Figure 4: Superimpose Ellipses, $w = 6$



(a) Result for I1.png



(b) Result for I3.png

Figure 5: Superimpose Ellipses, $w = 9$

From figures above, we can see that while $w = 3$ gives ellipses with the most distinguishable shapes, and $w = 4$ gives reasonable result. However, the ellipses' changes are too small to be noticed when $w = 6$ and $w = 9$.

In conclusion, as window size becomes larger, the ellipses change would be more difficult to be observed.

Codes are attached below:

```

1  %% EECS 442 - HW 04 - Q1 Harris Corner Detection
2  %
3  % Date: 11 / 21 / 2016
4  % Author: Fan Bu
5  %
6  % ellipse(ra,rb,ang,x0,y0,C,Nb)
7  %
8  % Instructions(Latex code)
9  % Key steps of the algorithm:\\
10 % Step1, Compute magnitude of the x and y gradients at each pixel.\\\
11 % Step2, Due to Harris, construct a Gaussian window M around each pixel.\\\
12 % Step3, Compute $\lambda_s$ of second moment matrix $M$ \\
13 % Step4, Compute $R=\det(M)-k(\text{tr}(M))^2$ \\
14 % Step5, Set proper threshold $T$ \\
15 % Step6, If $R> T$, a corner is detected; then, retain point of local maxima. \\
16 % Step7, Superpose corners on images.\\\
17
18 %% Initialization
19 clear all;
20 close all;
21 clc;
22 %% ----- Load data -----
23 img1 = imread('I1.png');
24 img2 = imread('I3.png');
25
26 img = img2;
27 w = 3; %choose window size
28
29 %% ----- Part 1: compute M,R and ellipse parameter -----
30 % translate to gray image
31 gray_img = rgb2gray(img);
32 [height,width] = size(gray_img);
33 % initialize corner position
34 is_edge = zeros(height, width);
35 % initialize R score
36 R = zeros(height, width);
37
38 % applying sobel edge detector in the horizontal direction
39 fx = [-1 0 1;
40         -1 0 1;
41         -1 0 1];
42 Ix = filter2(fx,gray_img);
43 % applying sobel edge detector in the vertical direction
44 fy = [1 1 1;
45         0 0 0;
46         -1 -1 -1];
47 Iy = filter2(fy,gray_img);
48 % compute terms for second moment matrix M
49 Ix2 = Ix.^2;
50 Iy2 = Iy.^2;
51 Ixy = Ix.*Iy;
52
53 % perform gaussian filtering for eliminating noises

```

```

54 h= fspecial('gaussian',[w w],2);
55 Ix2 = filter2(h,Ix2);
56 Iy2 = filter2(h,Iy2);
57 Ixy = filter2(h,Ixy);
58
59 % Initialize ellipsePerem
60 ellipsePerem = [];
61 Rmax = 0;
62 figure;
63 imshow(img(1:height/2,1:width/2,:),'border','tight','initialmagnification','fit');
64 hold on;
65 for i = 1:height
66     for j = 1:width
67         % compute second moment matrix M
68         M = [
69             Ix2(i,j), Ixy(i,j);
70             Ixy(i,j) Iy2(i,j)
71         ];
72         % compute R(i,j) = det(M)-0.04*(trace(M))^2
73         R(i,j) = det(M)-0.04*(trace(M))^2;
74         if R(i,j) > Rmax
75             Rmax = R(i,j);
76         end;
77
78     end;
79 end;
80 num_edge = 0;
81 Rmax
82 T=0.1*Rmax
83 % determine corner by comparing with threshold and all neighbor points
84 for i = 2:height-1
85     for j = 2:width-1
86         M = [
87             Ix2(i,j), Ixy(i,j);
88             Ixy(i,j) Iy2(i,j)
89         ];
90         % compute M's eigenvalue
91         [vector,lambda] = eig(M);
92         lambda = diag(lambda);
93         % compute radius associate to the ellipse
94         [lambda1,index] = max(lambda);
95         lambda2 = min(lambda);
96         a = lambda1^(-0.5);
97         b = lambda2^(-0.5);
98         vector = vector(:,index);
99         angle = atan(vector(2)/vector(1));
100        % form an ellipsePerem
101        if mod(i,4) == 0 && mod(j,4) == 0
102            ellipsePerem = [ellipsePerem;a b angle j i];
103        end
104        if R(i,j) > T ...
105            && R(i,j) > R(i-1,j-1) ...
106            && R(i,j) > R(i-1,j) ...
107            && R(i,j) > R(i-1,j+1) ...
108            && R(i,j) > R(i,j-1) ...
109            && R(i,j) > R(i,j+1) ...
110            && R(i,j) > R(i+1,j-1) ...
111            && R(i,j) > R(i+1,j) ...
112            && R(i,j) > R(i+1,j+1)
113
114            is_edge(i,j) = 1;
115            num_edge = num_edge+1;
116        end;
117    end;
118 end;

```

```

119 [edge_col, edge_row] = find(is_edge == 1);
120 num_edge
121 %% ===== Part 2: Plot ellipse and corner on image =====
122
123 ellipse(ellipsePerem(:,1), ellipsePerem(:,2), ellipsePerem(:,3), ...
124     ellipsePerem(:,4), ellipsePerem(:,5));
125 print(gcf,'-djpeg', ['HW4_q1_b_w', num2str(w), '_I3.jpeg'], '-r400')

```

(c)

The thresholds are chosen as $T = t \cdot R_{max}$, where R_{max} is defined as the maximum corner response and takes value of 9.4191×10^9 for I1.png, and 1.0741×10^{10} for I3.png.

Take $t = 0.001, 0.01, 0.05, 0.1$, and the results are shown in figures 6,7,8,9:

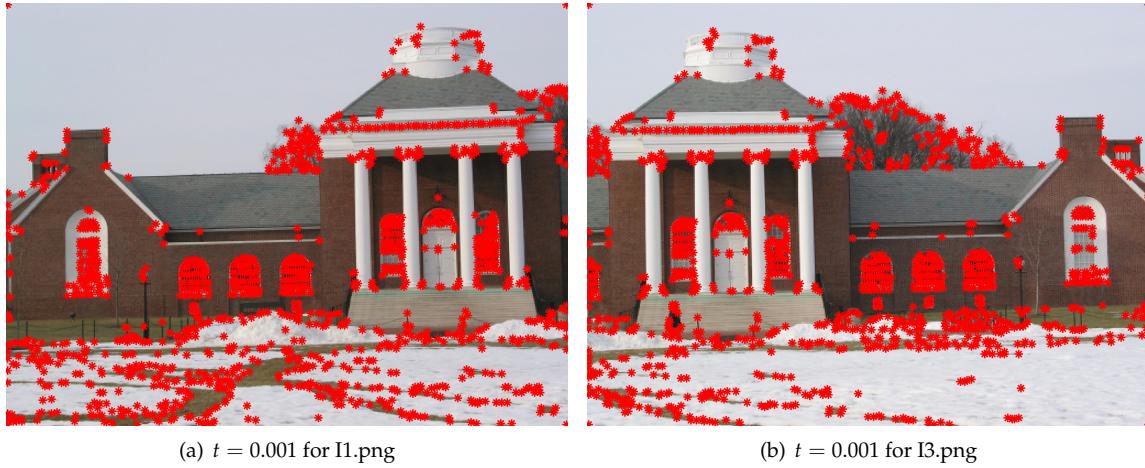


Figure 6: Harris Corner Detection results with different threshold, $T = tR_{max}, t = 0.001$

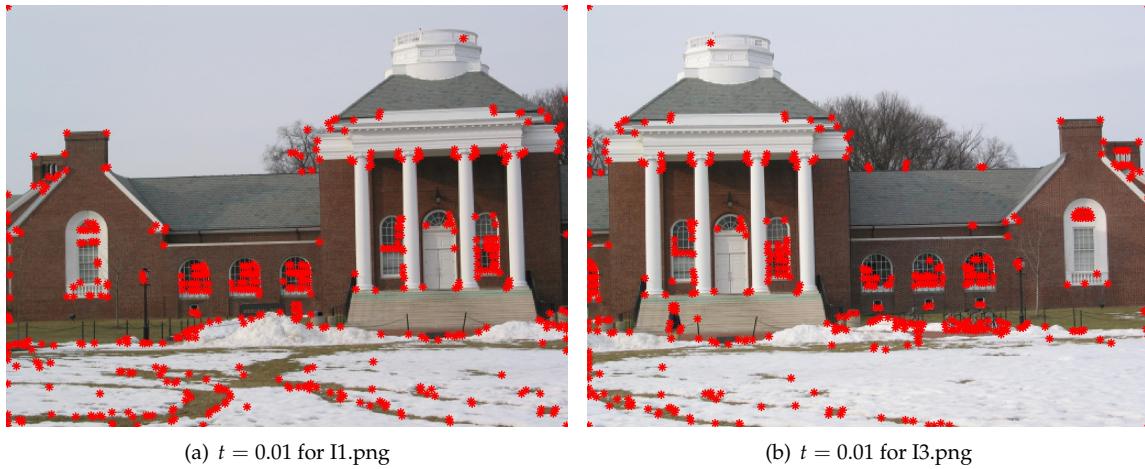


Figure 7: Harris Corner Detection results with different threshold, $T = tR_{max}, t = 0.01$

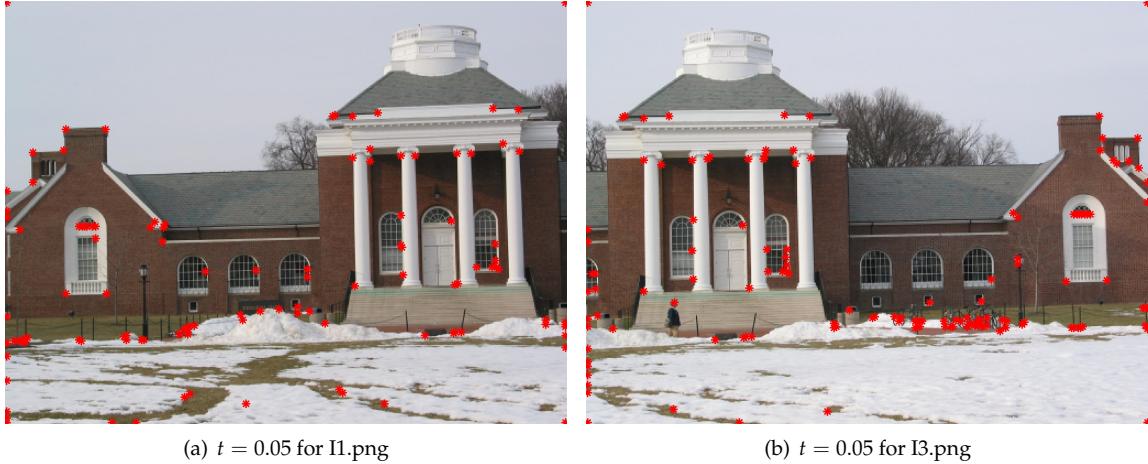


Figure 8: Harris Corner Detection results with different threshold, $T = tR_{max}$, $t = 0.05$

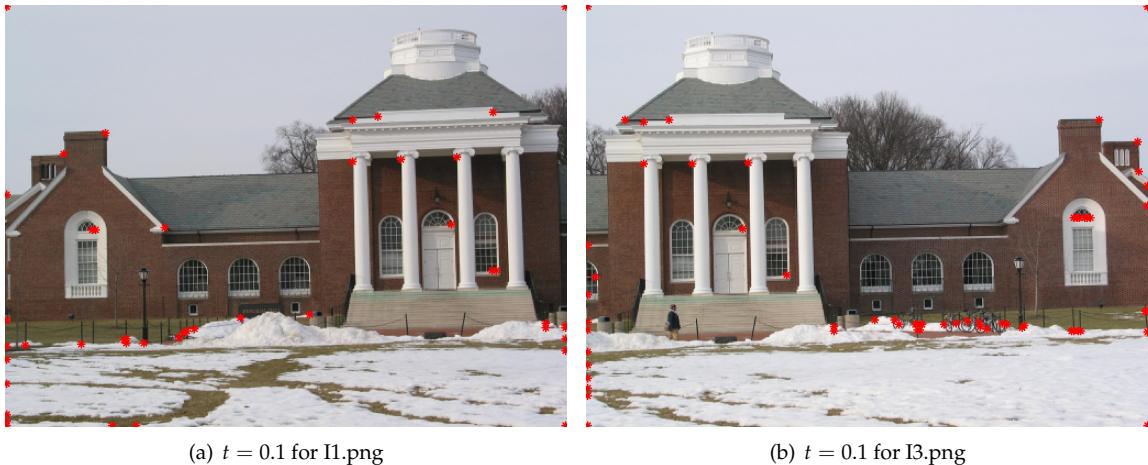


Figure 9: Harris Corner Detection results with different threshold, $T = tR_{max}$, $t = 0.1$

By comparing these Figures, we can see that a reasonable result is shown when $T = 0.01R_{max}$.

From Figure 6, more edge points were detected than the actual corner points when $T = 0.001R_{max}$, indicating this threshold is too low.

Same reasoning, when $T = 0.1R_{max}$ and $T = 0.05R_{max}$, detection results tend to lose some corner points, indicating these thresholds are too high.

In conclusion, $t = 0.01$ may be an ideal threshold for $T = t \cdot R_{max}$.

Question 2, Blob Detection

(a)

(1)

Algorithm outline:

1. Build a Laplacian scale space, starting with some initial scale and going for n iterations.
2. Generate a scale-normalized Laplacian of Gaussian filter at a given scale "sigma".
3. Filter image with the scale-normalized Laplacian.
4. Save square of Laplacian filter response for current level of scale space.
5. Determine threshold by multiply the scale with a factor k .
6. Perform non-maximum suppression in scale space.
7. Display resulting circles at their characteristic scales.

Codes are attached at the bottom of this question.

(2)

The safety factor k should depend on the largest scale at which we want regions to be detected. Therefore, the thresholds are chosen as $T = k \cdot \text{scaleSpace}_{\max}$, where scaleSpace_{\max} is defined as the maximum scale space representation and takes value of 1.1669×10^4 . Take $k = 0.1, 0.2, 0.3, 0.4$, and results are shown in Figure 10,11,12,13. The exact threshold T can be seen on the top of the figures.



Figure 10: Harris Corner Detection results with $k = 0.1, T \approx 1166$



Figure 11: Harris Corner Detection results with $k = 0.2$, $T \approx 2333$



Figure 12: Harris Corner Detection results with $k = 0.3$, $T \approx 3500$



Figure 13: Harris Corner Detection results with $k = 0.4$, $T \approx 4667$

From the figures above, we can see that while $k = 0.2$ captures a reasonable result. When $k = 0.3$, less blobs are detected but still acceptable.

While $k = 0.1$ includes some dummy points in the picture, for example those blobs in the sky, indicating this threshold is too small. Same reasoning, when $k = 0.4$, the detection only capture very limited nearby blobs. In that case, the threshold is too high.

(b)

The detection result of $T = 1500$ is shown below in Figure 14:



Figure 14: Harris Corner Detection results with $k = 0.12855$, $T = 1500$

The histogram of normalized radii is shown in Figure 15, and maximum radius is also shown on the title.

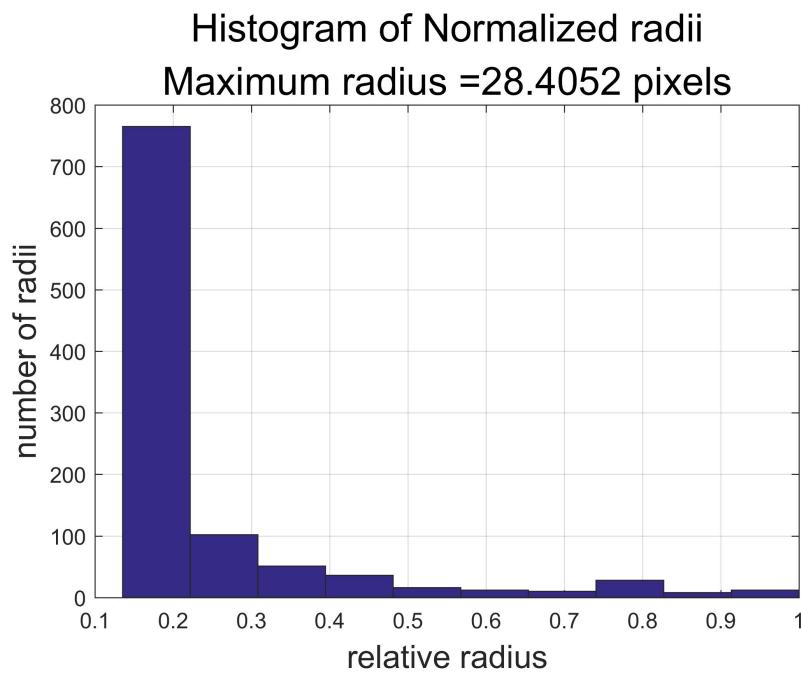


Figure 15: Histogram of Normalized radii with $T = 1500$, $bin = 10$

Matlab Codes

```
1 %% EECS 442 - HW 04 - Q2 Blob Detection
2 % -----
3 % Date: 11 / 21 / 2016
4 % Author: Fan Bu
5 %
6 % show_all_circles(I, cx, cy, rad, color, ln_wid)
7 %
8 % Instructions:
9 % Implement a blob detector based on the Laplacian operator and
10 % characteristic scale concept introduced by Lindeberg. Compute the radius
11 % of each blob (in pixels) and report the histogram (dis- tribution)
12 % of radii computed for the whole image.
13 %
14 % Algorithm outline:\\
15 % 1. Build a Laplacian scale space, starting with some initial scale and going for $n$ ...
16 %    iterations.\\
17 % 2. Generate a scale-normalized Laplacian of Gaussian filter at a given scale "$sigma$".\\
18 % 3. Filter image with the scale-normalized Laplacian. \\
19 % 4. Save square of Laplacian filter response for current level of scale space. \\
20 % 5. Determine threshold by multiply the scale with a factor $k$. \\
21 % 6. Perform non-maximum suppression in scale space. \\
22 % 7. Display resulting circles at their characteristic scales.\\
23 %
24 %% Initialization
25 clear all;
26 close all;
27 clc;
28 %% ----- Parameter selection -----
29 t = 1:0.1:3;
30 sigma = exp(t);
31 sigma = sigma';
32 levels = length(sigma);
33 %% ----- Load data -----
34 img = imread('3.bmp');
35 img = rgb2gray(img);
36 double_img = double(img);
37 %% Build a Laplacian scale space, starting with some initial scale and going for $n$ ...
38 %    iterations.
39 [height width] = size(img);
40 scale_space = zeros(height,width,levels);
41 %% ----- Perform detection -----
42
43 for i=1:levels
44
45 % Generate a scale-normalized Laplacian of Gaussian filter at a given scale "$sigma$".
46 filter_size = 2*ceil(3*sigma(i))+1;
47 kernel = fspecial('log', filter_size, sigma(i));
48 % Filter image with the scale-normalized Laplacian.
49 nLoG = sigma(i)^2 * kernel;
50 Filtered_img = imfilter(double_img, nLoG, 'same', 'replicate');
51 Filtered_img = Filtered_img .^ 2;
52 % Save square of Laplacian filter response for current level of scale space.
53 scale_space(:,:,:i) = Filtered_img;
54 end
55 % choose the right thresold T, as T = k * scaleSpace_{max}.
56 k = 0.4;
57 max_scale_space = max(max(max(scale_space)))
58 T = k * max_scale_space
59 % choose the right thresold T =1500 for this question.
```

```

60 T = 1500
61 k = T/max_scale_space
62 % Perform non-maximum suppression in the 3D scale space
63 scale_space_filtered = zeros(height,width,levels);
64 for i=1:levels
65     scale_space_filtered(:,:,:,i) = ordfilt2(scale_space(:,:,:,i),9,ones(3));
66 end
67 for i = 1:levels
68     scale_space_filtered(:,:,:,i) = ...
69         max(scale_space_filtered(:,:,:,:max(i-1,1):min(i+1,levels)),[],3);
70 end
71 scale_space_filtered = scale_space_filtered .* (scale_space_filtered == scale_space);
72 % calculate parameters for show_all_circles
73 row = [];
74 col = [];
75 radius = [];
76 for i=1:levels
77     [new_rows,new_cols] = find(scale_space_filtered(:,:,:,i)≥T);
78     num = length(new_rows);
79     radii = sigma(i)*sqrt(2);
80     radii = repmat(radii,num,1);
81     row = [row;new_rows];
82     col = [col;new_cols];
83     radius = [radius;radii];
84 end
85 show_all_circles(img,col,row,radius);
86 set(gcf,'units','normalized','position',[0,0,0.8,1]);
87 h_title=title({'Number of circles = ', num2str(size(col,1)) ...
88     ['Safety factor k = ',num2str(k),'; Threshold T = ',num2str(T)]});
89 set(h_title,'FontSize',18);
90 print(gcf,'-djpeg',[ 'HW4_q2_c_k_',num2str(k),'.jpeg'], '-r400')
91
92 % recompute 3d-indices of maxima
93 s = [height width levels];
94 %% ----- draw blobs on image -----
95 figure
96 max_rad = max(radius);
97 radius = radius / max_rad;
98 hist(radius,10);
99 h_title=title({'Histogram of Normalized radii' ...
100     ['Maximum radius =', num2str(max_rad), ' pixels']});
101 set(h_title,'FontSize',18);
102 grid on
103 xlabel('relative radius','FontSize',15)
104 ylabel('number of radii','FontSize',15)
105 print(gcf,'-djpeg',[ 'HW4_q2_c_hist.jpeg'], '-r400')

```