

CS-867 Spring 2021

Computer Vision

Assignment # 01

Harris Corner Detector & Its robustness to
Rotation & Scaling

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Harris Keypoint Detector and its Robustness to Rotation and Scaling

PART- A

Implement the Harris keypoint detector as discussed in the lectures. You do not need to implement a descriptor. Apply it to the two images provided and threshold the cornerness response so that about a few hundred keypoints are returned. Submit a result showing the (top-scoring) detected Harris keypoints superimposed on the original image (e.g., you can place a cross or a circle at the location of the keypoint; do this on the grayscale version to facilitate visualization). Report the threshold that you choose. Describe which objects or regions in the image seem to generate large numbers of Harris keypoints, and why. You may **not** use the built-in library function for Harris detection, but must implement it on your own.

Programming platform: MATLAB

The following built-in functions are used for computations:

- `im1 = imread('famous_five.png')` for image reading
- `im = rgb2gray(image)` for converting color image to grayscale
- `ix = conv2(double(im),dx,'same')` for horizontal derivative
- `iy = conv2(double(im),dy,'same')` for vertical derivative
- PARAMETERS FOR GAUSSIAN FILTER
 - `+sigma = 3`
 - `+order = 16 * 16`
- `Ix2 = conv2(double(ix.^2),G,'same')` for Ix^2
- `Iy2 = conv2(double(iy.^2),G,'same')` for Iy^2
- `Ixy = conv2(double(ix.*iy),G,'same')` for $Ix*Iy$
- Threshold =4000
- `maximum_point = ordfilt2(r, order^2,ones(order))` filter for non-max suppression
- Order of NMS filter = $32 * 32$
- `harris_corners = (r==maximum_point) & (r>threshold)` for finding Corners

Note: I resized the given images to a standard size i.e. 640 by 480 in a photo editor before using them in MATLAB, just to avoid any ambiguity and to focus on the main task.

Code

```
1 % CS-867 COMPUTER VISION
2 %
3 % ASSIGNMENT-1, PART-A
4 %
5 % Read given images
6 im1 = imread('famous_five.png');
7 im2 = imread('mausoleum.jpg');
8 harris(im1);
9 harris(im2);

10
11 [-function harris(image)
12 % Displaying given image
13 figure;
14 imshow(image);
15 title('GIVEN IMAGE');

16
17 % Converting given RGB image to GRayscale
18 im = rgb2gray(image);
19 % Displaying GRAYSCALE image
20 figure;
21 imshow(im);
22 title('GRAYSCALE IMAGE');
```

```

23 % Finding Horizontal & Vertical Gradient of GRAYSCALE image
24 [dx,dy]=meshgrid(-1:1, -1:1);
25 ix = conv2(double(im),dx,'same');
26 % Displaying Horizontal gradient/ Vertical Edge
27 figure; imshow(ix);
28 title('HORIZONTAL GRADIENT, Ix');
29 iy = conv2(double(im),dy,'same');
30 % Displaying Vertical gradient/ Horizontal Edge
31 figure; imshow(iy); title('VERTICAL GRADIENT, Iy');

32
33 % PARAMETERS FOR GAUSSIAN FILTER
34 sigma = 3;
35 radius=1;
36 order = (2*radius+1)^2;
37
38 % DEFINING GAUSSIAN FILTER
39 len = max(1,fix(6*sigma));
40 p=len; q=len;
41 [ul,u2]=meshgrid(-(p-1)/2:(p-1)/2, -(q-1)/2: (q-2)/2);
42 ug = exp(-(ul.^2+u2.^2)/(2*sigma^2));
43 [w,z] = size(ug);
44 sum = 0;
45 for i=1:w
46   for j=1:z
47     sum = sum+ug(i,j);
48   end
49 end
50 G = ug ./sum;

51
52 % COMPUTING ELEMENTS OF SECOND MOMENT MATRIX, M
53 Ix2 = conv2(double(ix.^2),G,'same');
54 Iy2 = conv2(double(iy.^2),G,'same');
55 Ixy = conv2(double(ix.*iy),G,'same');

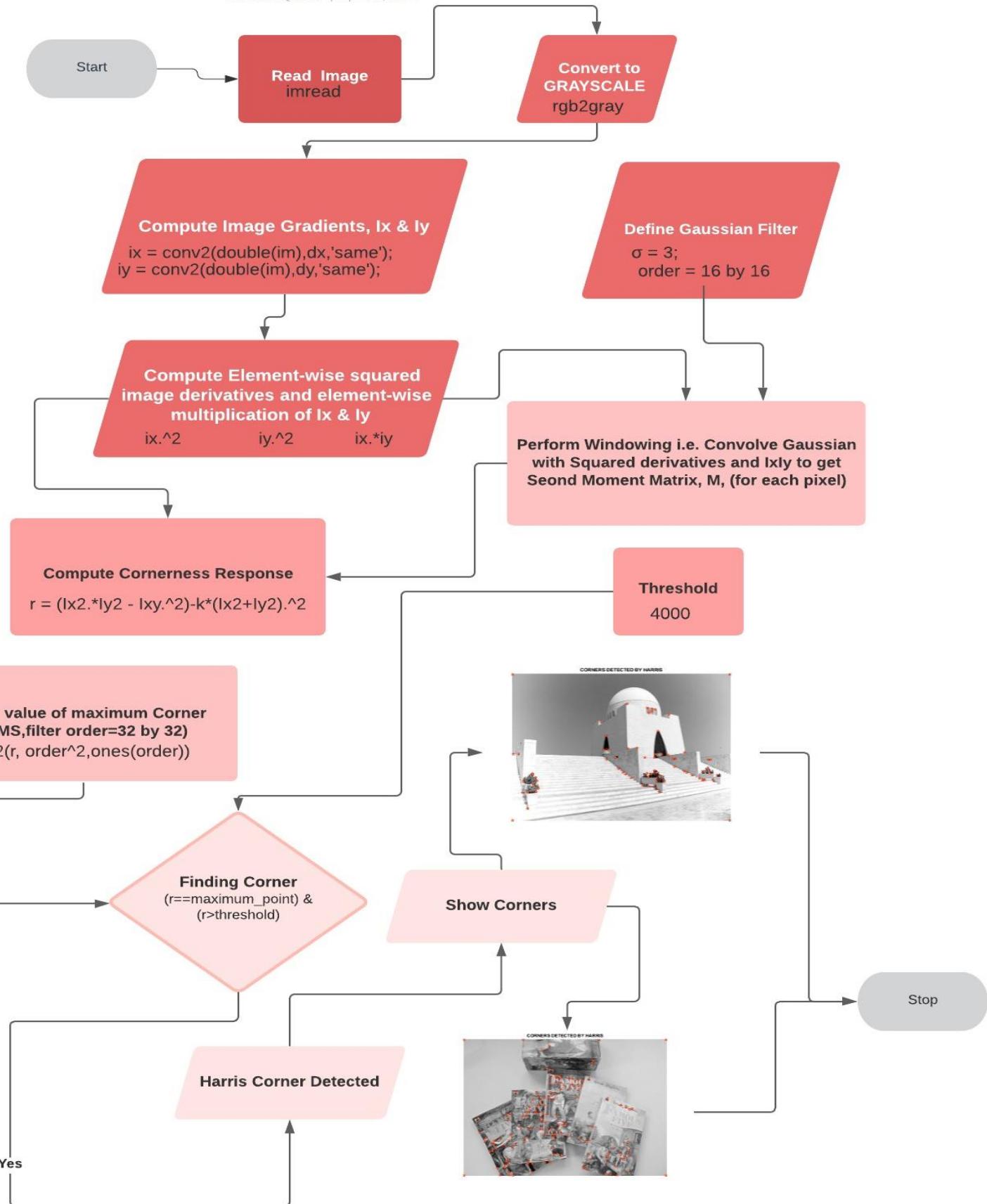
56
57 % CORNERNESS MEASURE
58 r = (Ix2.*Iy2 - Ixy.^2)./(Ix2+Iy2 + eps);

59
60 % Value of Threshold set empirically
61 threshold = 4000;
62
63 % FINDING MAX POINT FOR NON-MAX SUPPRESSSION
64 maximum_point = ordfilt2(r, order^2,ones(order));
65
66 % FINDING CORNERS
67 harris_corners = (r==maximum_point) & (r>threshold);
68 [R,C]=find(harris_corners);
69 figure,imshow(im),hold on,
70 plot(C,R,'yp','MarkerFaceColor','y'),
71 title('CORNERS DETECTED BY HARRIS');
72 end

```

HARRIS CORNER DETECTOR

Monazza Qadeer | April 30, 2021



Results

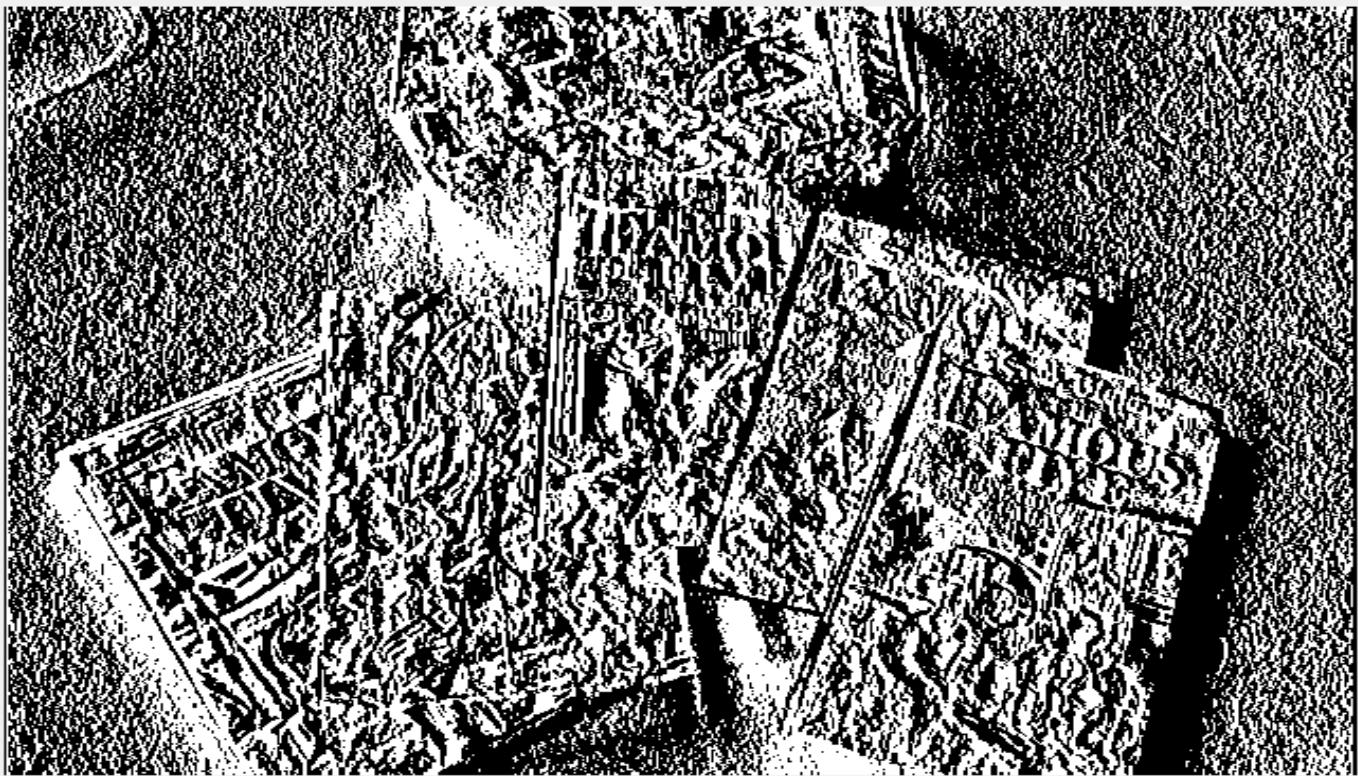
GIVEN IMAGE



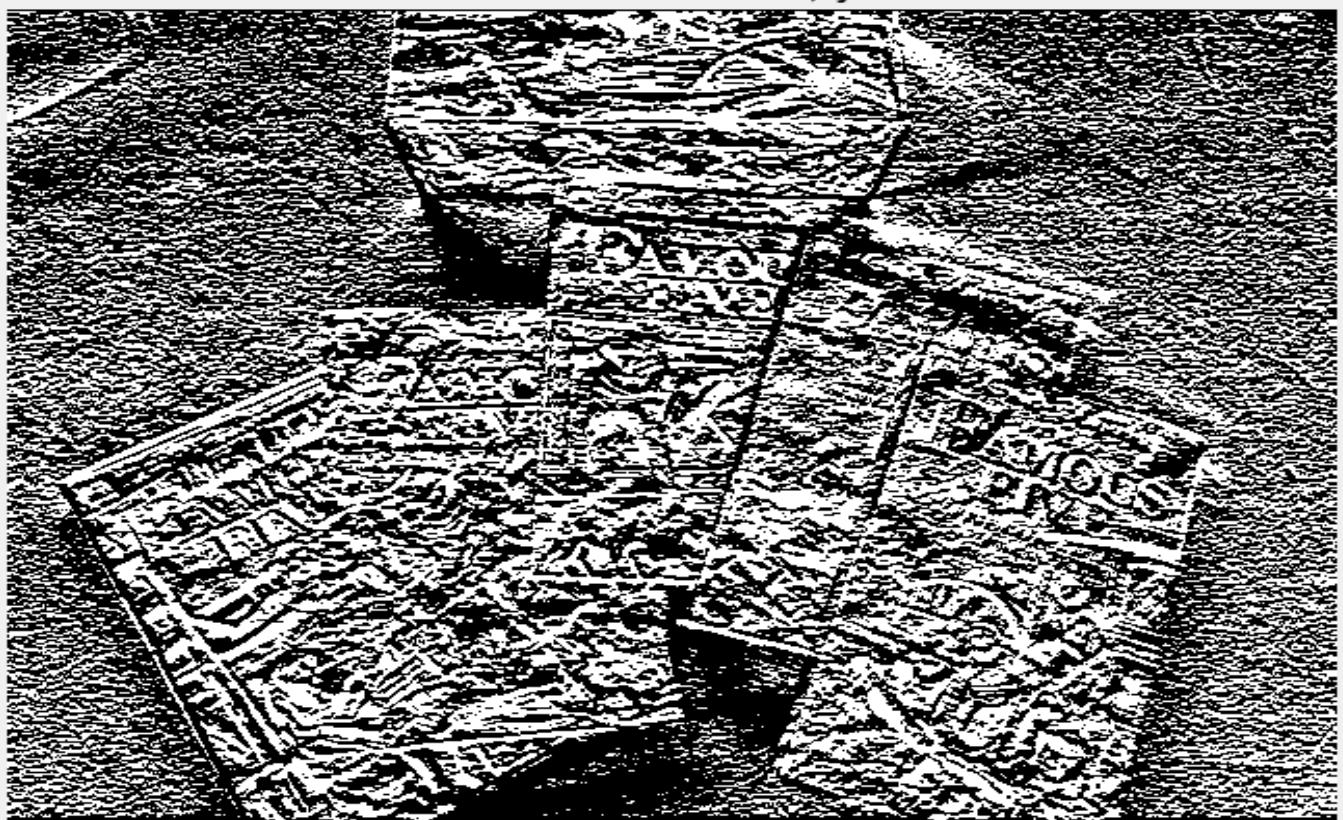
GRAYSCALE IMAGE



HORIZONTAL GRADIENT, I_x



VERTICAL GRADIENT, I_y

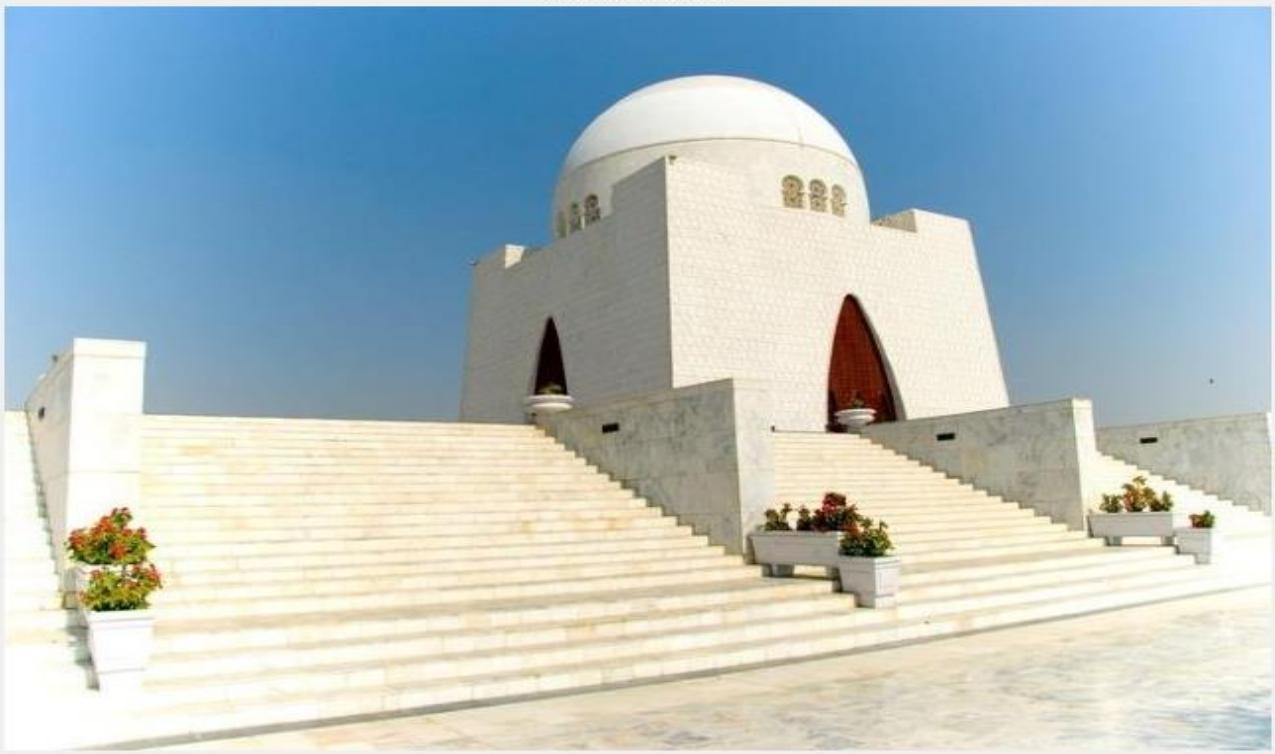


Yellow stars in the following image shows detected corner points.

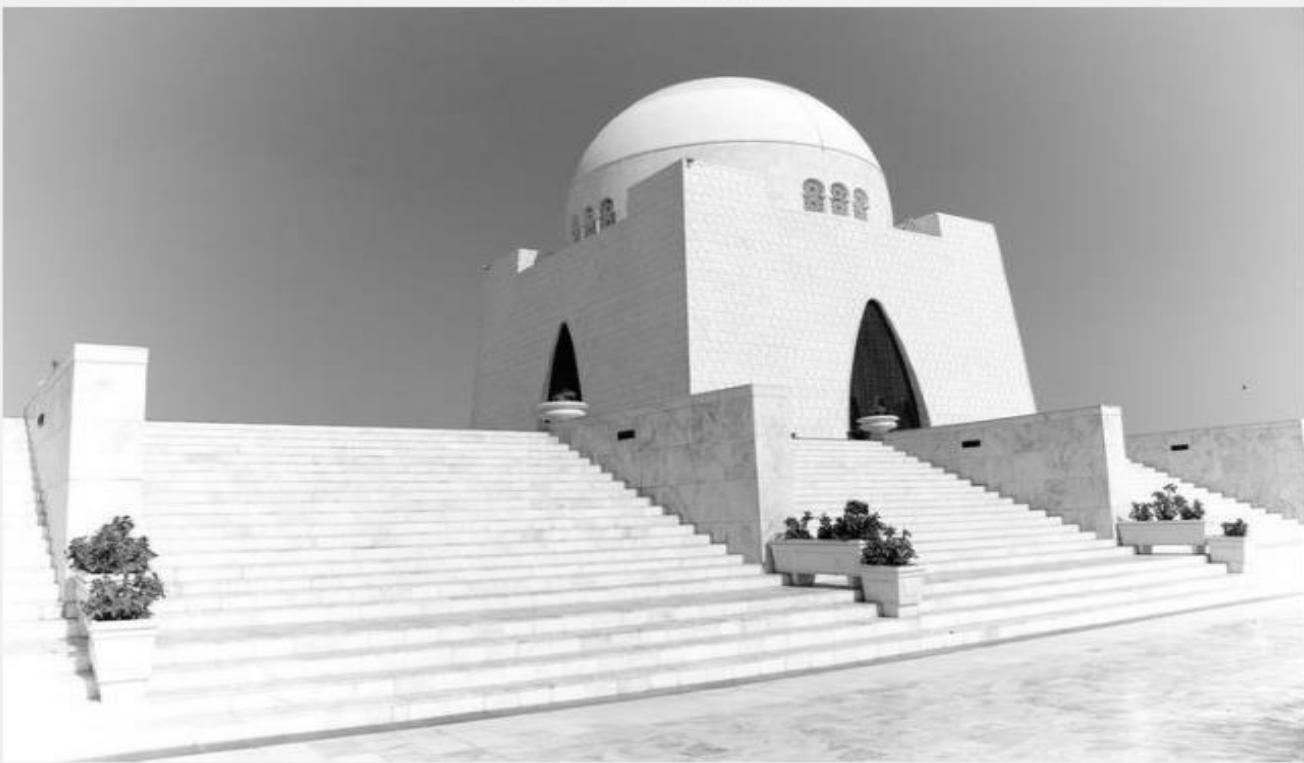
CORNERS DETECTED BY HARRIS



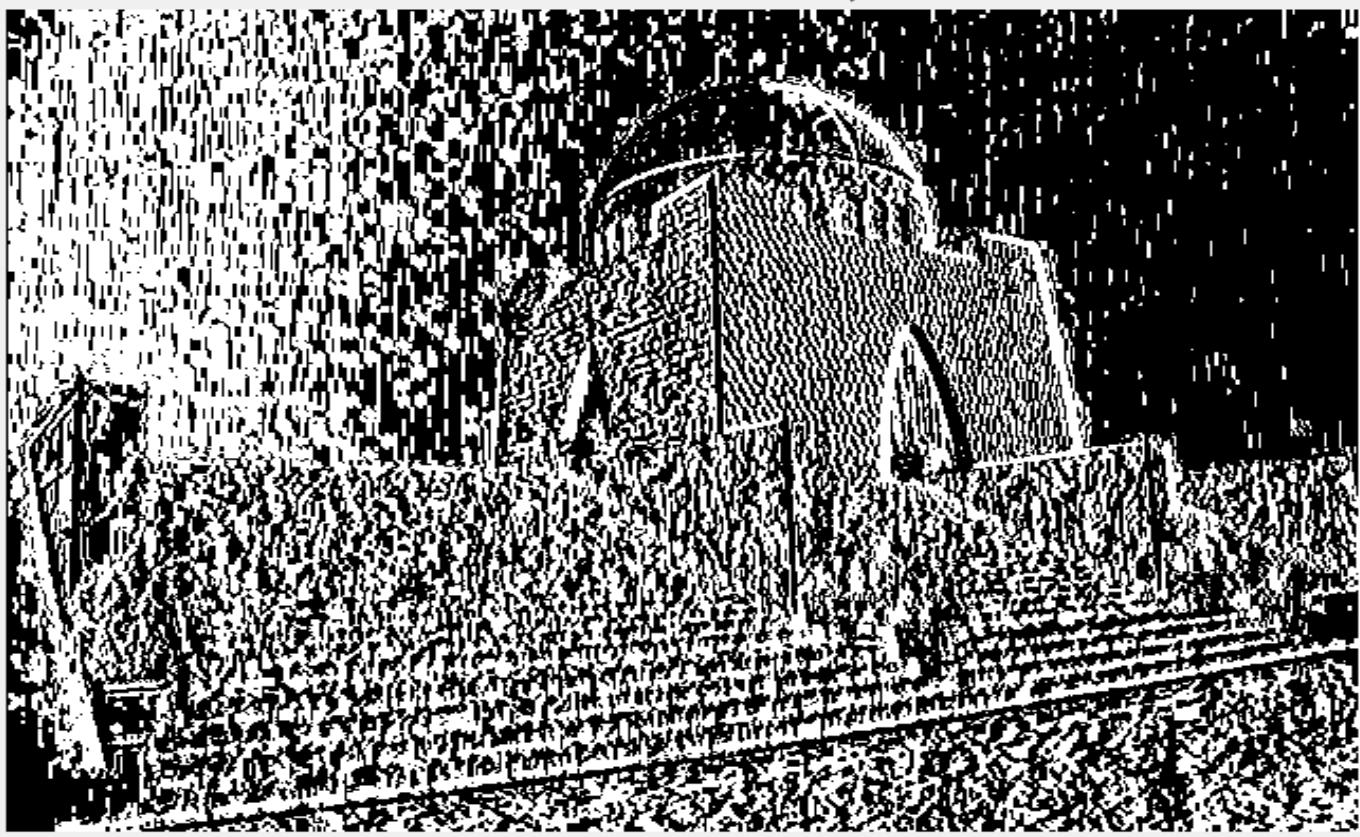
GIVEN IMAGE



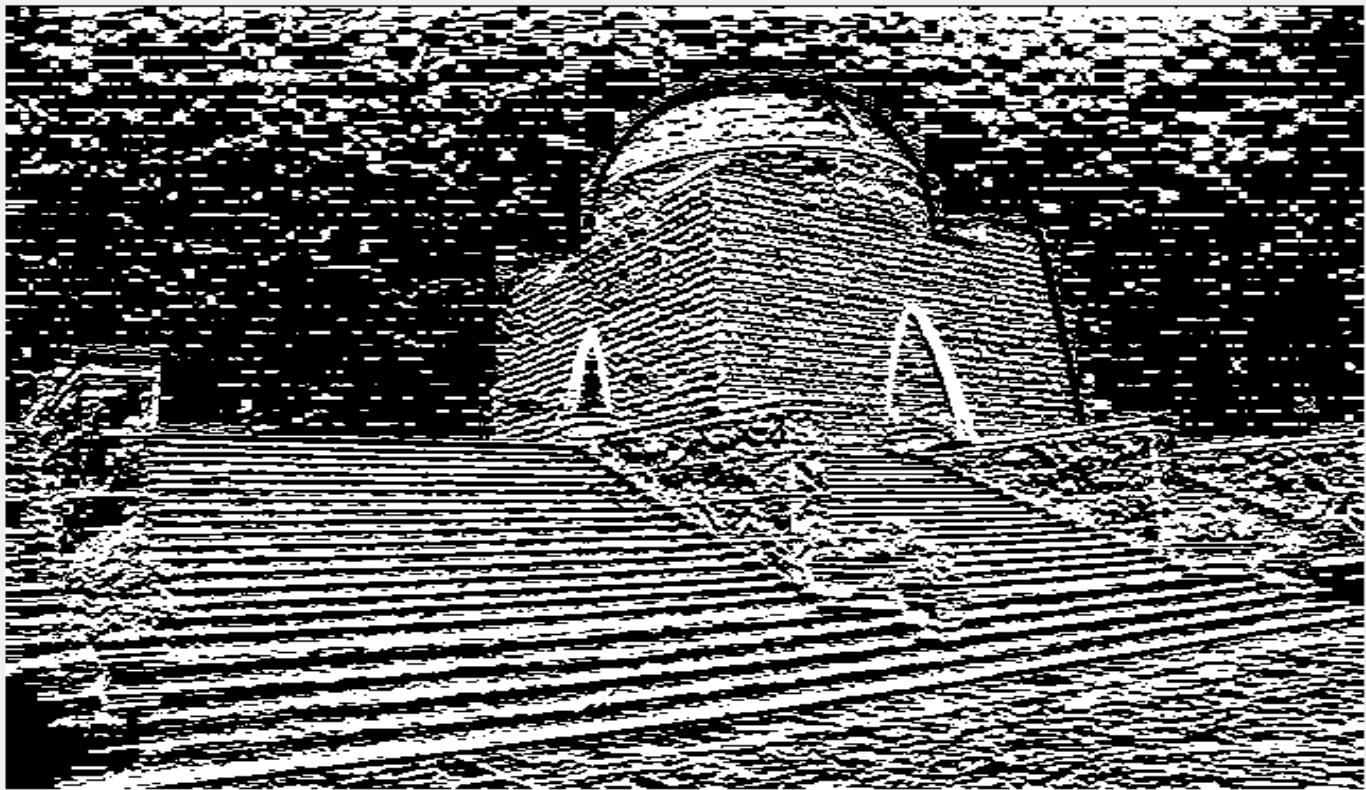
GRAYSCALE IMAGE



HORIZONTAL GRADIENT, I_x

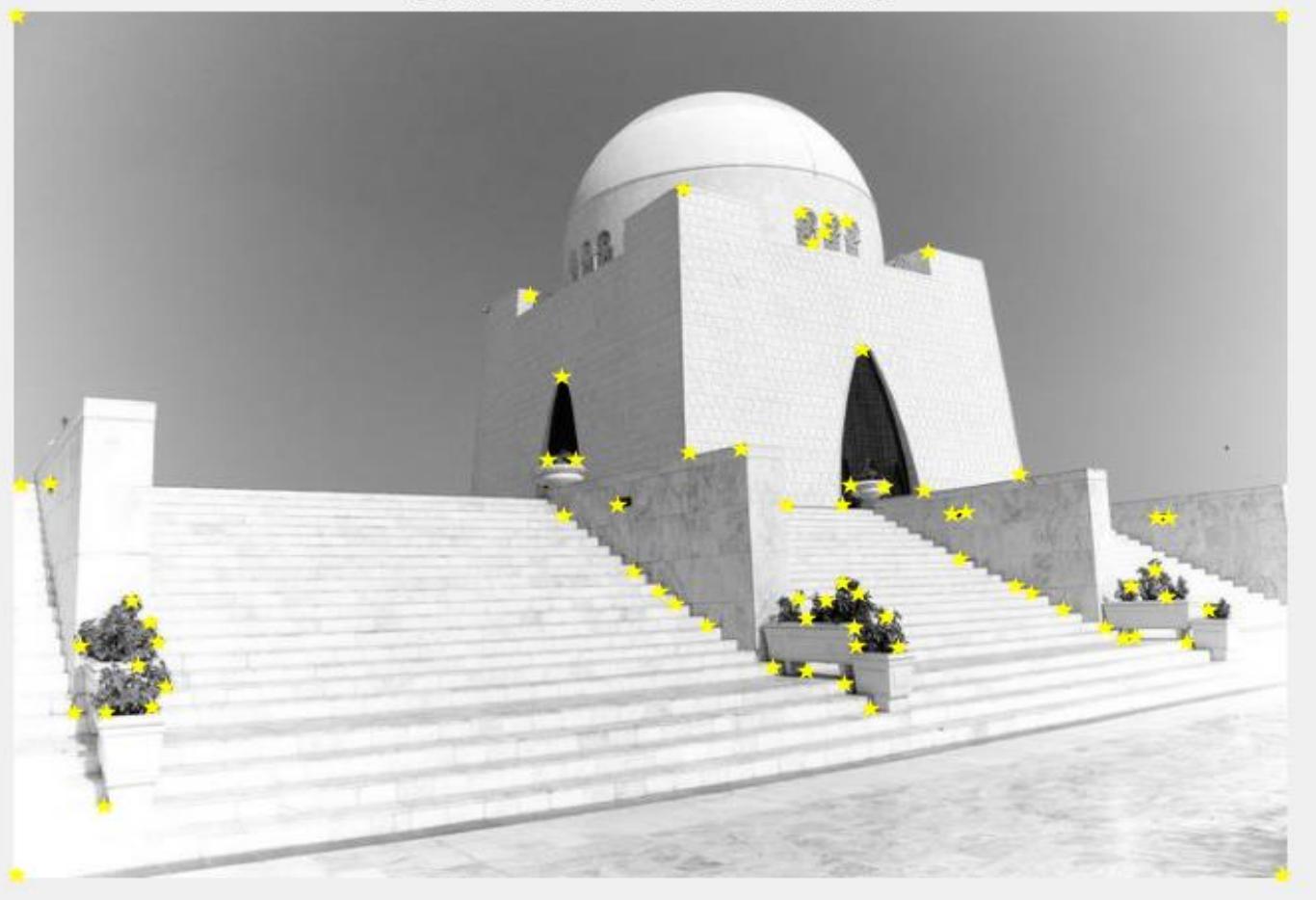


VERTICAL GRADIENT, I_y



Yellow stars in the following image shows detected corner points.

CORNERS DETECTED BY HARRIS



Conclusion

The yellow stars in the grayscale images depicts the corner points. A medium value of threshold (4000) is set, not too high to miss out on potential keypoints and not too low to get redundant keypoints and also to avoid any ambiguity i.e. to get clear, visible and well separated keypoints for better visualization.

Famous five, the image

This image has good corners as there is a lot of color change even within a single book cover. So, a set of books gives rise to good potential corners such as:

- Name of author at the top of book cover, “Enid Blyton”
- Title of the book series (Famous Five), written in large font
- Paper embossing giving depth to the written words, it creates good corners
- Kids playing around doing different kinds of activities, giving detail to the image, creates good corner points

Mausoleum, the image

As the image contains good and sharp corners so it can be easily found in the following locations:

- **flower pots**, as there is a sharp rate of change in intensity of pixels in every direction with green leaves and red flowers making a good combination for a corner point
- the finishing points or **ends of stairs** also makes a good corner point
- **the sharp corners of mausoleum itself**
- **the small windows** in the dome or vault of mausoleum

PART- B

Rotate the image in increments of 15° , from 0° to 360° . You are allowed use of library function(s) for rotation. For each rotated image, compute Harris keypoints using the same settings that you chose in part (a). Then, compute the repeatability using the following procedure:

- Set number of feature matches M to be 0
- For each Harris keypoint at $[x \ y]$ in the original image:
 - Predict the position $\mathbf{x}_r = [x_r, y_r]$ where the keypoint should ideally appear in the rotated image (by applying a rotation transformation on $[x \ y]$).
 - Search for a nearby Harris keypoint which is detected in the rotated image with coordinates $\mathbf{x}_0 = [x_0, y_0]$ satisfying $\|\mathbf{x}_0 - \mathbf{x}_r\| \leq T$, where T is some suitably chosen threshold.
 - If such an $\mathbf{x}_0 = [x_0, y_0]$ is found, increment number of feature matches M by 1. This matching point \mathbf{x}_0 in the rotated image should now not be considered for matching in subsequent iterations.
- Compute repeatability as M/N , where M is the number of feature matches and N is the number of Harris keypoints in the original image.

Plot repeatability against rotation angle and comment on the Harris keypoint detector's robustness against in-plane rotation.

The following built-in functions are used for computations:

- Rimage1 = imrotate(Oimage1,angle) for image rotation
- RotMatrix = [cosd(rot_ang) -sind(rot_ang);sind(rot_ang) cosd(rot_ang)], this is Transformation Matrix for rotation of Corner point
- RotatedP = RotMatrix*((Ocpoints(i))'-CenterO)+CenterR for finding rotated keypoint
- thresh = 100, value of threshold as a reference for finding corner whose Euclidean distance must be less than the set threshold
- Euc_D = sqrt(sum((Rcpoints(i)-RotatedP).^2)) for finding Euclidean distance between predicted keypoint location and keypoint in rotated image

Code

```
1 % CS-867 COMPUTER VISION
2 % ASSIGNMENT-1, PART-B
3 % ROBUSTNESS OF HARRIS KEYPOINT DETECTOR TO ROTATION
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5
6 - Oimage1 = imread('famous_five.png');
7 - Oimage2 = imread('mausoleum.jpg');
8 - rot_eff(Oimage1);
9 - rot_eff(Oimage2);
10
11 [- function rot_eff(Oimage1)
12 - [Orows,Ocols] = harris(Oimage1);
13 - Ocpoints = [Orows,Ocols];
14
```

```
--> 15 % ROTATING IMAGE IN INCREMENTS OF 15 Deg ALL THE WAY FROM 0 Deg to 360 Deg
--> 16 - for angle=0:15:360
--> 17 -     Rimage1 = imrotate(Oimage1,angle); % ROTATE THE IMAGE
--> 18 -     [Rrows,Rcols] = harris(Rimage1);
--> 19 -     Rcpoints = [Rrows,Rcols];
--> 20 -     threshold = 4000;
--> 21 -     Ocpointssize = length(Ocpoints);
--> 22 -     N=length(Ocpoints);% COUNT OF HARRIS CORNERS IN ORIGINAL IMAGE
--> 23 -     rot_ang=15*pi/180; % ROTATION ANGLE FOR IMAGE
--> 24
--> 25 % TRANSFORMATION MATRIX FOR ROTATION OF KEYPOINT
--> 26 - RotMatrix = [cosd(rot_ang) -sind(rot_ang);sind(rot_ang) cosd(rot_ang)]
--> 27
--> 28 % RETURNS THE CENTER OF ORIGINAL IMAGE (GRAYSCALE)
--> 29 - CenterO = (size(rgb2gray(Oimage1))/2)';
--> 30
--> 31 % RETURNS THE CENTER OF ROTATED IMAGE (GRAYSCALE)
--> 32 - CenterR = (size(rgb2gray(Rimage1))/2)';
--> 33 - M=0; % COUNT OF MATCHED CORNERS
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    % PREDICTING IDEAL POSITION OF KEYPOINT
    for i=1:l:Ocpointssize
        % KEYPOINT ROTATED BY TRANSFORMATION MATRIX
        RotatedP = RotMatrix*((Ocpoints(i))-CenterO)+CenterR;
        % EUCLIDEAN DISTANCE BTW TRANSFORMED & ROTATED IMAGE KEYPOINT
        Euc_D = sqrt(sum((Rcpoints(i)-RotatedP).^2));
        if(Euc_D<threshold) % IF DISTANCE LESS THAN THRESHOLD, THEN
            M=M+1; % INCREMENT IN COUNT OF MATCHED KEYPOINTS
        end
    end
    rep=M/N; %REPEATABILITY, M= MATCHED KEYPOINTS & N= ORIGINAL KEYPOINTS
    figure,imshow(Rimage1),hold on,
    plot(Rcols,Rrows,'yp','MarkerFaceColor','y'),
    title(['Rotated image with ',num2str(angle),
           ' degree where M = ',num2str(M),', N = ',num2str(N),
           ' and Repeatability M/N = ',num2str(rep)]);
end
end

% HARRIS KEYPOINT DETECTOR AS IN PART-A
function [R,C] = harris(image)
im = rgb2gray(image);
[dx,dy]=meshgrid(-1:1, -1:1);
ix = conv2(double(im),dx, 'same');
iy = conv2(double(im),dy, 'same');
% PARAMETERS FOR GAUSSIAN FILTER
sigma = 3;
radius=1;
order = (2*radius+1)^2;
% DEFINING GAUSSIAN FILTER
len = max(1,fix(6*sigma));
p=len; q=len;
[u1,u2]=meshgrid(-(p-1)/2:(p-1)/2, -(q-1)/2: (q-2)/2);
ug = exp(-(u1.^2+u2.^2)/(2*sigma^2));
[w,z] = size(ug);
sum = 0;
for i=1:w
    for j=1:z
        sum = sum+ug(i,j);
    end
end
G = ug ./sum;

```

```

79
80      %COMPUTING ELEMENTS OF SECOND MOMENT MATRIX, M
81  Ix2 = conv2(double(ix.^2),G,'same');
82  Iy2 = conv2(double(iy.^2),G,'same');
83  Ixy = conv2(double(ix.*iy),G,'same');
84  % CORNERNESS MEASURE
85  r = (Ix2.*Iy2 - Ixy.^2)./(Ix2+Iy2 + eps);
86  % Value of Threshold set empirically
87  threshold = 4000;
88  % FINFDING MAX POINT FOR NON-MAX SUPPRESSSION
89  maximum_point = ordfilt2(r, order^2,ones(order));
90  % FINDING CORNERS
91  harris_corners = (r==maximum_point) & (r>threshold);
92  [R,C]=find(harris_corners);
93  end
94

```

Important Note:

Repeatability of matched keypoints is checked against two values of threshold. In first observation it is set to be 100 i.e. keeping it minimum as it should be ideally, as the distance between predicted keypoint and the one in rotated image should be close to each other. In second observation the threshold value is set the same as used in Harris corner detector in Part-A and for some reason it's giving a 100 percent repeatability. Although, it is not supported logically. The repeatability plots for both observations can be seen in the results section.

Observation # 01

Threshold value is set to be 100 and the Euclidean distance is compared with it as shown in the code snippet below. Plots showing corners can be seen in the results section. The number of matched keypoints comes out be less than the keypoints in original image

Code snippet

```
34 % PREDICTING IDEAL POSITION OF KEYPOINT
35 thresh = 100;
36 - for i=1:l:Ocpointssize
37 -     % KEYPOINT ROTATED BY TRANSFORMATION MATRIX
38 -     RotatedP = RotMatrix*((Ocpoints(i))'-Center0)+CenterR;
39 -     % EUCLIDEAN DISTANCE BTW TRANSFORMED & ROTATED IMAGE KEYPOINT
40 -     Euc_D = sqrt(sum((Rcpoints(i)-RotatedP).^2));
41 -     if(Euc_D
```

Results



Rotated image with 15 degree where M = 36 , N = 152 and Repeatability M/N = 0.23684



Rotated image with 30 degree where M = 42 , N = 152 and Repeatability M/N = 0.27632



Rotated image with 45 degree where M = 31 , N = 152 and Repeatability M/N = 0.20395



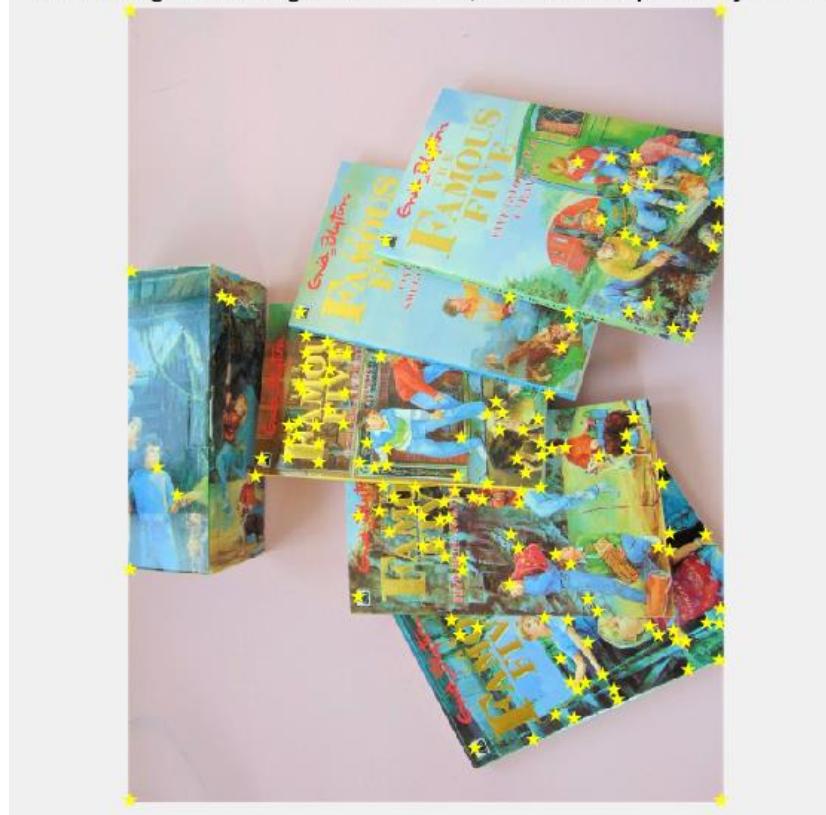
Rotated image with 60 degree where M = 21 , N = 152 and Repeatability M/N = 0.13816



Rotated image with 75 degree where M = 13 , N = 152 and Repeatability M/N = 0.085526



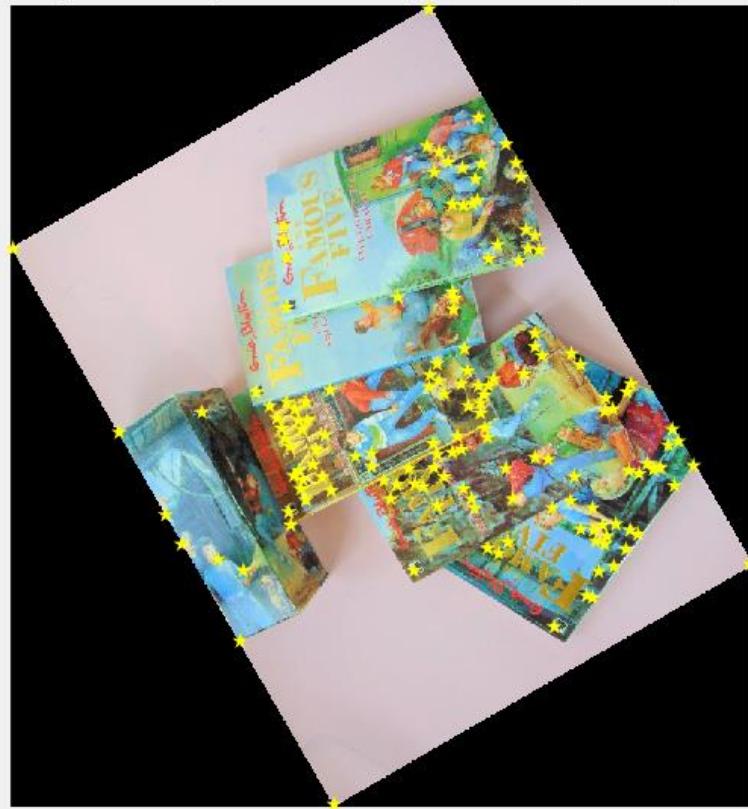
Rotated image with 90 degree where M = 0 , N = 152 and Repeatability M/N = 0



Rotated image with 105 degree where M = 9 , N = 152 and Repeatability M/N = 0.059211



Rotated image with 120 degree where M = 31 , N = 152 and Repeatability M/N = 0.20395



Rotated image with 135 degree where M = 40 , N = 152 and Repeatability M/N = 0.26316



Rotated image with 150 degree where M = 35 , N = 152 and Repeatability M/N = 0.23026



Rotated image with 165 degree where M = 41 , N = 152 and Repeatability M/N = 0.26974



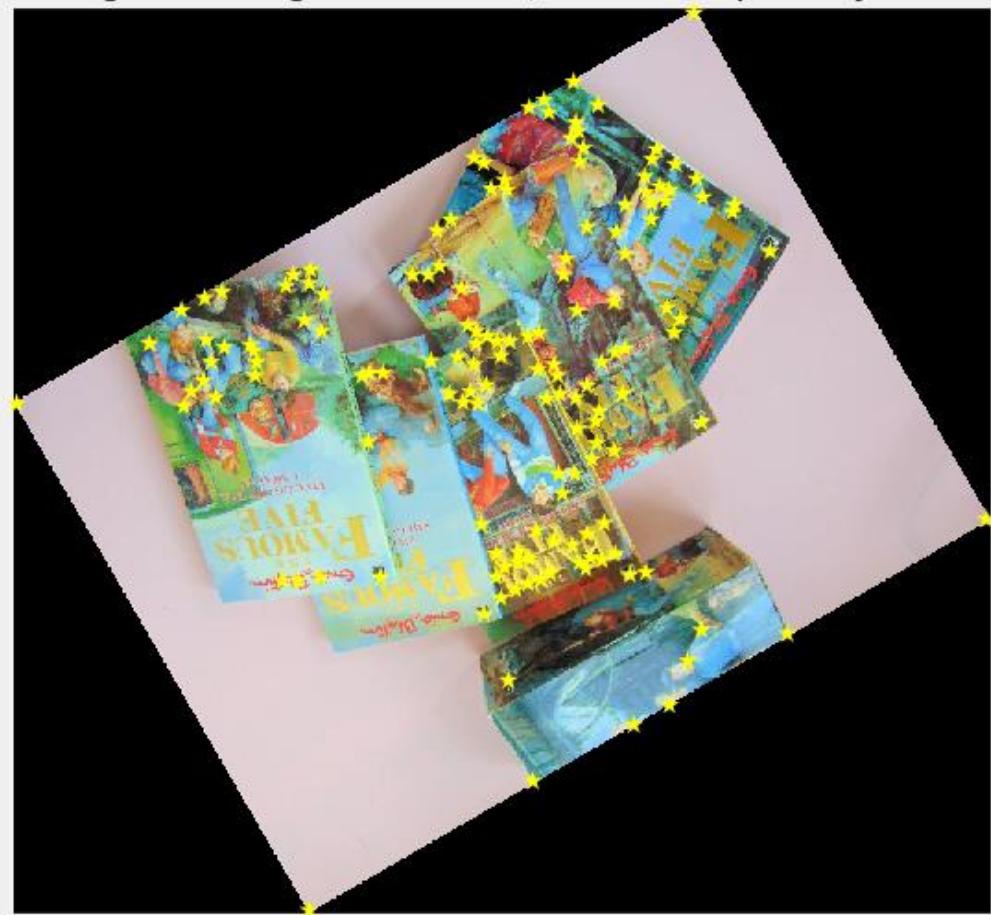
Rotated image with 180 degree where M = 46 , N = 152 and Repeatability M/N = 0.30263



Rotated image with 195 degree where M = 28 , N = 152 and Repeatability M/N = 0.18421



Rotated image with 210 degree where M = 36 , N = 152 and Repeatability M/N = 0.23684



Rotated image with 225 degree where M = 31 , N = 152 and Repeatability M/N = 0.20395



Rotated image with 240 degree where M = 34 , N = 152 and Repeatability M/N = 0.22368



Rotated image with 255 degree where M = 6 , N = 152 and Repeatability M/N = 0.039474



Rotated image with 270 degree where M = 0 , N = 152 and Repeatability M/N = 0



Rotated image with 285 degree where M = 5 , N = 152 and Repeatability M/N = 0.032895



Rotated image with 300 degree where M = 28 , N = 152 and Repeatability M/N = 0.18421



Rotated image with 315 degree where M = 39 , N = 152 and Repeatability M/N = 0.25658



Rotated image with 330 degree where M = 52 , N = 152 and Repeatability M/N = 0.34211



Rotated image with 345 degree where M = 37 , N = 152 and Repeatability M/N = 0.24342

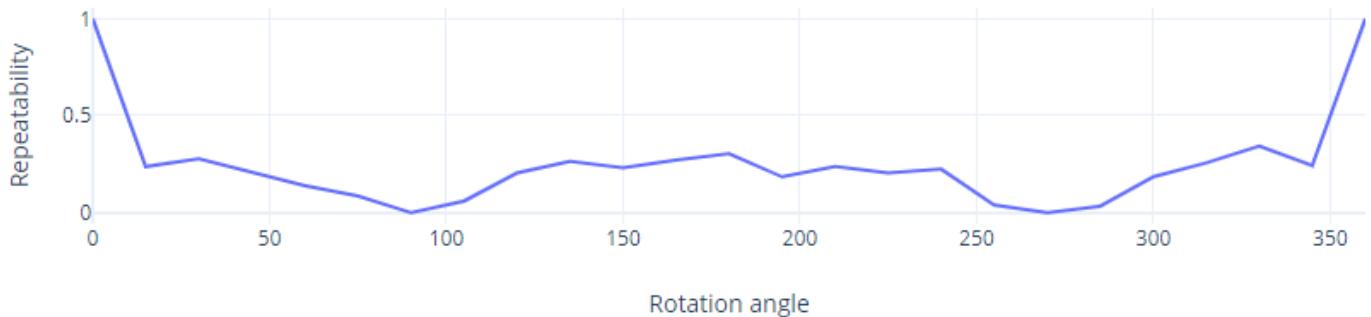


Rotated image with 360 degree where M = 152 , N = 152 and Repeatability M/N = 1



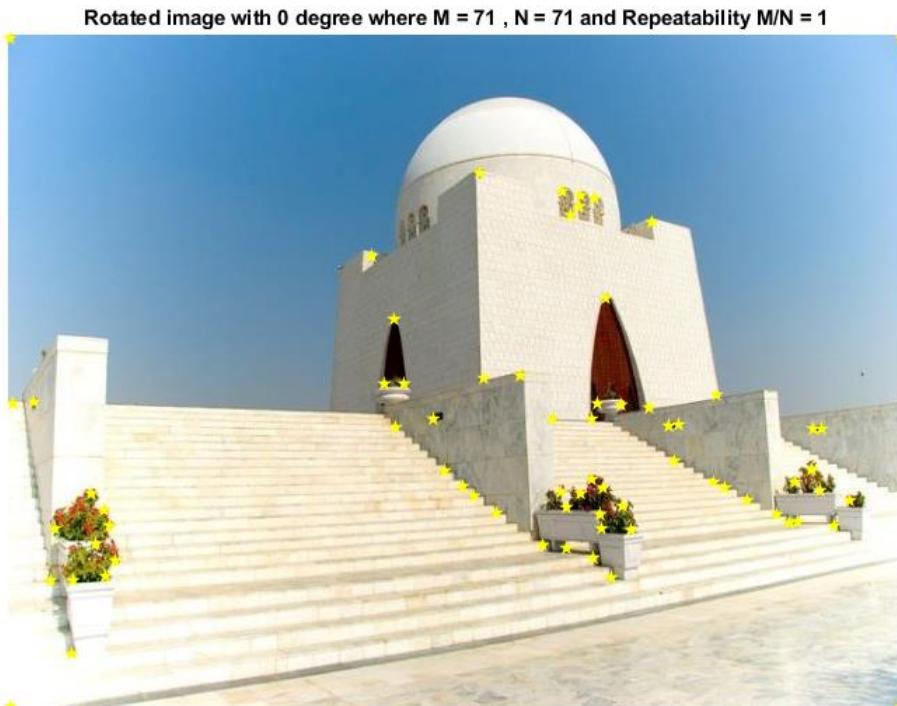
Repeatability vs Rotation Plot for the Image Famous Five

Robustness of Harris Corner Detector to Rotation



Conclusion

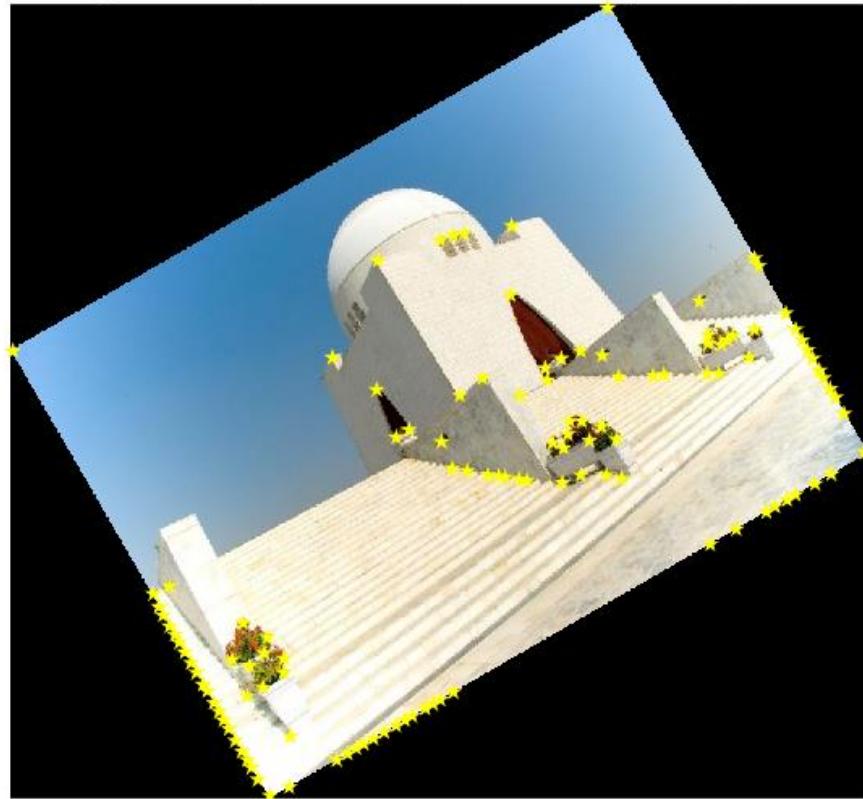
Ideally, the repeatability plot for rotation must maintain the ratio at 0.95 but due to some unknown reason or inefficient programming, it falls as soon it gets rotated and couldn't locate the rotated keypoints. Harris detector is robust to rotation but here the results are not so favorable.



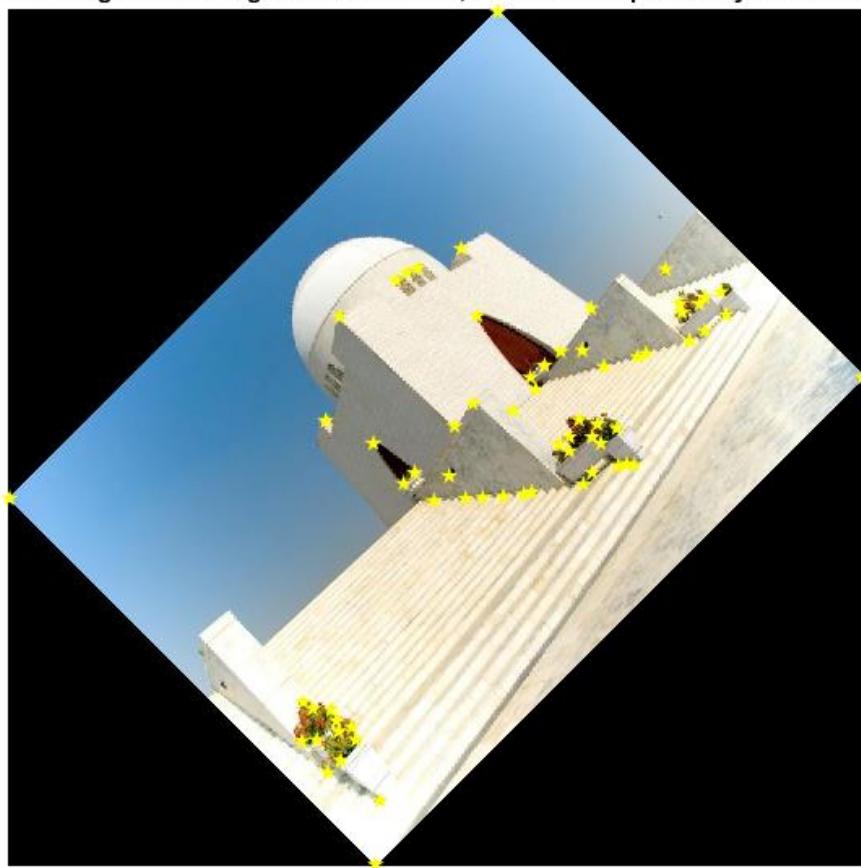
Rotated image with 15 degree where M = 18 , N = 71 and Repeatability M/N = 0.25352



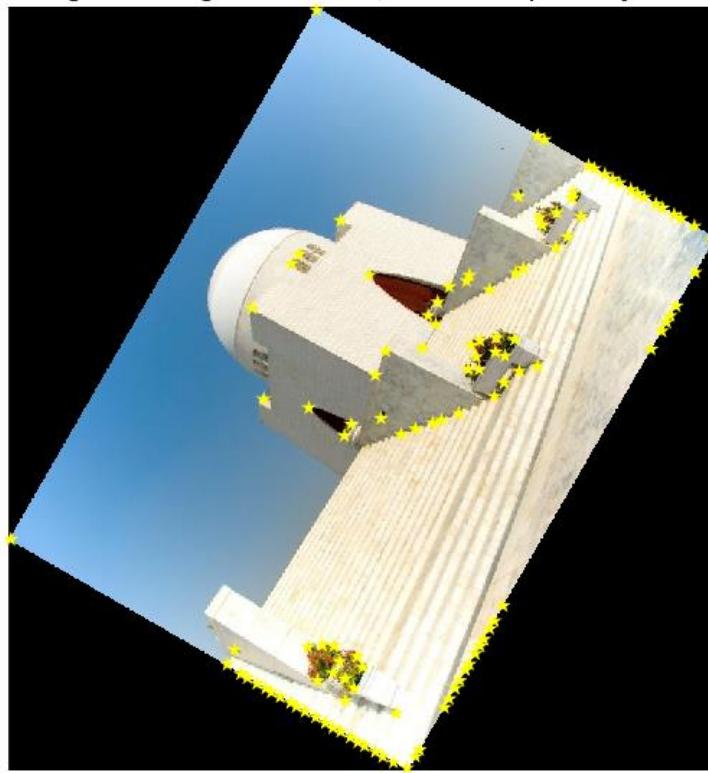
Rotated image with 30 degree where M = 14 , N = 71 and Repeatability M/N = 0.19718



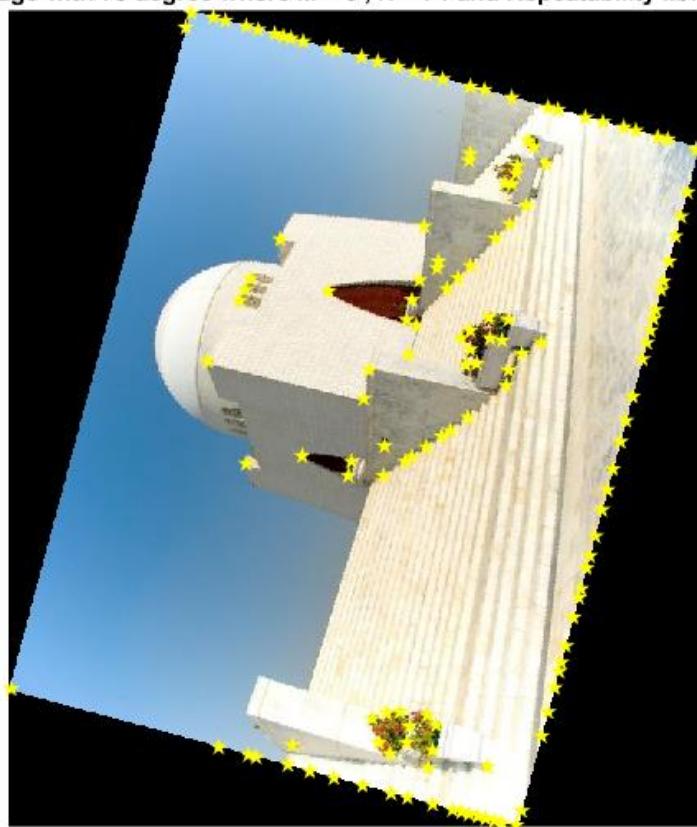
Rotated image with 45 degree where M = 16 , N = 71 and Repeatability M/N = 0.22535



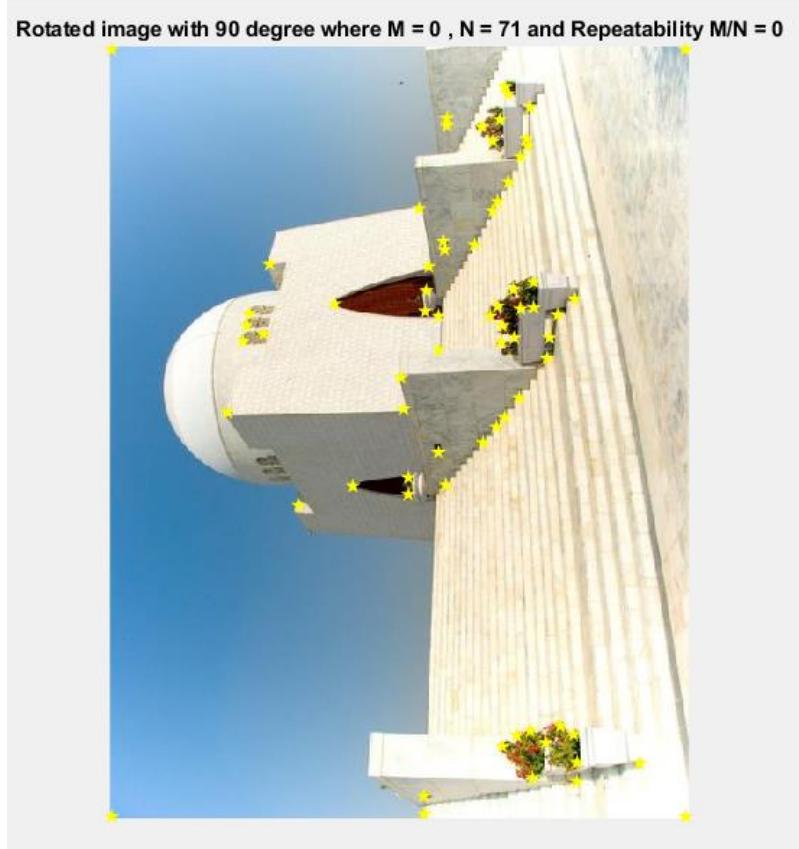
Rotated image with 60 degree where M = 9 , N = 71 and Repeatability M/N = 0.12676



Rotated image with 75 degree where M = 5 , N = 71 and Repeatability M/N = 0.070423



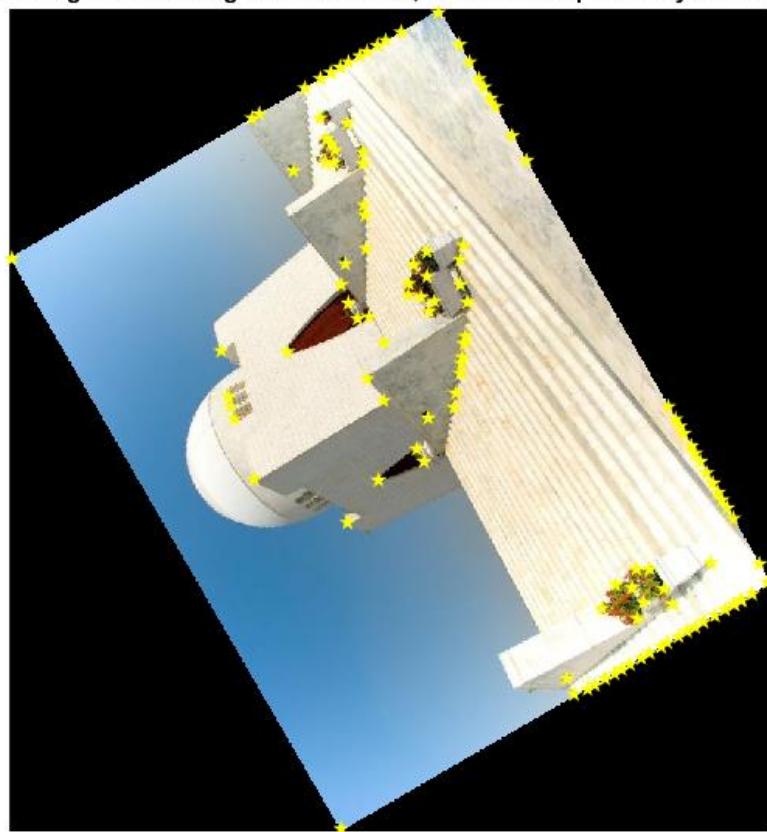
Rotated image with 90 degree where M = 0 , N = 71 and Repeatability M/N = 0



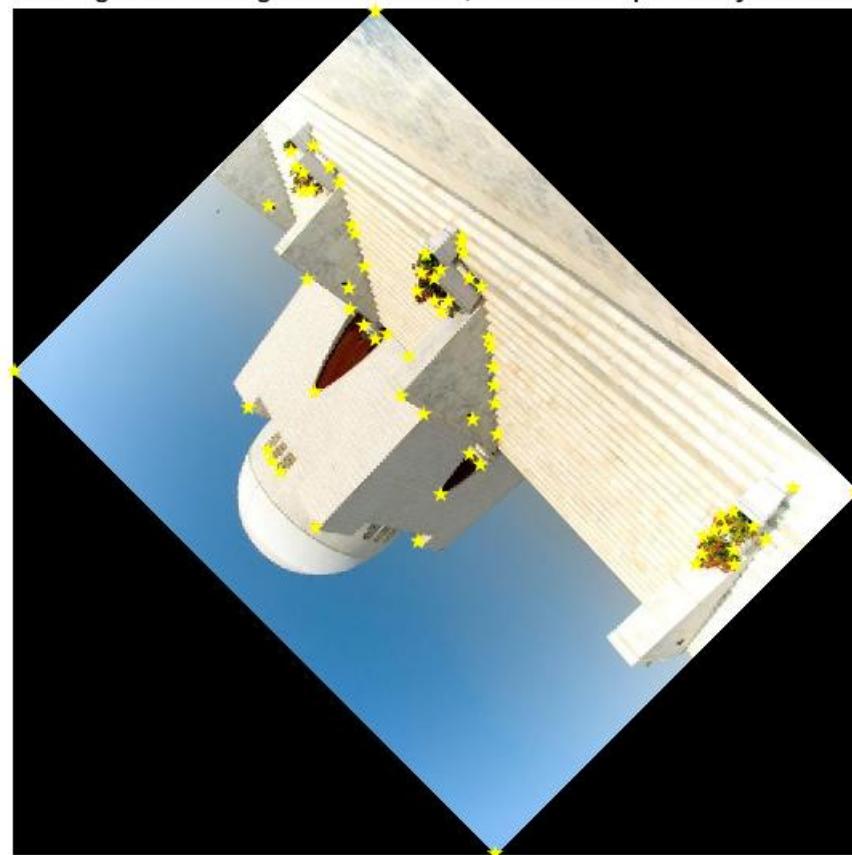
Rotated image with 105 degree where M = 2 , N = 71 and Repeatability M/N = 0.028169



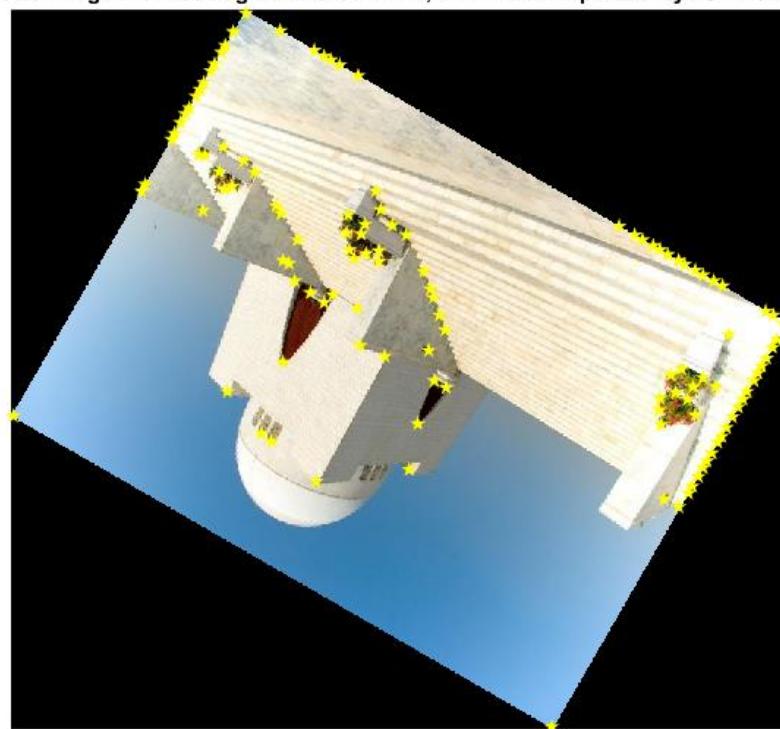
Rotated image with 120 degree where M = 9 , N = 71 and Repeatability M/N = 0.12676



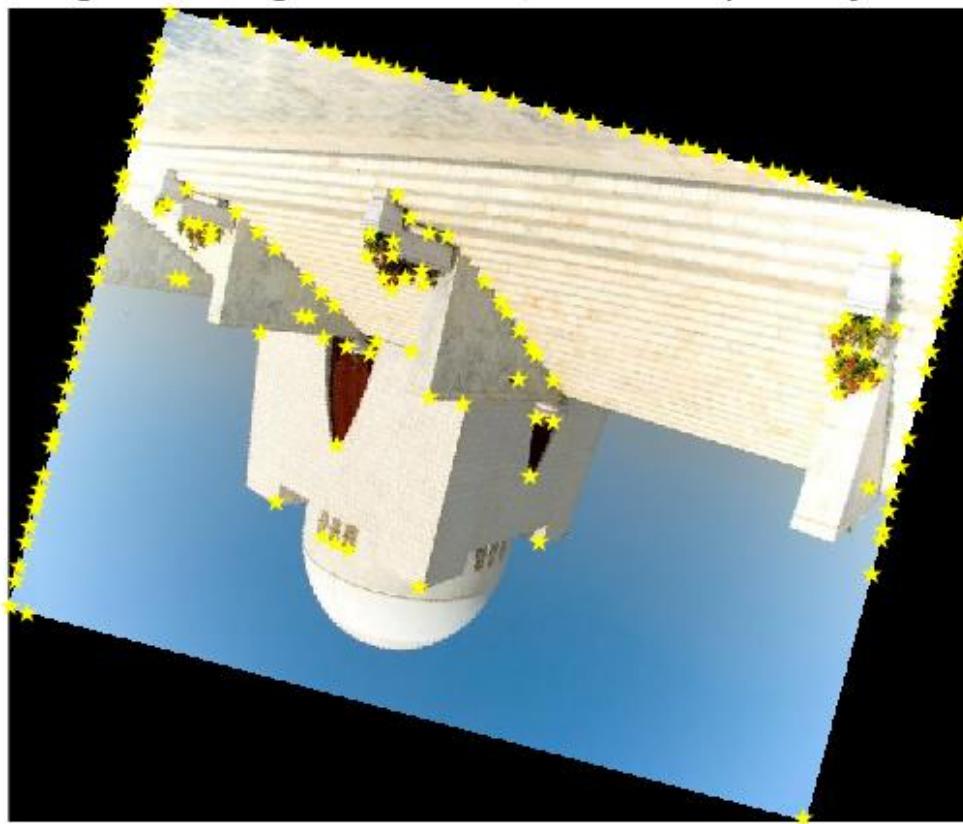
Rotated image with 135 degree where M = 22 , N = 71 and Repeatability M/N = 0.30986



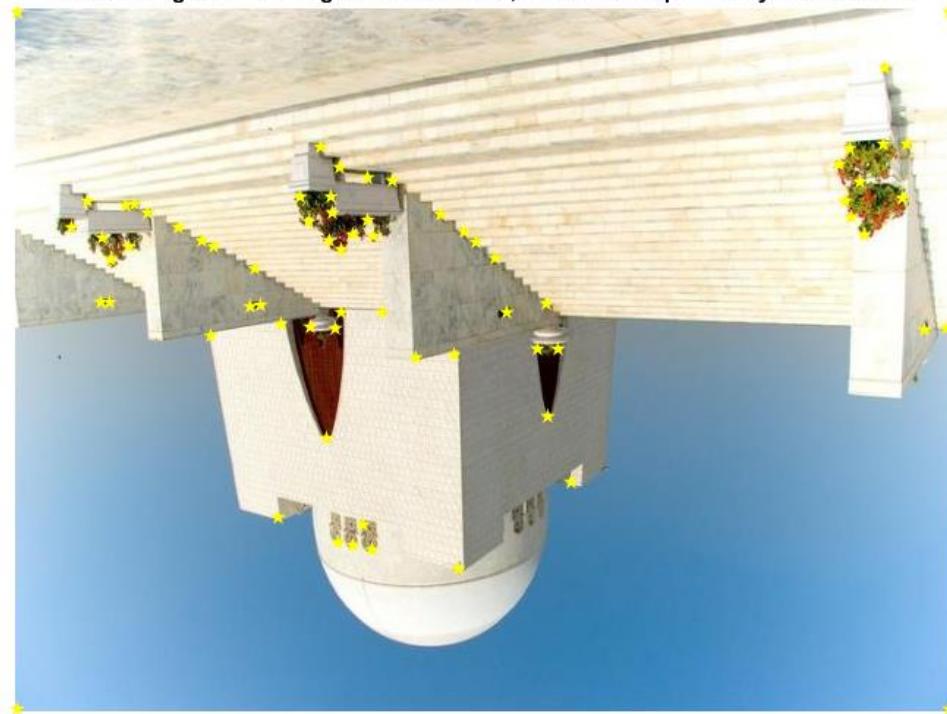
Rotated image with 150 degree where M = 11 , N = 71 and Repeatability M/N = 0.15493



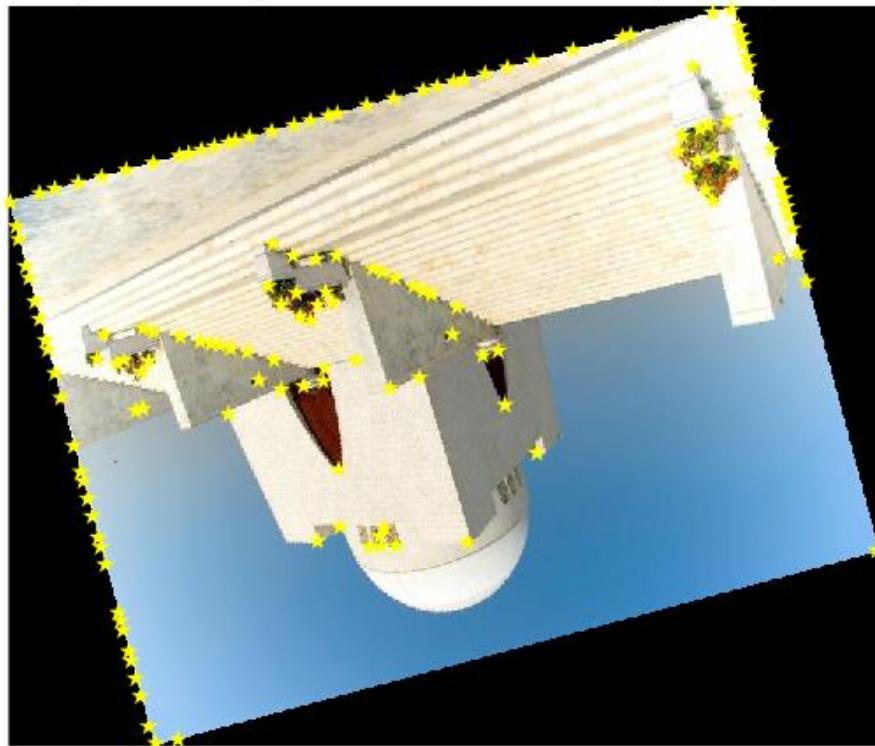
Rotated image with 165 degree where M = 21 , N = 71 and Repeatability M/N = 0.29577



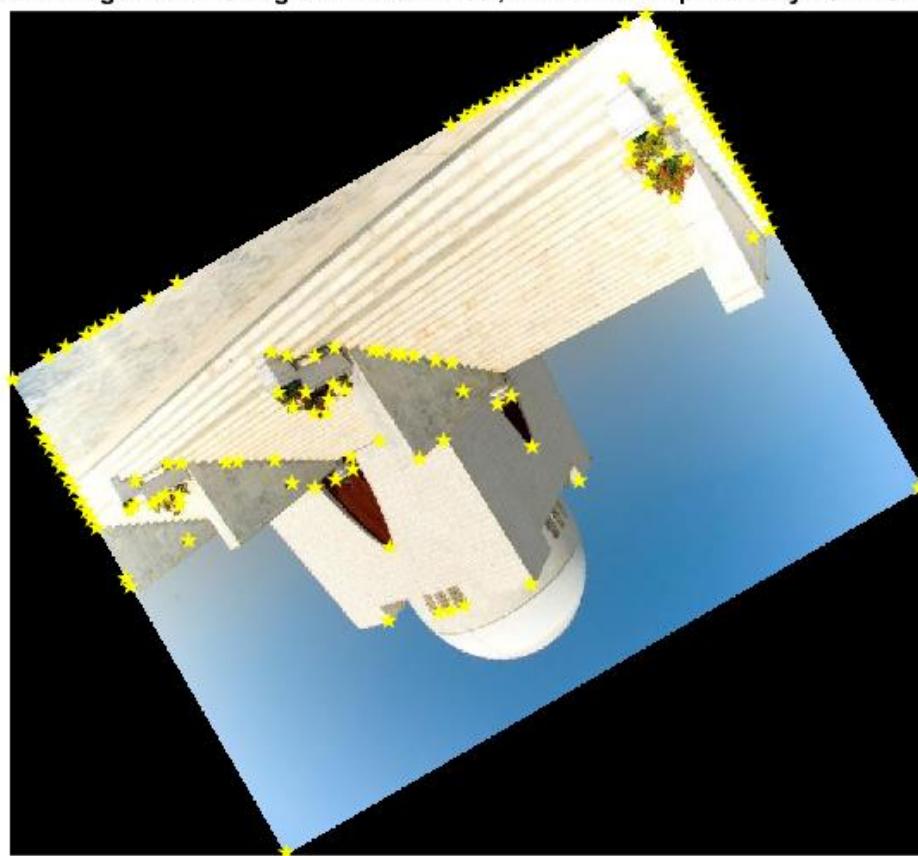
Rotated image with 180 degree where M = 18 , N = 71 and Repeatability M/N = 0.25352



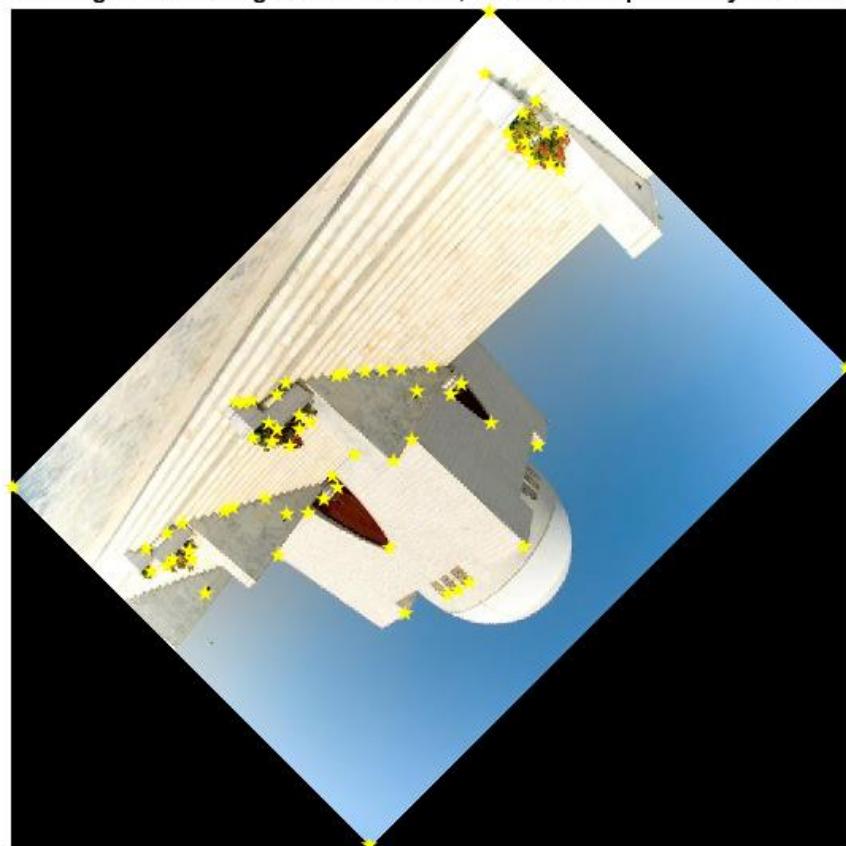
Rotated image with 195 degree where M = 9 , N = 71 and Repeatability M/N = 0.12676



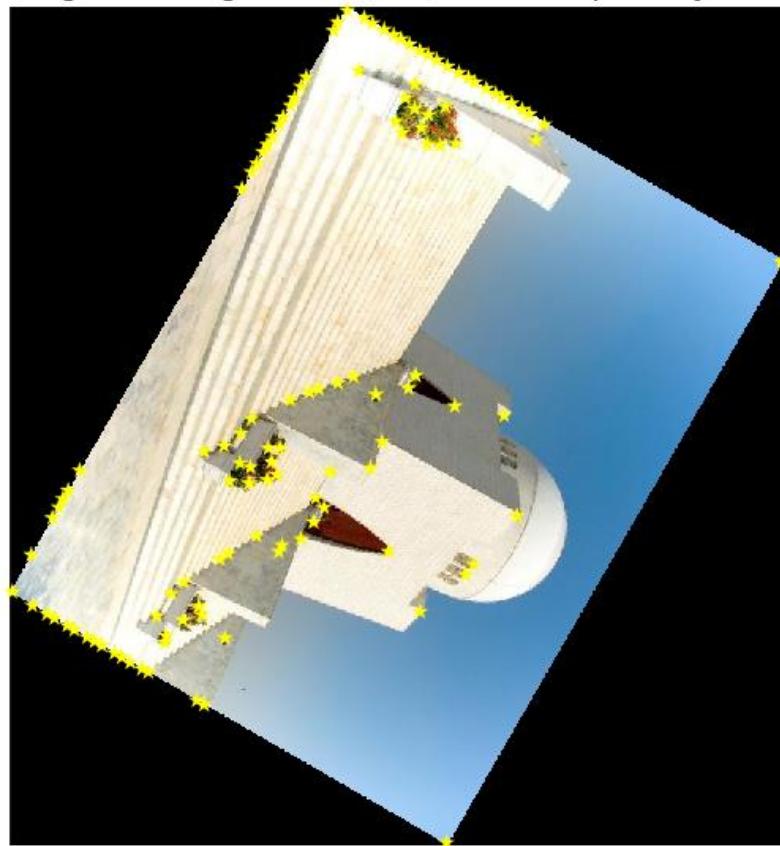
Rotated image with 210 degree where M = 30 , N = 71 and Repeatability M/N = 0.42254



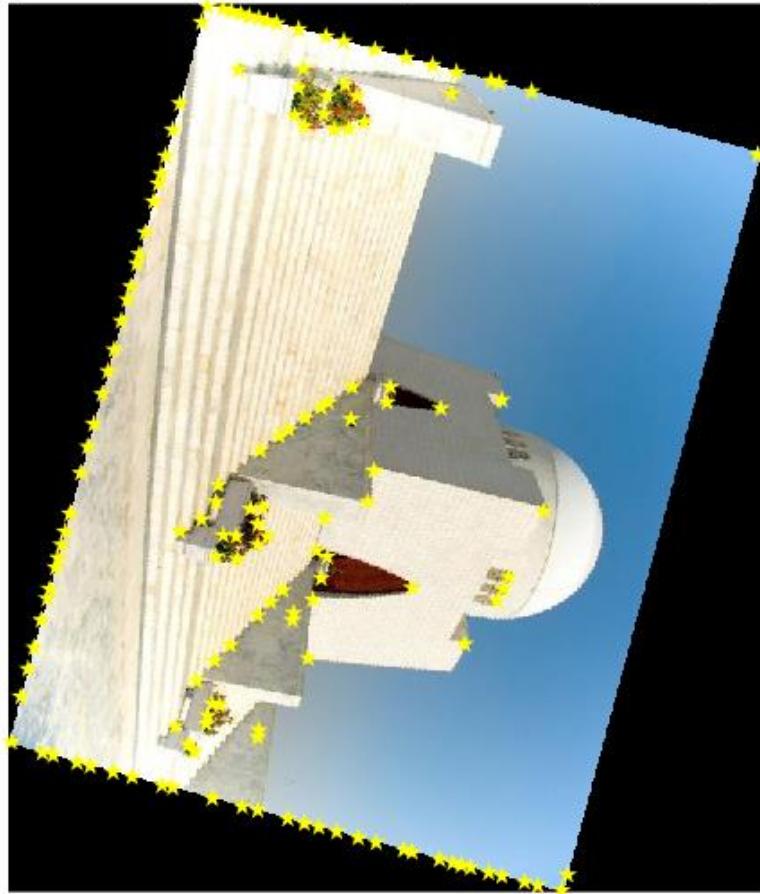
Rotated image with 225 degree where M = 21 , N = 71 and Repeatability M/N = 0.29577



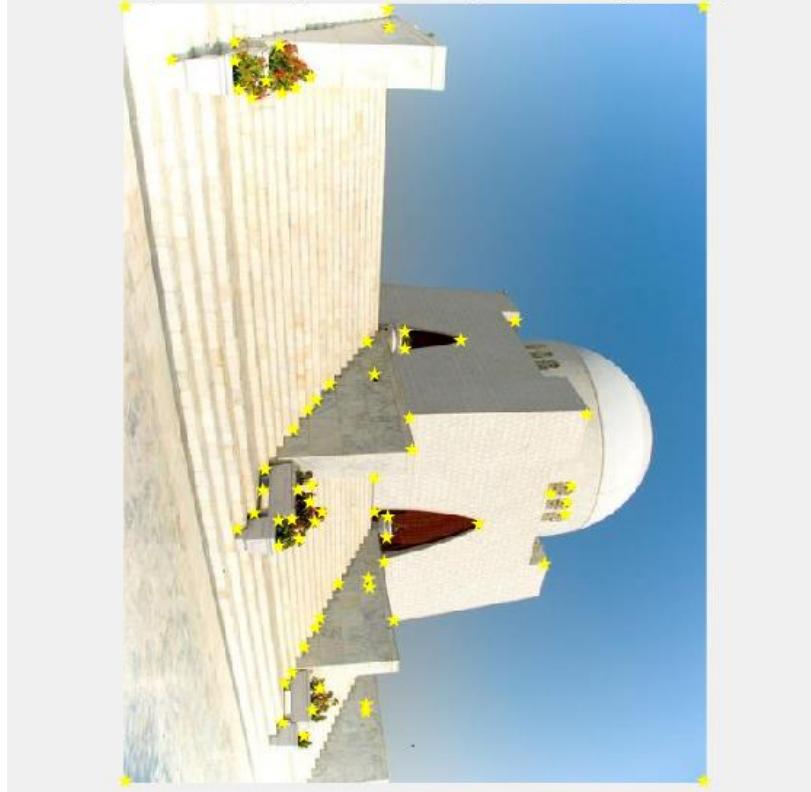
Rotated image with 240 degree where M = 14 , N = 71 and Repeatability M/N = 0.19718



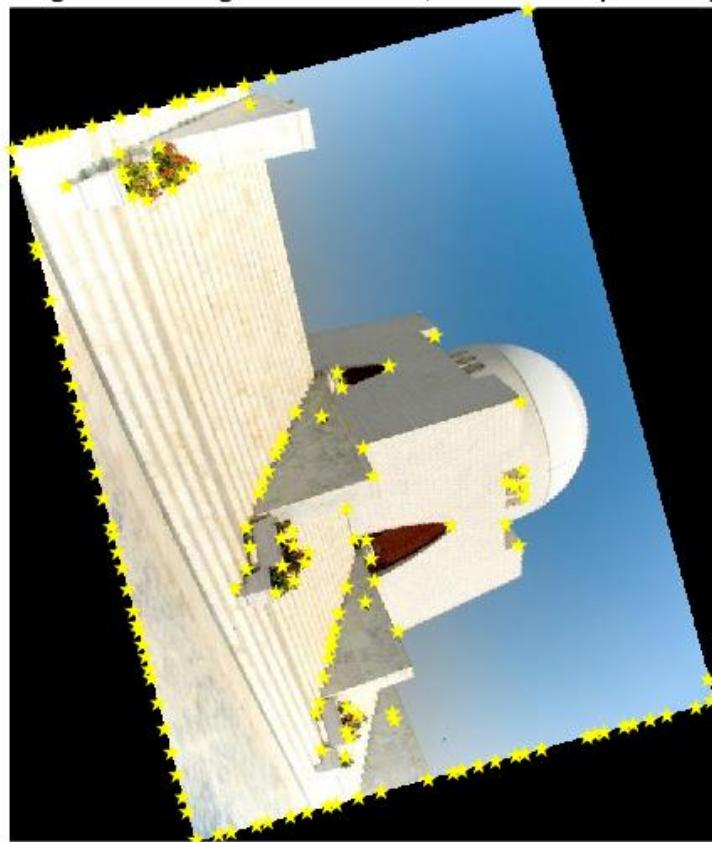
Rotated image with 255 degree where M = 6 , N = 71 and Repeatability M/N = 0.084507



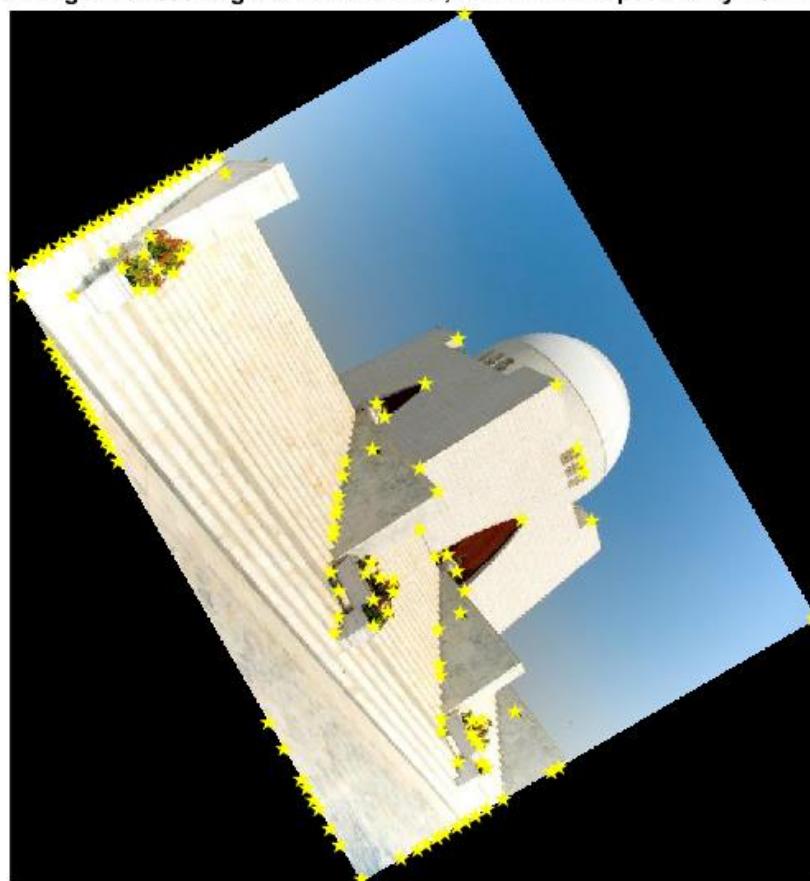
Rotated image with 270 degree where M = 0 , N = 71 and Repeatability M/N = 0



Rotated image with 285 degree where M = 0 , N = 71 and Repeatability M/N = 0



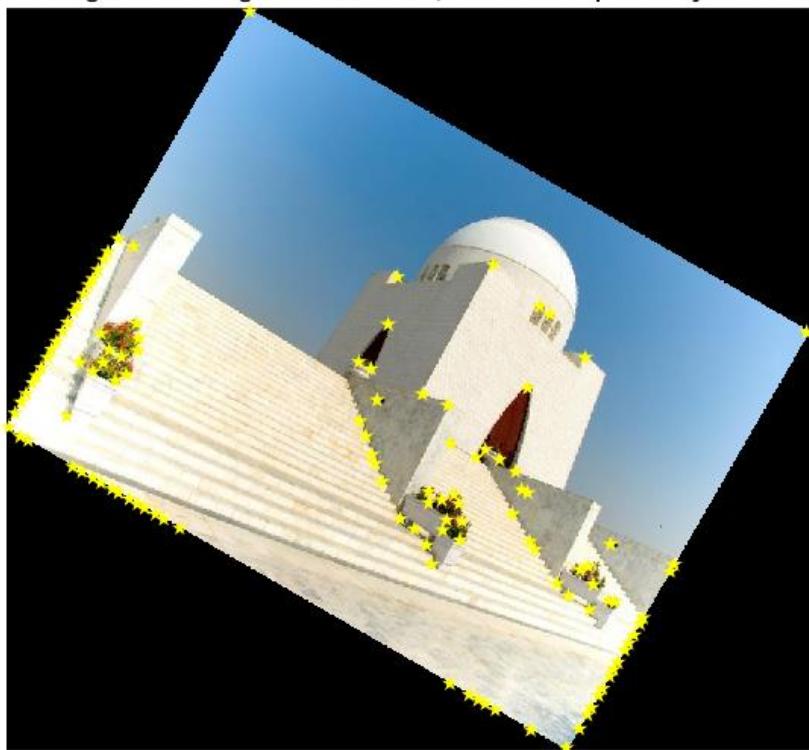
Rotated image with 300 degree where M = 16 , N = 71 and Repeatability M/N = 0.22535



Rotated image with 315 degree where M = 9 , N = 71 and Repeatability M/N = 0.12676



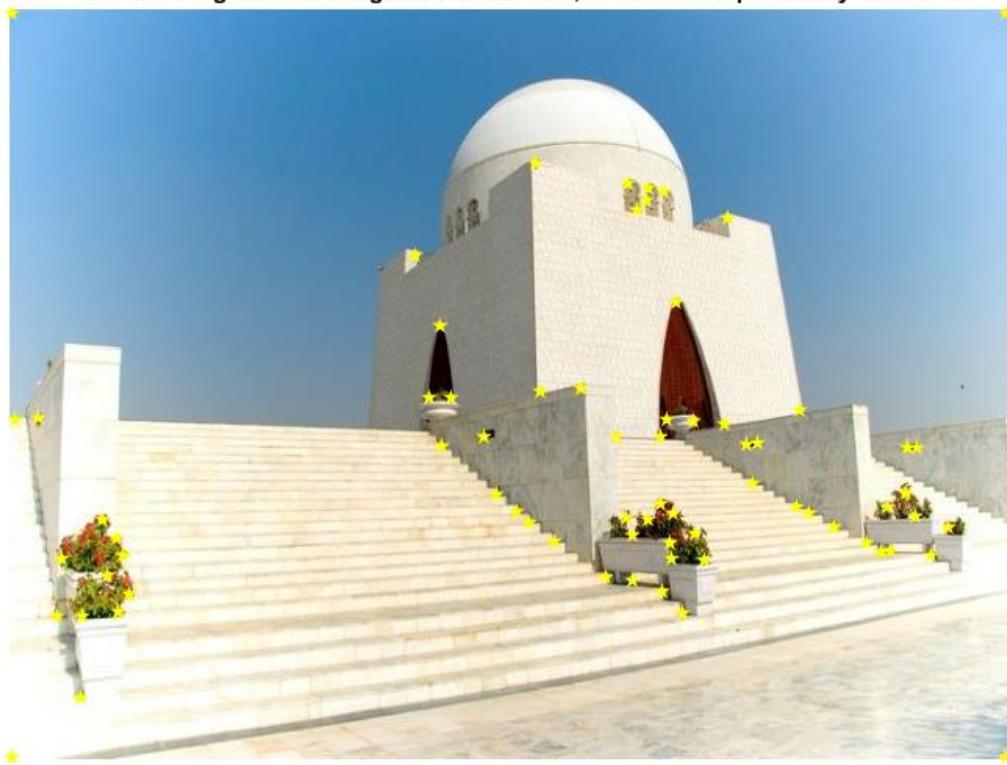
Rotated image with 330 degree where M = 24 , N = 71 and Repeatability M/N = 0.33803



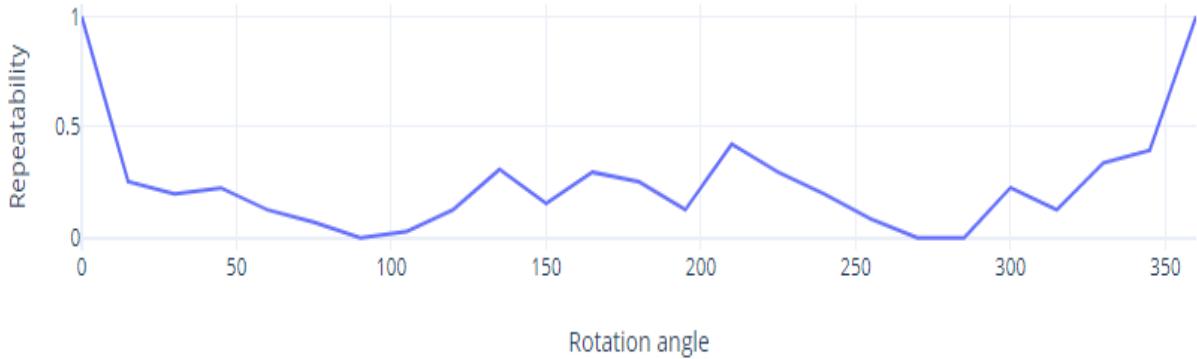
Rotated image with 345 degree where M = 28 , N = 71 and Repeatability M/N = 0.39437



Rotated image with 360 degree where M = 71 , N = 71 and Repeatability M/N = 1



Robustness of Harris Corner Detector to Rotation



Conclusion

For the second image we get almost the same result with slight difference in values but the overall pattern is same. Ideally, the repeatability plot for rotation must maintain the ratio at 0.95 but due to some unknown reason or inefficient programming, it falls as soon it gets rotated and couldn't locate the rotated keypoints. Harris detector is robust to rotation but here the results are not so favorable.

Observation # 02

The threshold value is set to be equal to that used in Harris corner detector in Part- A i.e. 4000. It was observed that number of matched corners (M) comes out to be equal to corners in original image (N), so, we get a repeatability factor equal to 1. It can be seen in the plot of repeatability vs rotation in the results section.

P.s. Ideally threshold value must have low value e.g. 60 or 80 when finding the keypoint in the rotated image that matches the ones in original image. Displaying the results for $T = 4000$ only as a finding whose reasoning couldn't be found out.

Code snippet

```
34
35      % PREDICTING IDEAL POSITION OF KEYPOINT
36 -      threshold = 4000;
37 -      for i=1:l:Ocpointssize
38 -          % KEYPOINT ROTATED BY TRANSFORMATION MATRIX
39 -          RotatedP = RotMatrix*((Ocpoints(i))'-Center0)+CenterR;
40 -          % EUCLIDEAN DISTANCE BTW TRANSFORMED & ROTATED IMAGE KEYPOINT
41 -          Euc_D = sqrt(sum((RpPoints(i))-RotatedP).^2));
42 -          if(Euc_D<threshold) % IF DISTANCE LESS THAN THRESHOLD, THEN
43 -              M=M+1; % INCREMENT IN COUNT OF MATCHED KEYPOINTS
44 -          end
45 -      end
46
```

Rotated image with 0 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 15 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 30 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 45 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 60 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 75 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 90 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 105 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 120 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 135 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 150 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 165 degree where M = 152 , N = 152 and Repeatability M/N = 1



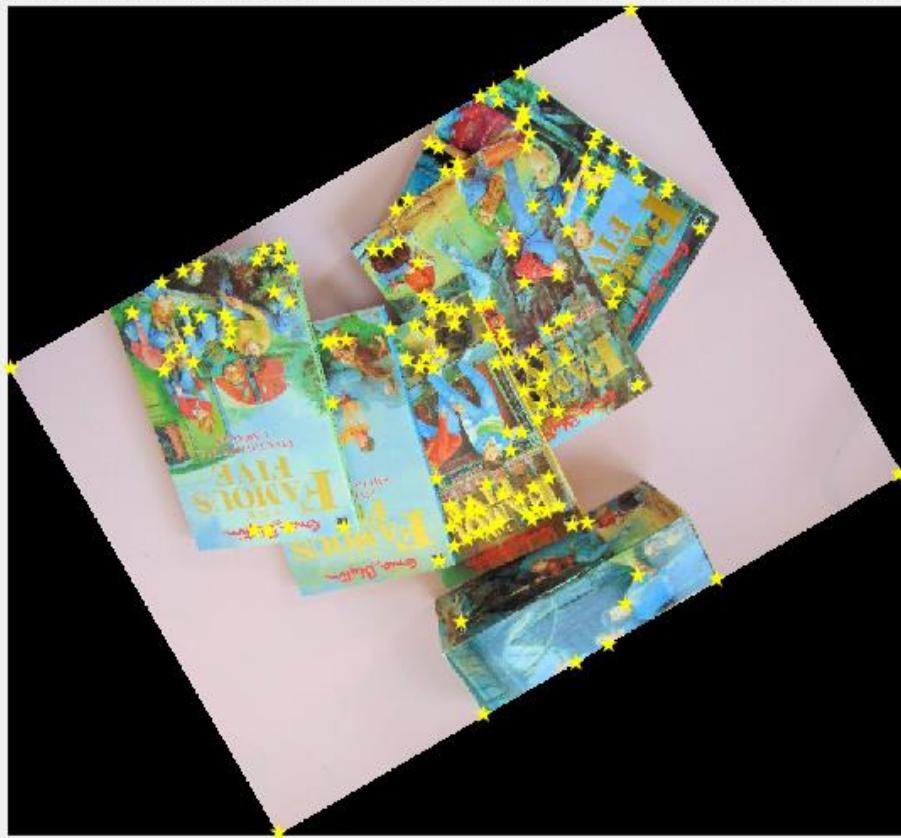
Rotated image with 180 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 195 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 210 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 225 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 240 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 255 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 270 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 285 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 300 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 315 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 330 degree where M = 152 , N = 152 and Repeatability M/N = 1



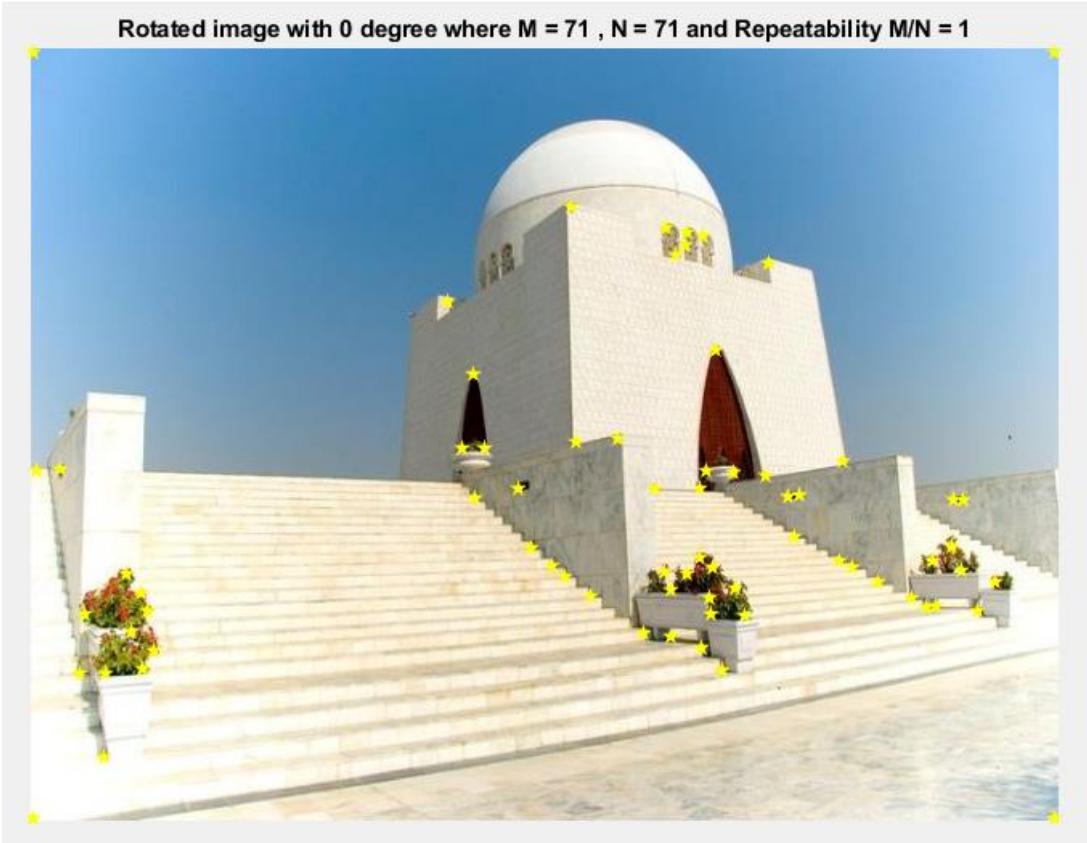
Rotated image with 345 degree where M = 152 , N = 152 and Repeatability M/N = 1



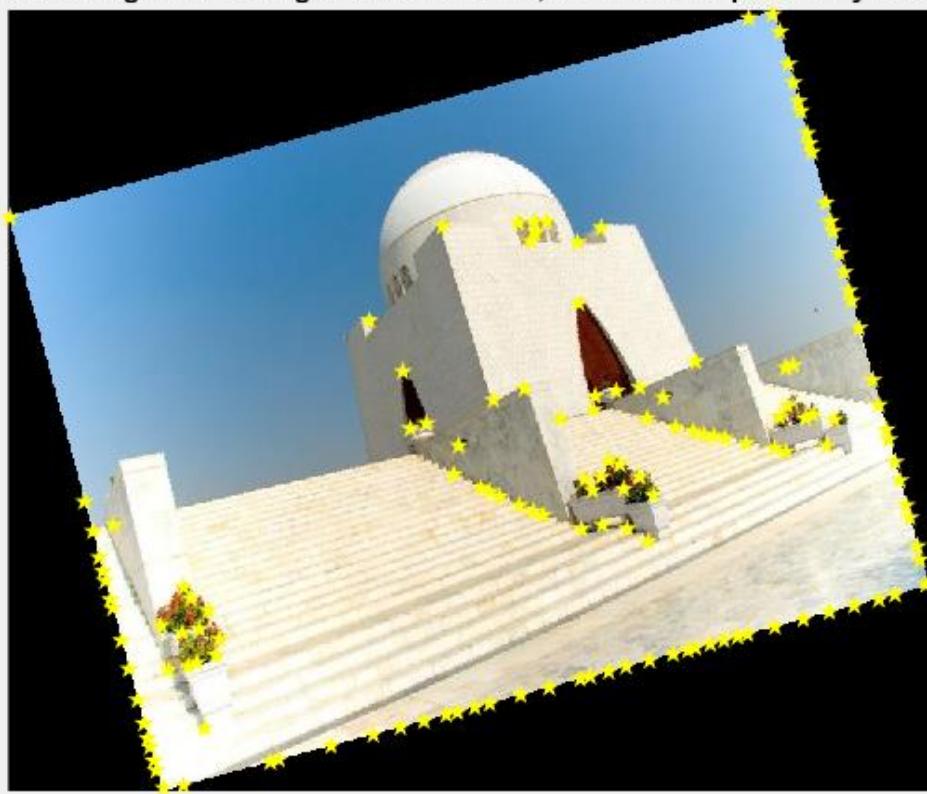
Rotated image with 360 degree where M = 152 , N = 152 and Repeatability M/N = 1



Rotated image with 0 degree where M = 71 , N = 71 and Repeatability M/N = 1



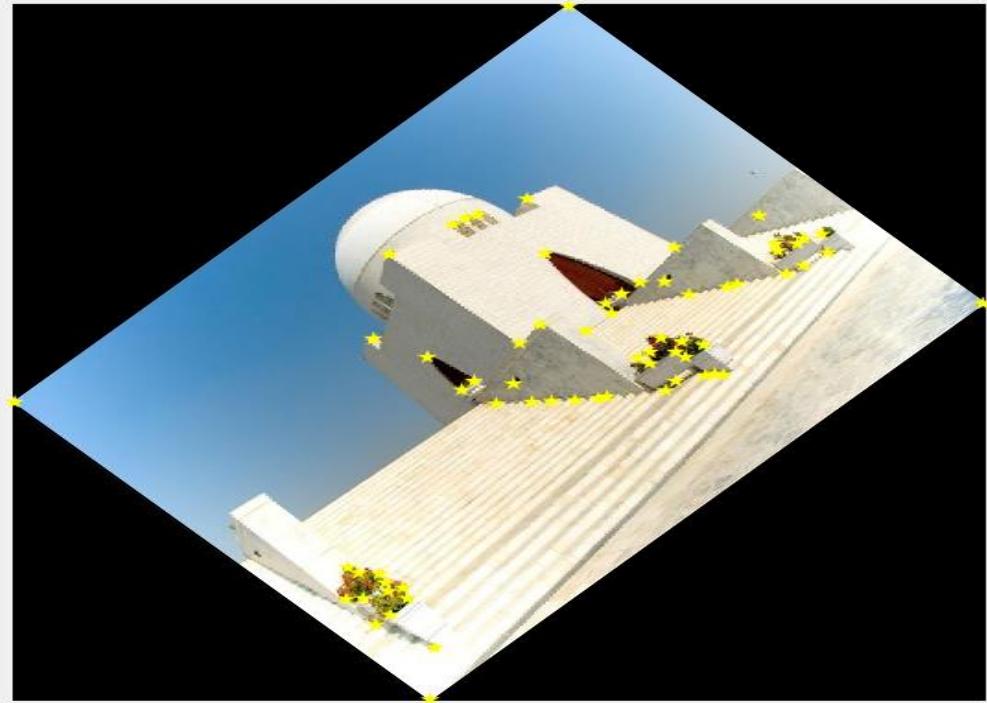
Rotated image with 15 degree where M = 71 , N = 71 and Repeatability M/N = 1



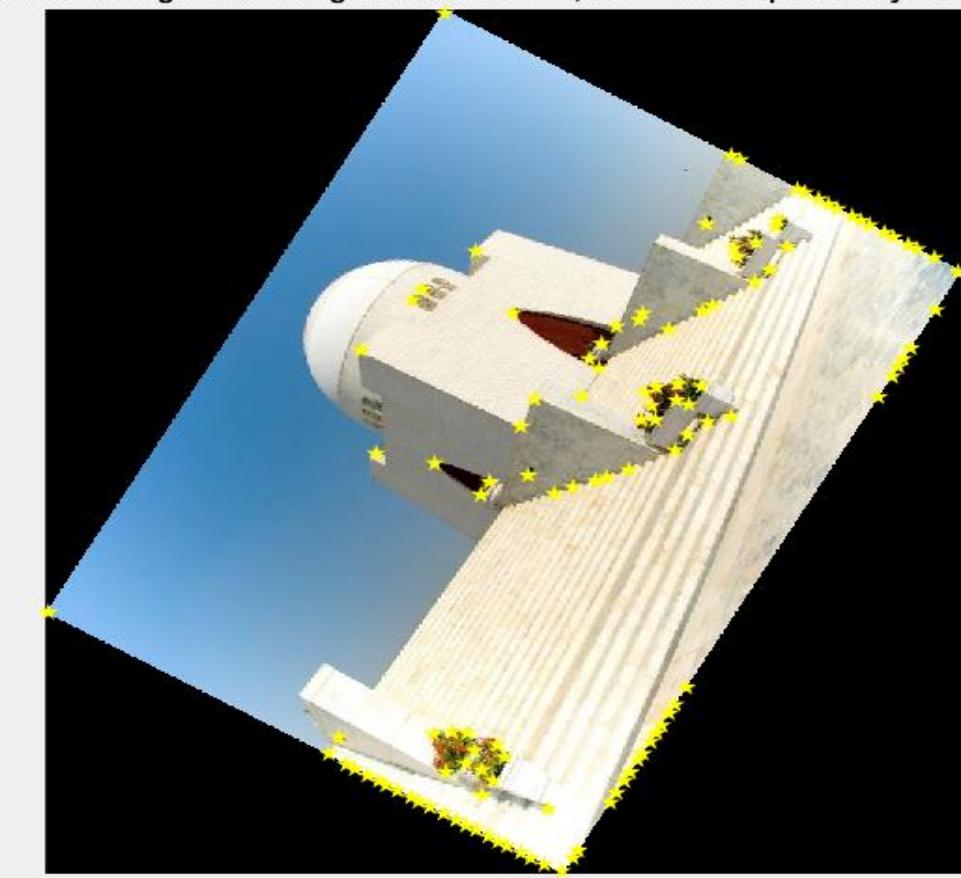
Rotated image with 30 degree where M = 71 , N = 71 and Repeatability M/N = 1



Rotated image with 45 degree where M = 71 , N = 71 and Repeatability M/N = 1



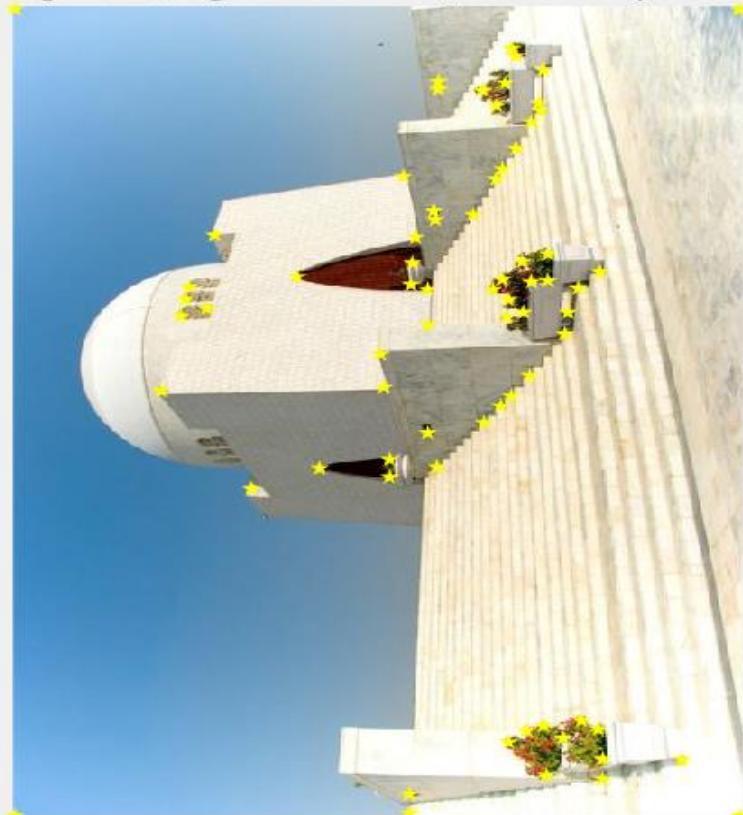
Rotated image with 60 degree where M = 71 , N = 71 and Repeatability M/N = 1



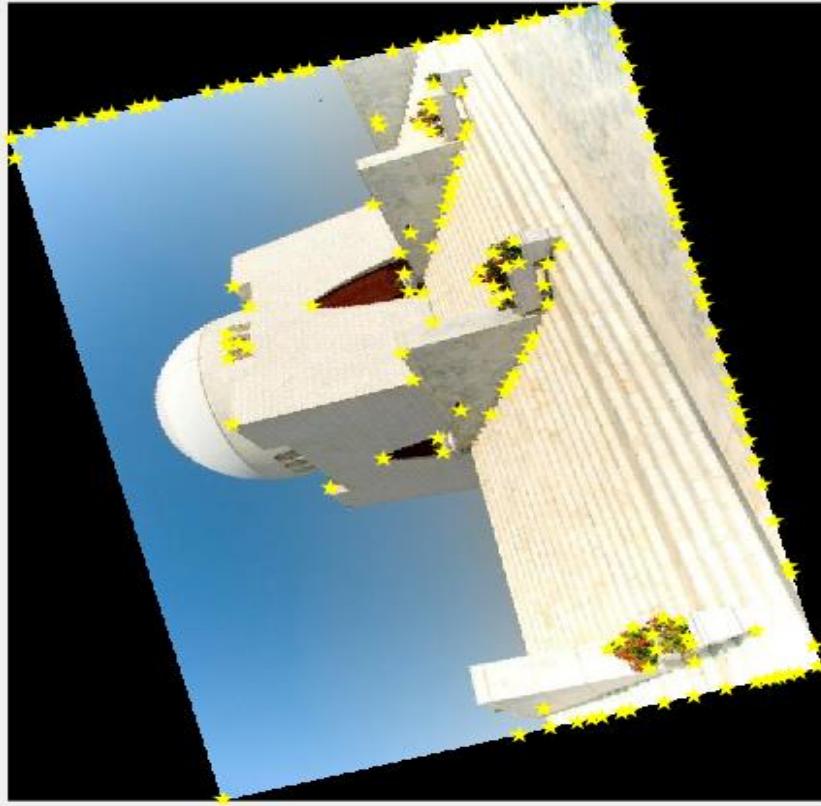
Rotated image with 75 degree where M = 71 , N = 71 and Repeatability M/N = 1



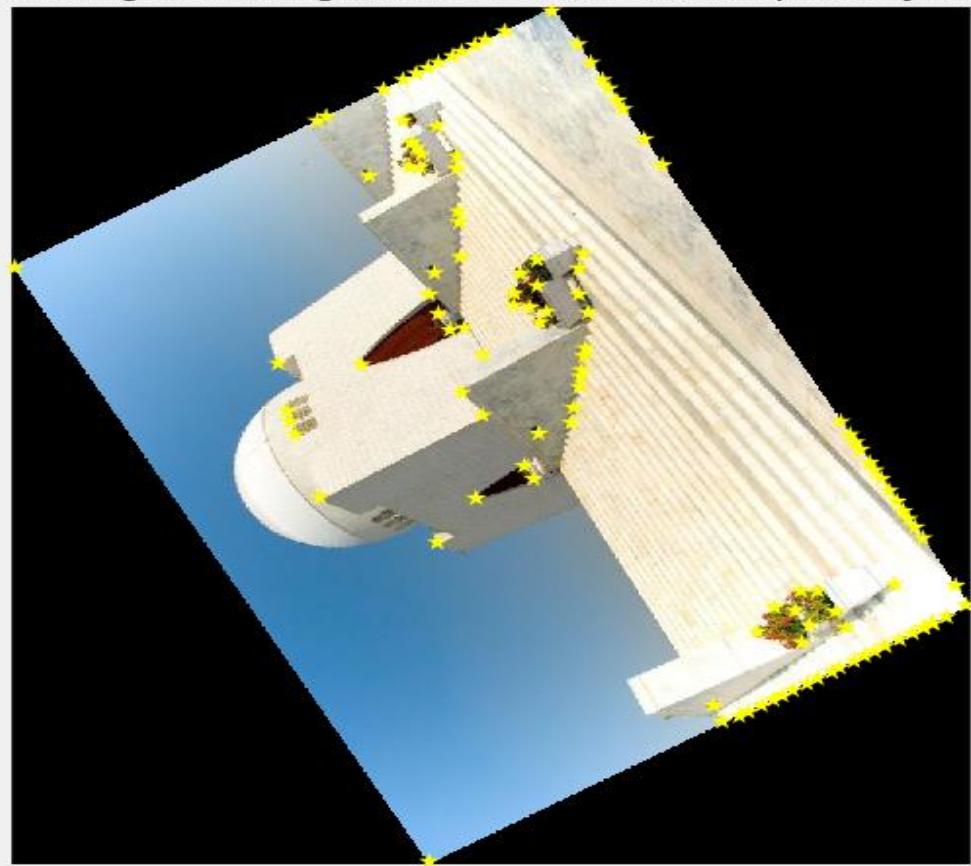
Rotated image with 90 degree where M = 71 , N = 71 and Repeatability M/N = 1



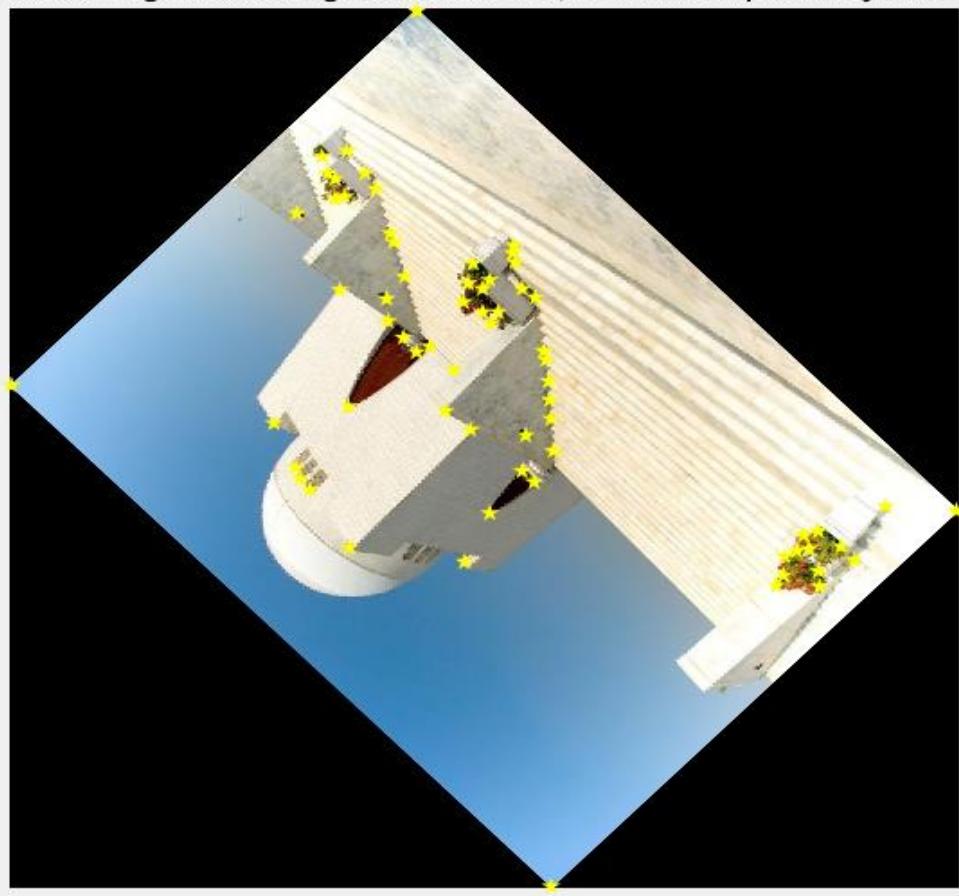
Rotated image with 105 degree where M = 71 , N = 71 and Repeatability M/N = 1



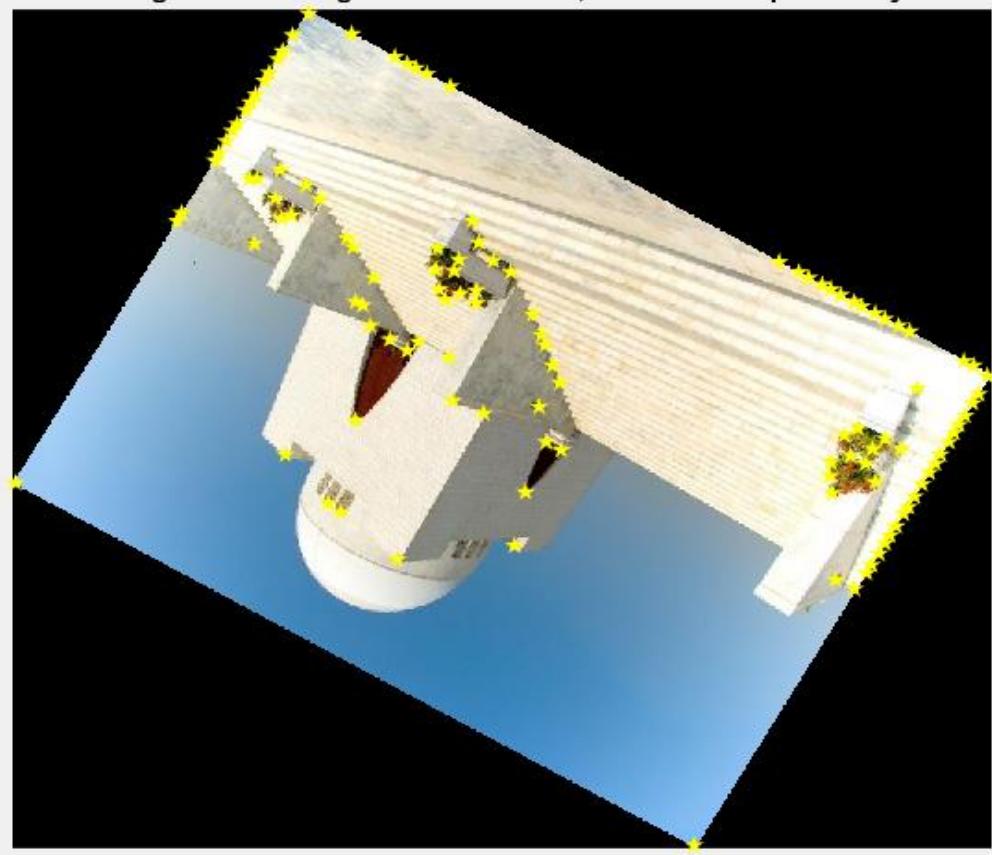
Rotated image with 120 degree where M = 71 , N = 71 and Repeatability M/N = 1



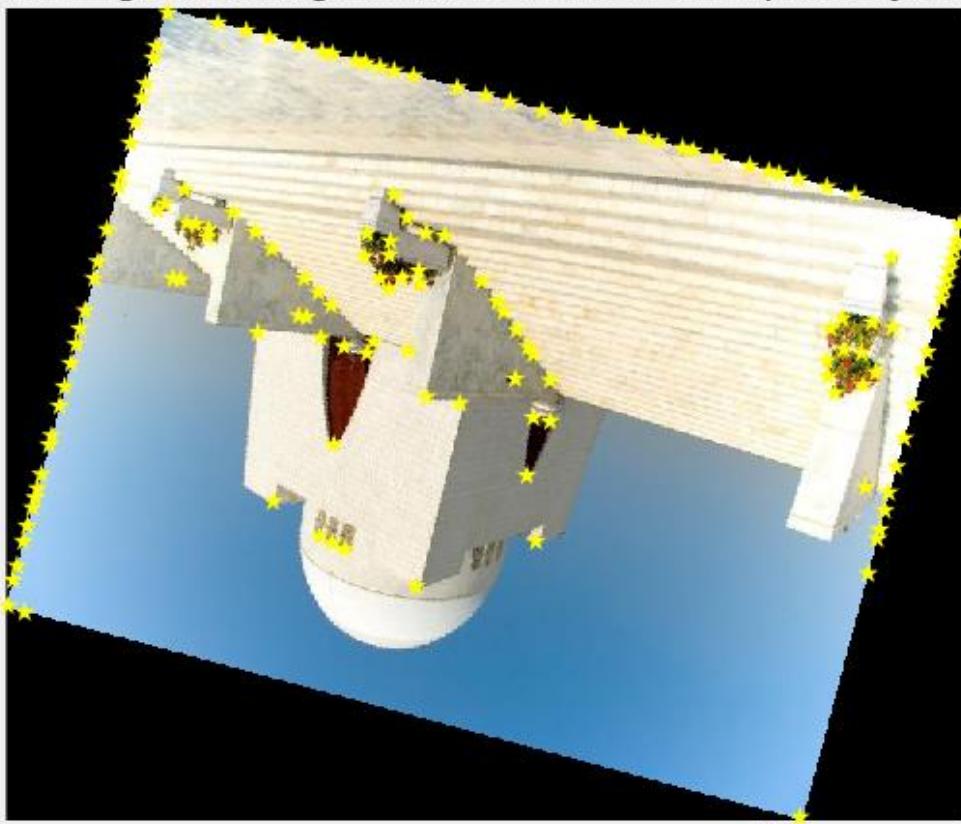
Rotated image with 135 degree where M = 71 , N = 71 and Repeatability M/N = 1



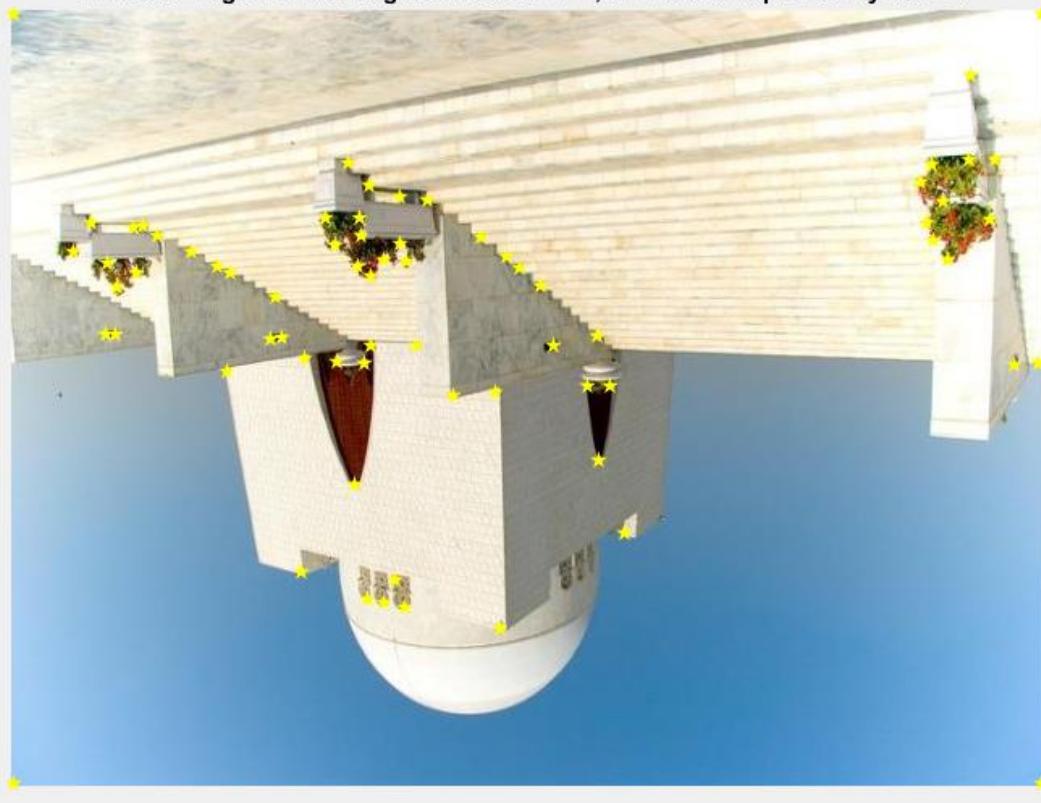
Rotated image with 150 degree where M = 71 , N = 71 and Repeatability M/N = 1



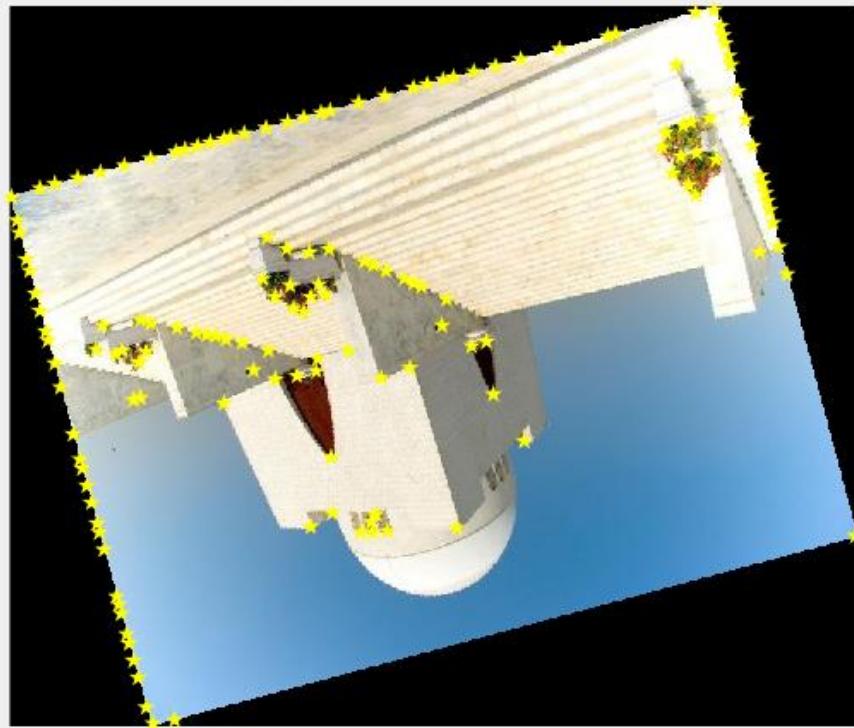
Rotated image with 165 degree where M = 71 , N = 71 and Repeatability M/N = 1



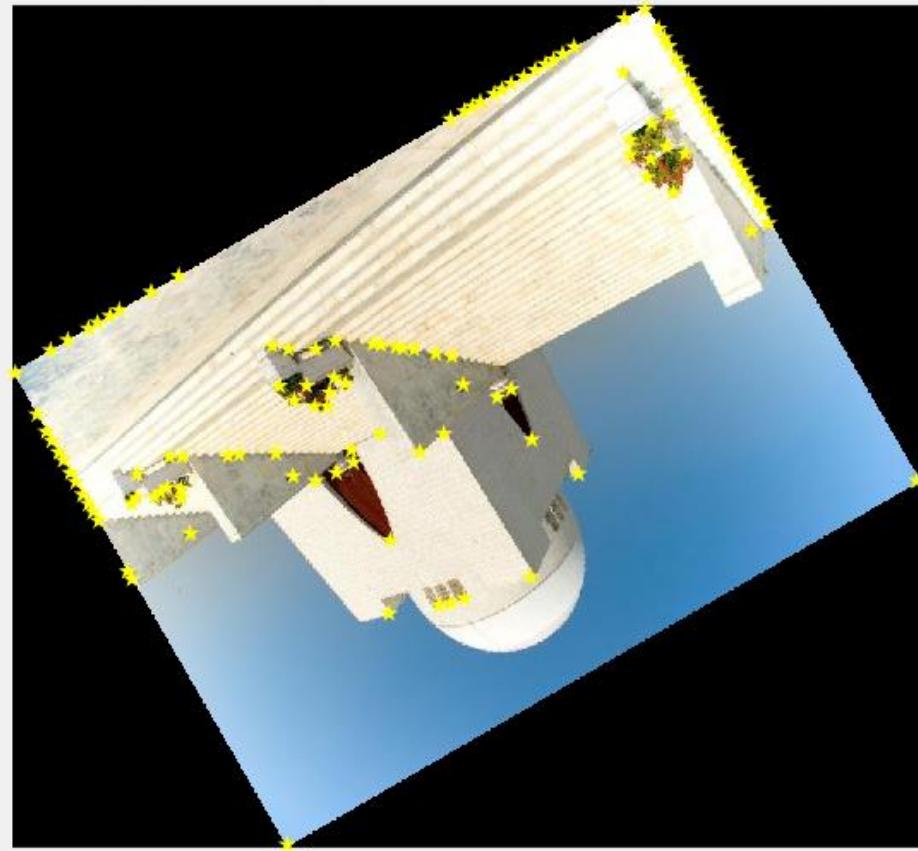
Rotated image with 180 degree where M = 71 , N = 71 and Repeatability M/N = 1



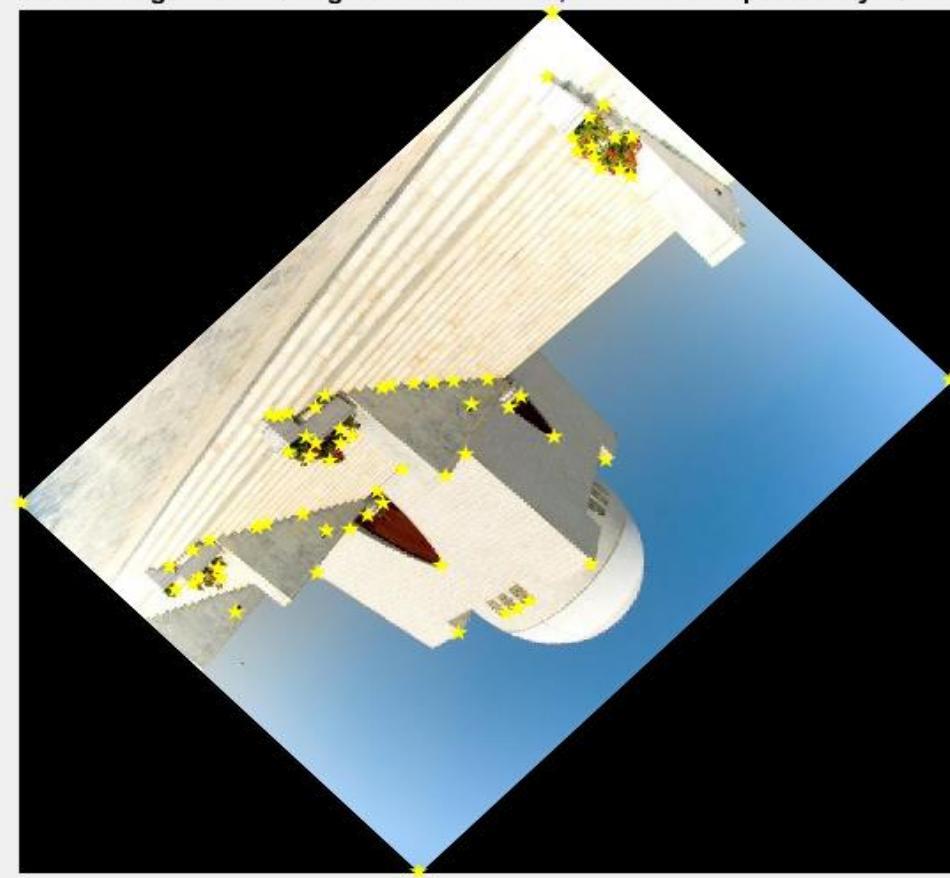
Rotated image with 195 degree where M = 71 , N = 71 and Repeatability M/N = 1



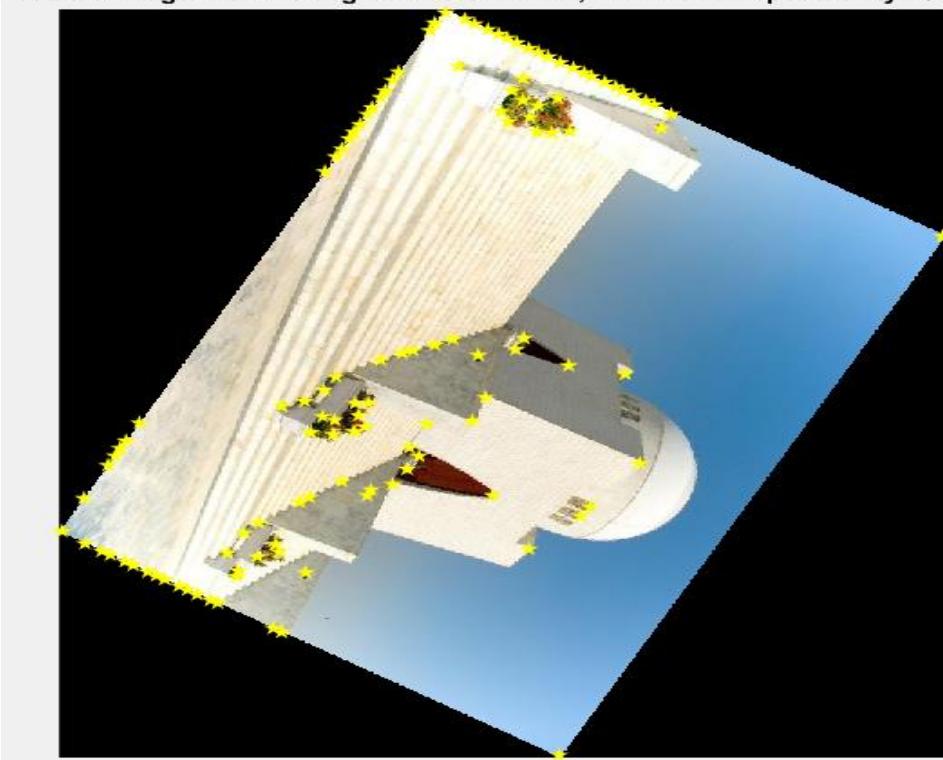
Rotated image with 210 degree where M = 71 , N = 71 and Repeatability M/N = 1



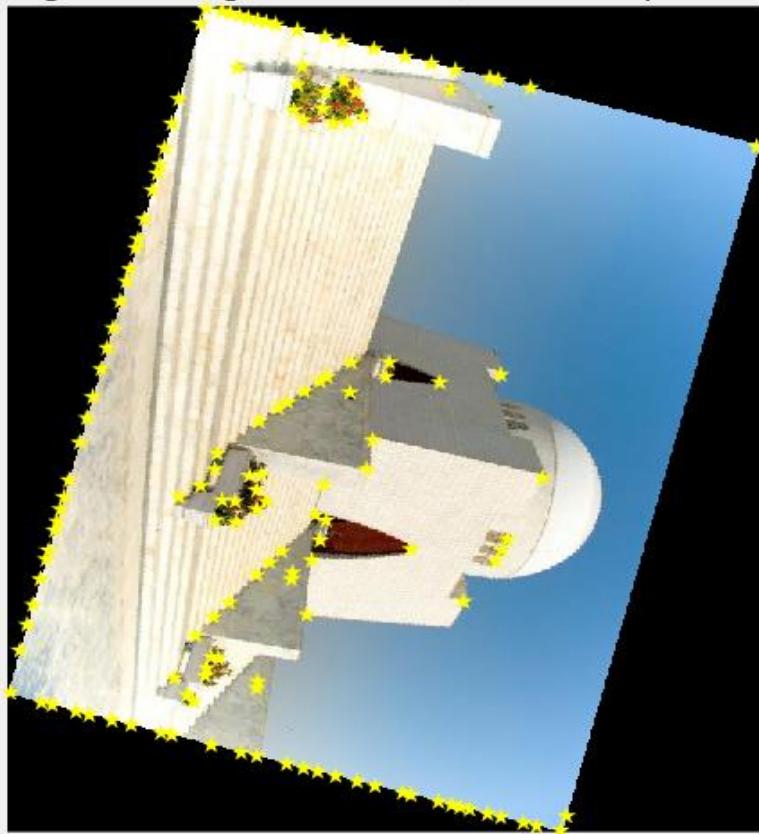
Rotated image with 225 degree where M = 71 , N = 71 and Repeatability M/N = 1



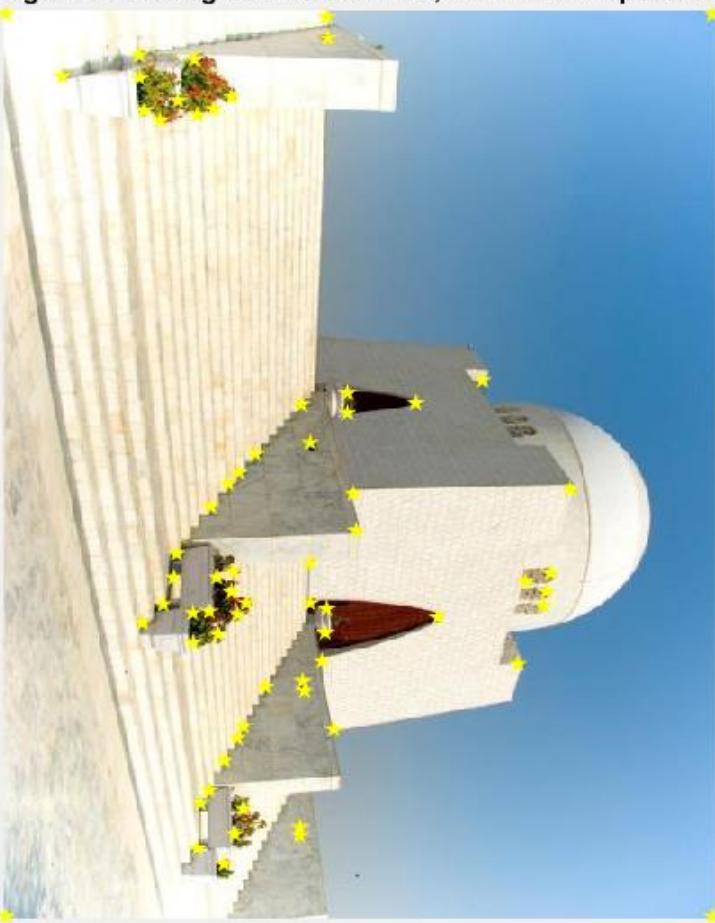
Rotated image with 240 degree where M = 71 , N = 71 and Repeatability M/N = 1



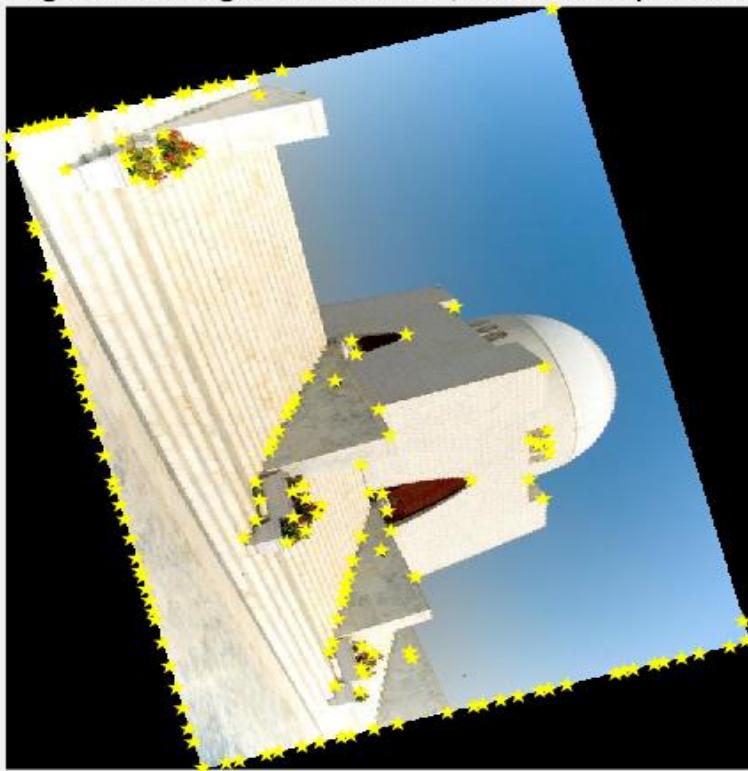
Rotated image with 255 degree where M = 71 , N = 71 and Repeatability M/N = 1



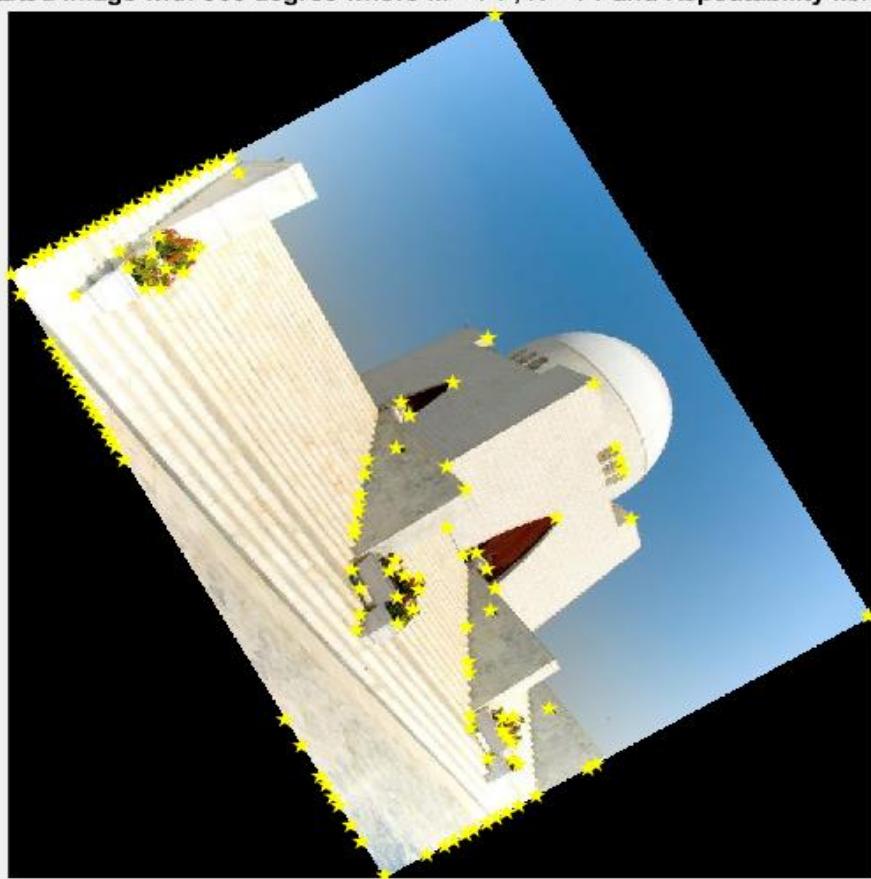
Rotated image with 270 degree where M = 71 , N = 71 and Repeatability M/N = 1



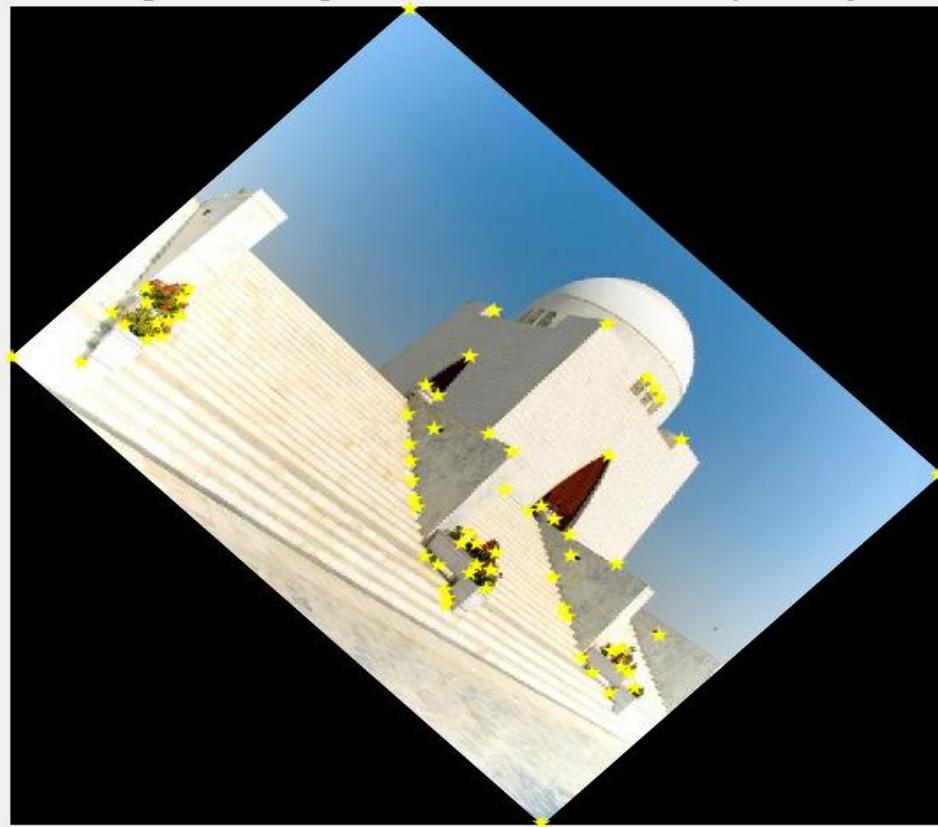
Rotated image with 285 degree where M = 71 , N = 71 and Repeatability M/N = 1



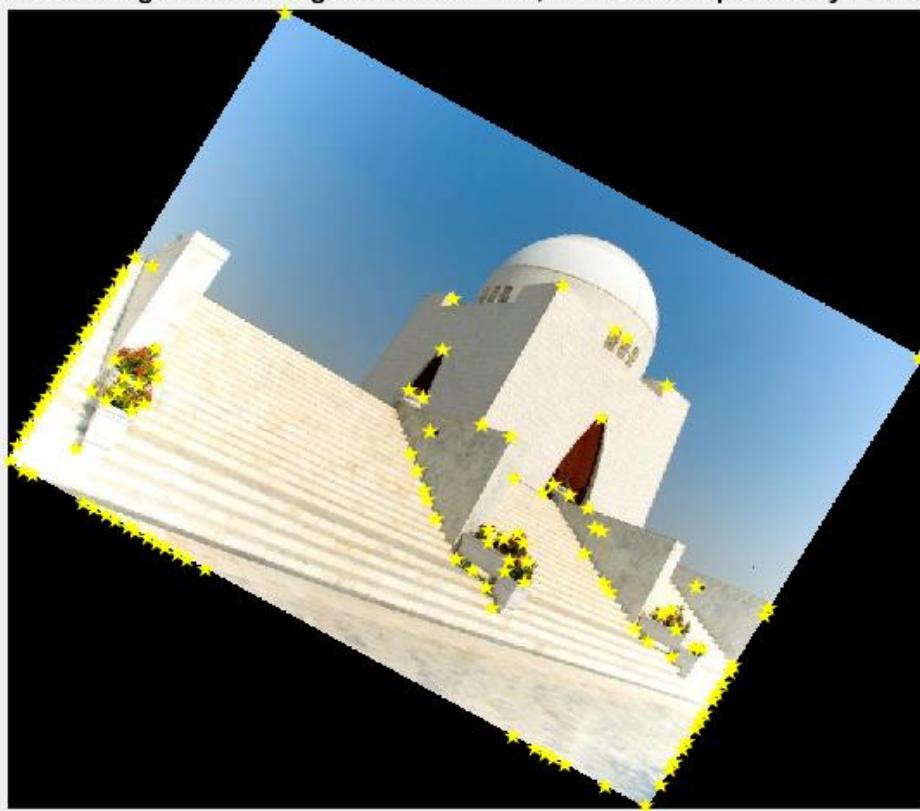
Rotated image with 300 degree where M = 71 , N = 71 and Repeatability M/N = 1



Rotated image with 315 degree where M = 71 , N = 71 and Repeatability M/N = 1



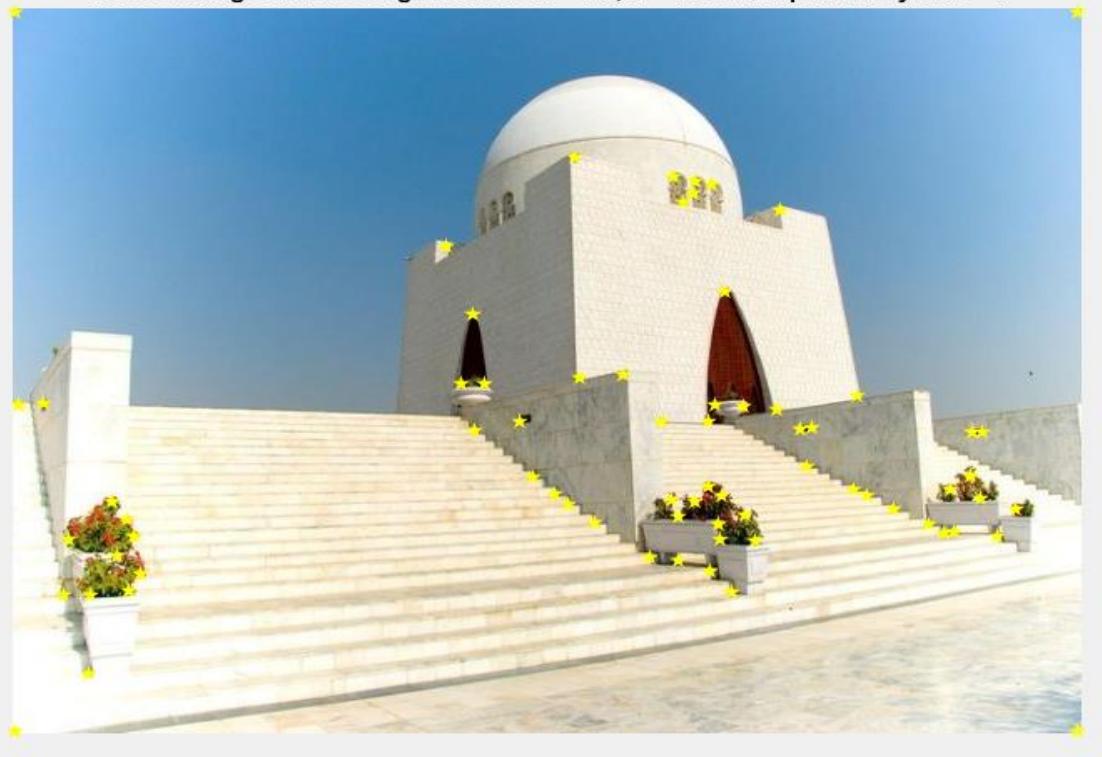
Rotated image with 330 degree where M = 71 , N = 71 and Repeatability M/N = 1



Rotated image with 345 degree where M = 71 , N = 71 and Repeatability M/N = 1

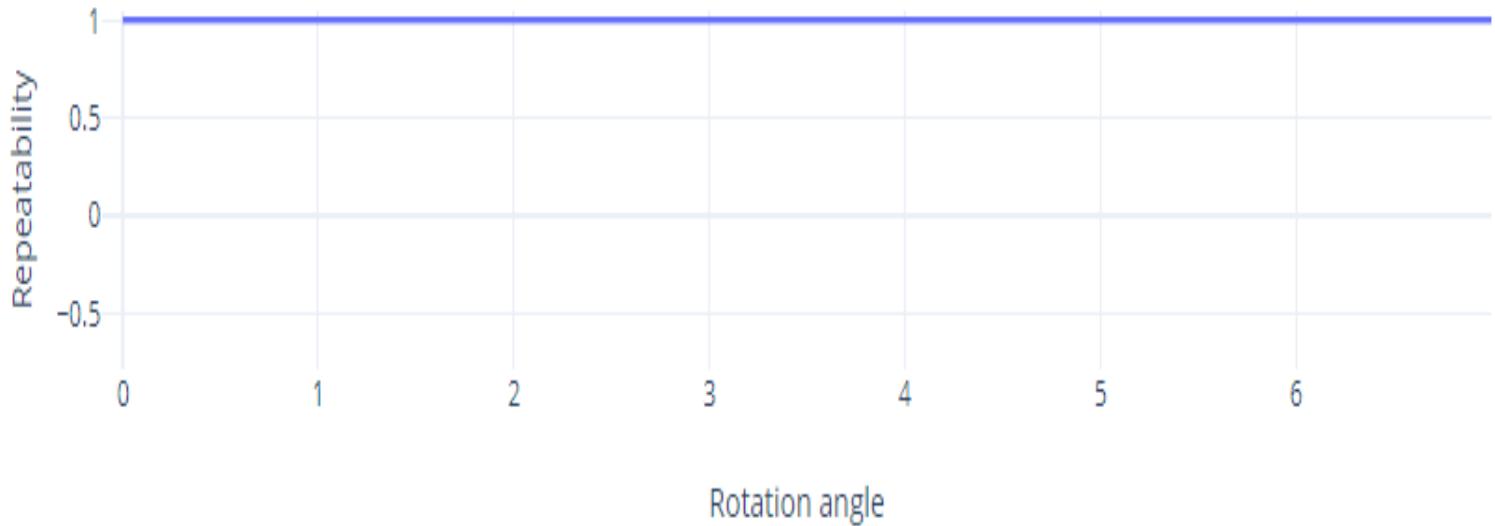


Rotated image with 360 degree where M = 71 , N = 71 and Repeatability M/N = 1



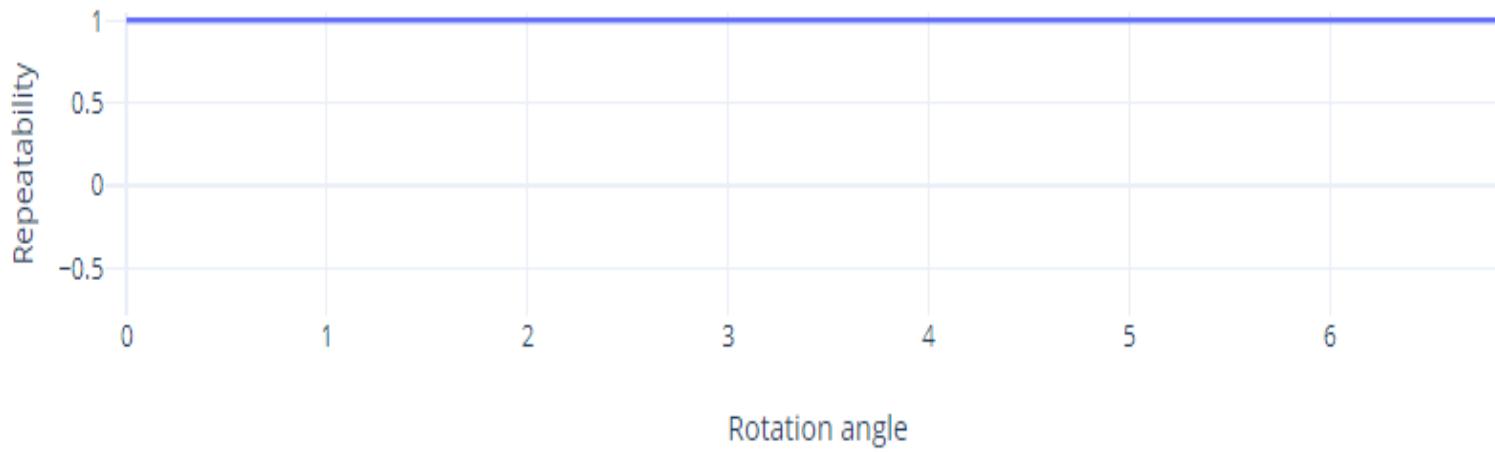
Repeatability vs Rotation Plot for the Image Famous Five

Robustness of Harris Corner Detector to Rotation



Repeatability vs Rotation Plot for the Image Mausoleum

Robustness of Harris Corner Detector to Rotation



Note: As a result of facing some error while plotting repeatability vs rotation, an online tool has been used. The link for online tool is given below.

<https://chart-studio.plotly.com/create/>

PART- C

Conduct an experiment analogous to part (b), but instead of rotating the image, resize the image by the scaling factors $m^0, m^1, m^2, \dots, m^8$, where $m = 1.2$. Compute Harris keypoints for each resized image. You are allowed to use built-in library functions to perform this scaling, preferably with bicubic interpolation. By comparing Harris keypoints of the original and resized images, compute and plot repeatability against scaling factor. Comment on the Harris keypoint detector's robustness against scale changes.

The following built-in functions are used for computations:

- `sf=(1.2)^sfp`, scaling factor for image
- `RimageI = imresize(Oimagek,sf)` for resizing image to different scales

Code

```
1 % CS-867 COMPUTER VISION
2 % ASSIGNMENT-1, PART-C
3 % ROBUSTNESS OF HARRIS KEYPOINT DETECTOR TO SCALING
4
5 OimageI = imread('famous_five.png');
6 scale_eff(OimageI);
7 function scale_eff(Oimagek)
8 [Orows,Ocols] = harris(Oimagek);
9 Ocpoints = [Orows,Ocols];
10
11 for sfp=0:1:8 % RANGE OF SCALING FACTOR FROM 0 TO 8
12 sf=(1.2)^sfp; %SCALING FACTOR FOR IMAGE
13 RimageI = imresize(Oimagek,sf); %RESIZE IMAGE WITH SCALE FACTOR
14 [Rrows,Rcols] = harris(RimageI);
15 Rcpoints = [Rrows,Rcols];
16 threshold = 4000;
17 Ocpointssize = length(Ocpoints);
18 N=length(Ocpoints); % HARRIS CORNERS IN ORIGINAL IMAGE
19 rot_ang=15*pi/180; % ROTATION ANGLE FOR IMAGE
20 % TRANSFORMATION MATRIX FOR ROTATION OF KEYPOINT
21 RotMatrix = [cosd(rot_ang) -sind(rot_ang);sind(rot_ang) cosd(rot_ang)];
22 % RETURNS THE CENTER OF ORIGINAL IMAGE (GRAYSCALE)
23 CenterO = (size(rgb2gray(Oimagek))/2)';
24 % RETURNS THE CENTER OF ROTATED IMAGE (GRAYSCALE)
25 CenterR = (size(rgb2gray(RimageI))/2)';
26 M=0; % COUNT OF MATCHED CORNERS
```

```

28
29          % PREDICTING IDEAL POSITION OF KEYPOINT
30 -    for i=1:1:Ocpointssize
31          % KEYPOINT ROTATED BY TRANSFORMATION MATRIX
32          RotatedP = RotMatrix*((Ocpoints(i))-CenterO)+CenterR;
33          % EUCLIDEAN DISTANCE BTW TRANSFORMED & ROTATED IMAGE KEYPOINT
34          Euc_D = sqrt(sum((Rcpoints(i)-RotatedP).^2));
35          if(Euc_D<threshold) % IF DISTANCE LESS THAN THRESHOLD, THEN
36              M=M+1; % INCREMENT IN COUNT OF MATCHED KEYPOINTS
37          end
38      end
39      % REPEATABILITY MEASURE, RATIO OF MATCHED CORNERS TO ORIGINAL CORNERS
40      rep=M/N;
41      figure,imshow(Rimage1),hold on,
42          plot(Rcols,Rrows,'yp','MarkerFaceColor','y','MarkerSize',15),
43          title(['RESIZED IMAGE WITH m^',num2str(sfpl),', M =',num2str(M),
44          ' & N =',num2str(N),' REPEATABILITY FACTOR,M/N =',num2str(rep)]);
45      end
46  end
47
48      % HARRIS KEYPOINT DETECTOR AS IN PART-A
49  function [R,C] = harris(image)
50      im = rgb2gray(image); % CONVERTING GIVEN IMAGE FROM RGB TO GRayscale
51      [dx,dy]=meshgrid(-1:1, -1:1);
52      ix = conv2(double(im),dx,'same'); % HORIZONTAL GRADIENT / VERTICAL EDGES
53      iy = conv2(double(im),dy,'same'); % VERTICAL GRADIENT / HORIZONTAL EDGES
54
55      % PARAMETERS FOR GAUSSIAN FILTER
56      sigma = 3;
57      radius=1;
58      order = (2*radius+1)^2;
59
60      % DEFINING GAUSSIAN FILTER
61      len = max(1,fix(6*sigma));
62      p=len; q=len;
63      [ul,u2]=meshgrid(-(p-1)/2:(p-1)/2, -(q-1)/2: (q-2)/2);
64      ug = exp(-(ul.^2+u2.^2)/(2*sigma^2));
65      [w,z] = size(ug);
66      sum = 0;
67      for i=1:w
68          for j=1:z
69              sum = sum+ug(i,j);
70          end
71      end
72      G = ug ./sum;
73

```

```

73
74      % COMPUTING ELEMENTS OF SECOND MOMENT MATRIX, M
75 -  Ix2 = conv2(double(ix.^2),G,'same');
76 -  Iy2 = conv2(double(iy.^2),G,'same');
77 -  Ixy = conv2(double(ix.*iy),G,'same');
78
79      % CORNERNESS MEASURE
80 -  r = (Ix2.*Iy2 - Ixy.^2)./(Ix2+Iy2 + eps);
81
82      % Value of Threshold set empirically
83 -  threshold = 4000;
84
85      % FINDING MAX POINT FOR NON-MAX SUPPRESSION
86 -  maximum_point = ordfilt2(r, order^2,ones(order));
87
88      % FINDING CORNERS
89 -  harris_corners = (r==maximum_point) & (r>threshold);
90 -  [R,C]=find(harris_corners);
91 - end
92

```

Results

As we increase the scaling factor, the number of matched keypoints decreases. Hence, Harris detector is not robust to scaling.

RESIZED IMAGE WITH m^0 , M =152 & N =152 REPEATABILITY FACTOR,M/N =1



RESIZED IMAGE WITH m^1 , M =89 & N =152 REPEATABILITY FACTOR,M/N =0.58553



RESIZED IMAGE WITH m^2 , M =36 & N =152 REPEATABILITY FACTOR,M/N =0.23684



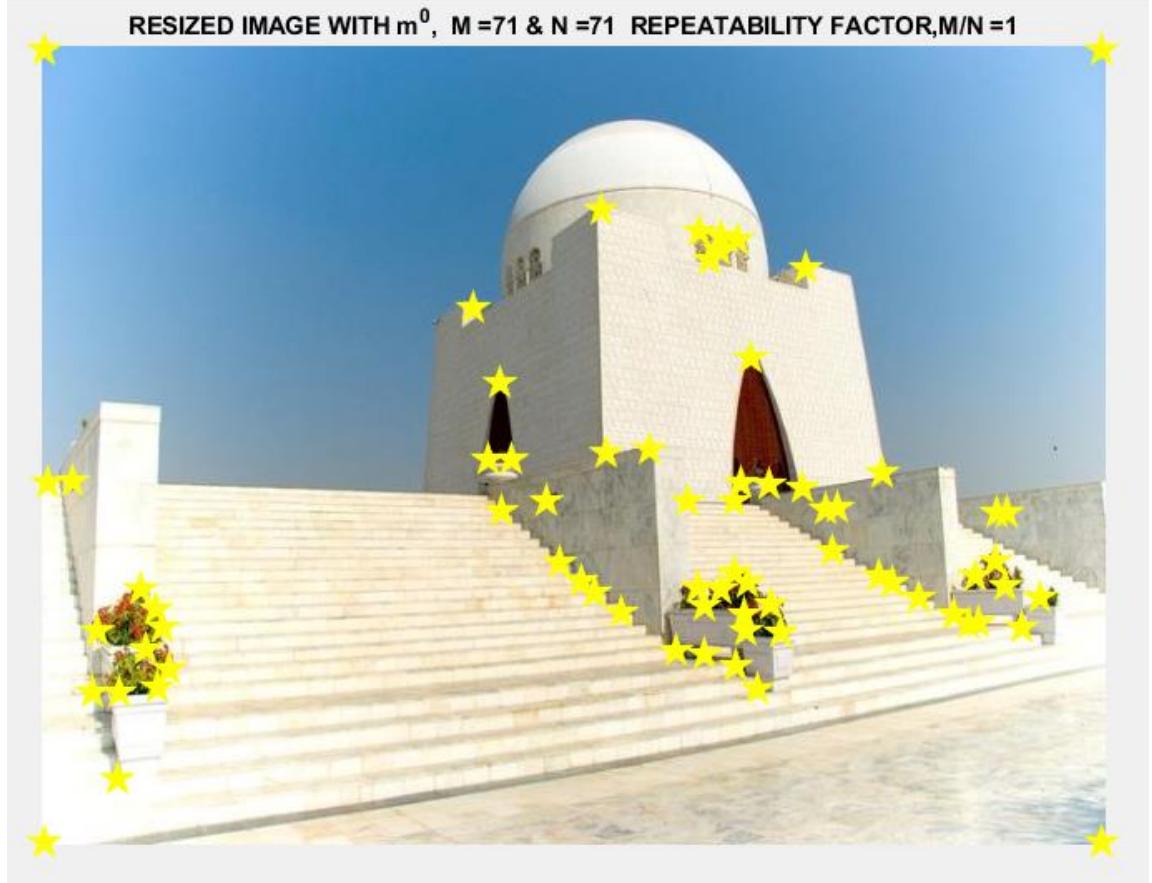
RESIZED IMAGE WITH m^3 , M =26 & N =152 REPEATABILITY FACTOR,M/N =0.17105



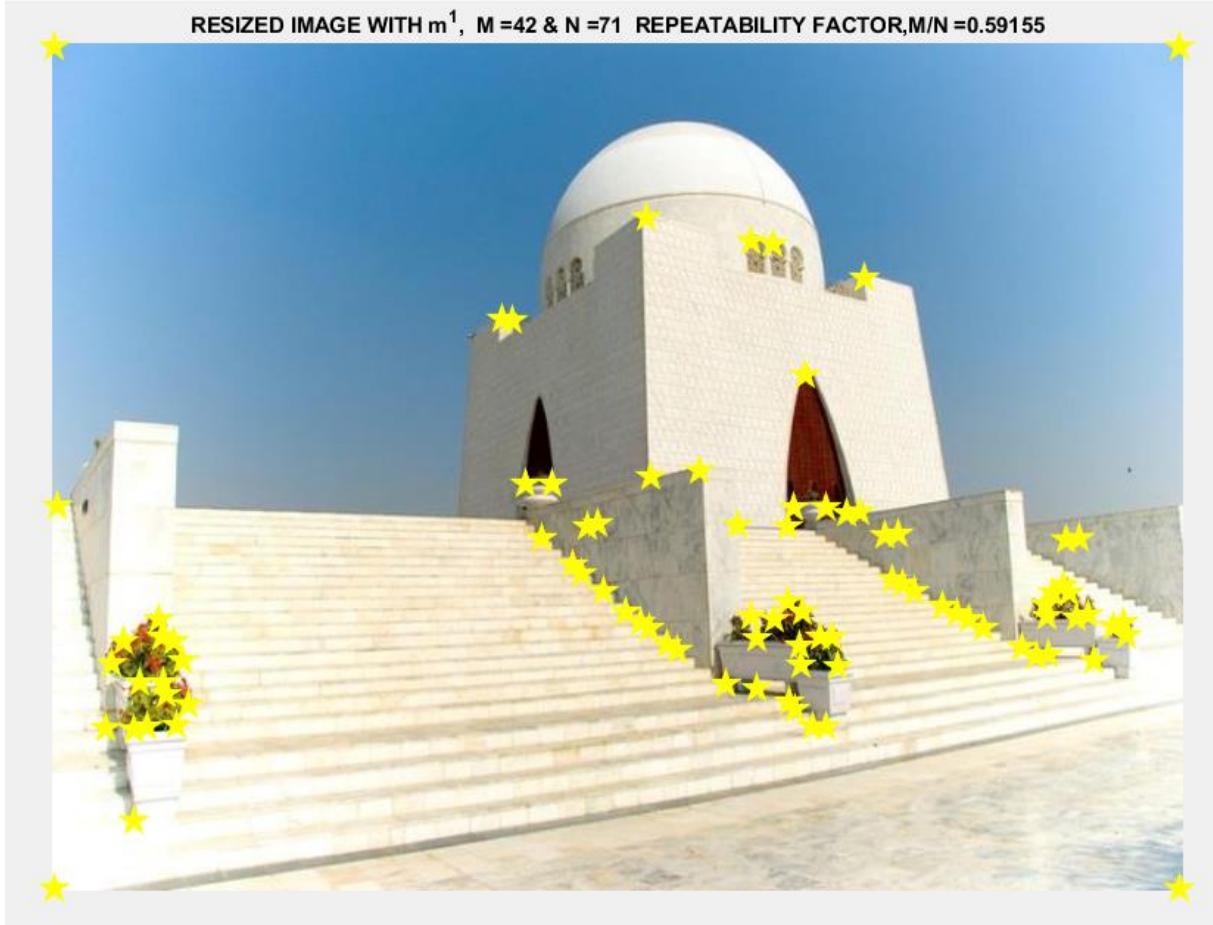
RESIZED IMAGE WITH m^4 , M =24 & N =152 REPEATABILITY FACTOR,M/N =0.15789



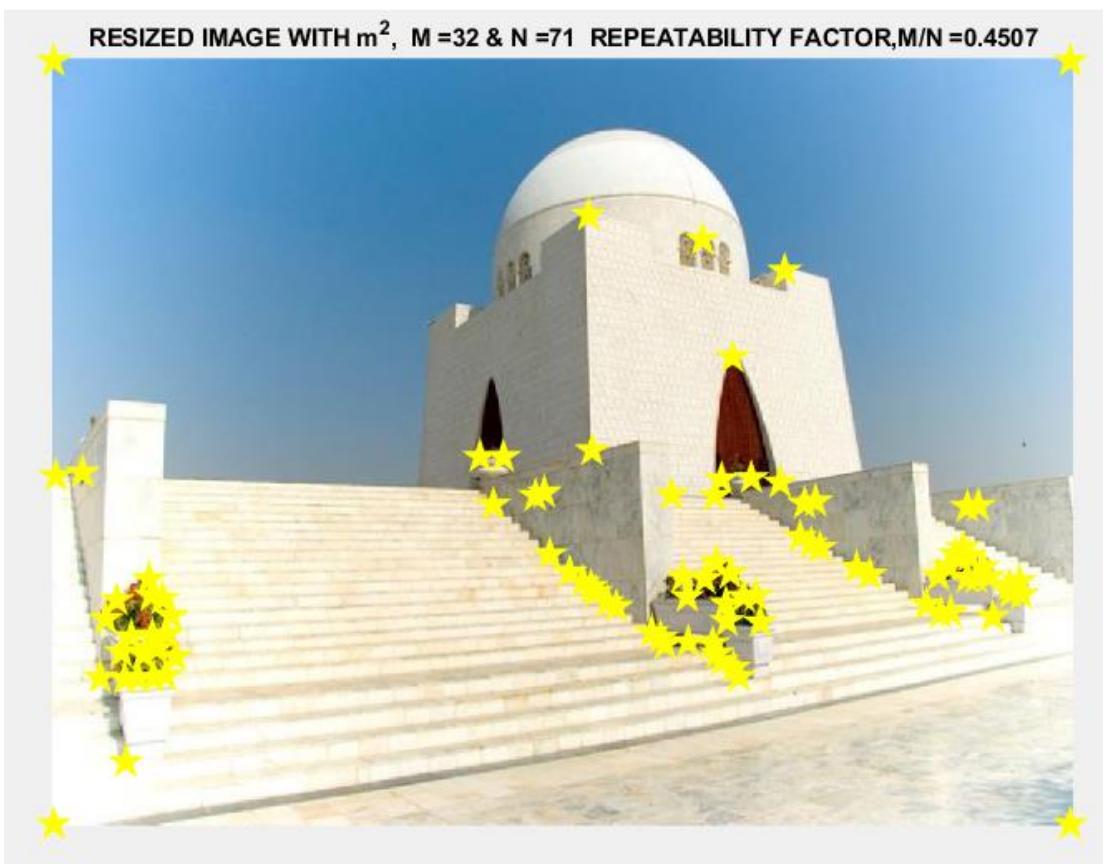
RESIZED IMAGE WITH m^0 , M =71 & N =71 REPEATABILITY FACTOR,M/N =1



RESIZED IMAGE WITH m^1 , M =42 & N =71 REPEATABILITY FACTOR,M/N =0.59155



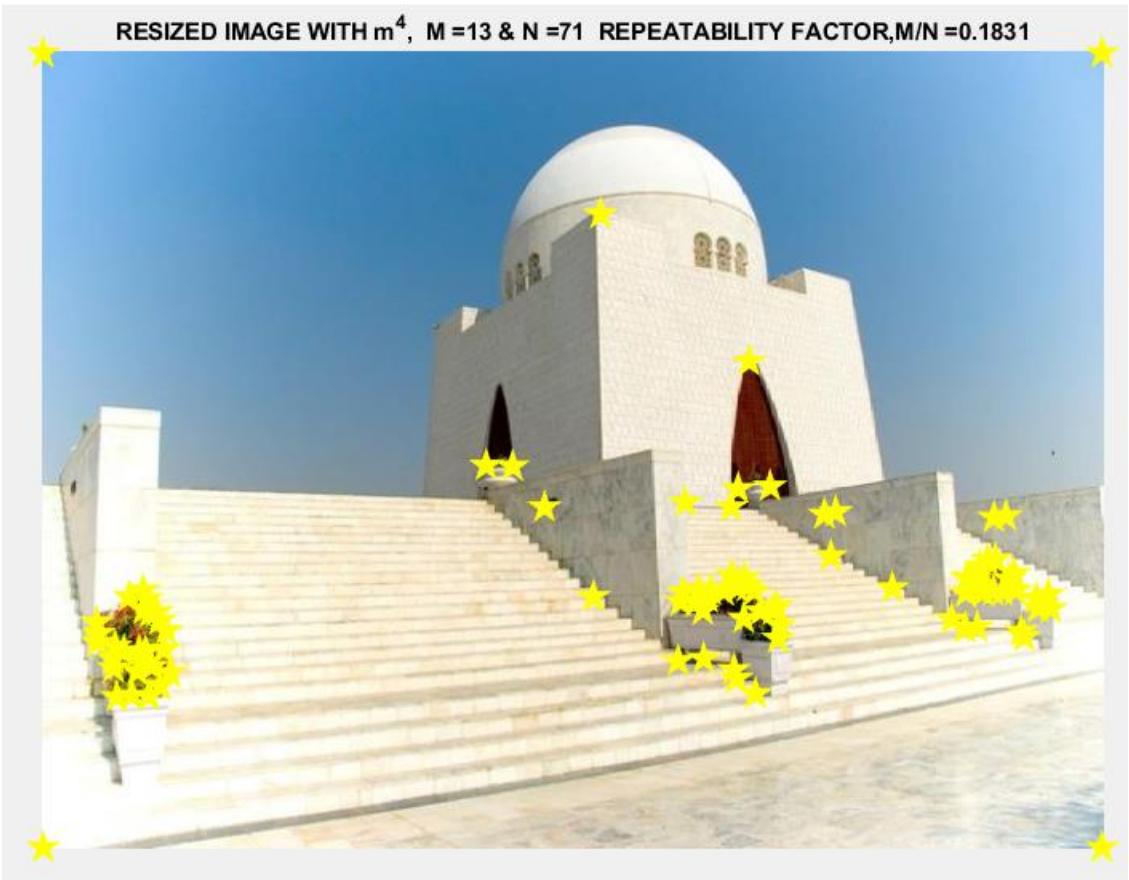
RESIZED IMAGE WITH m^2 , M =32 & N =71 REPEATABILITY FACTOR,M/N =0.4507



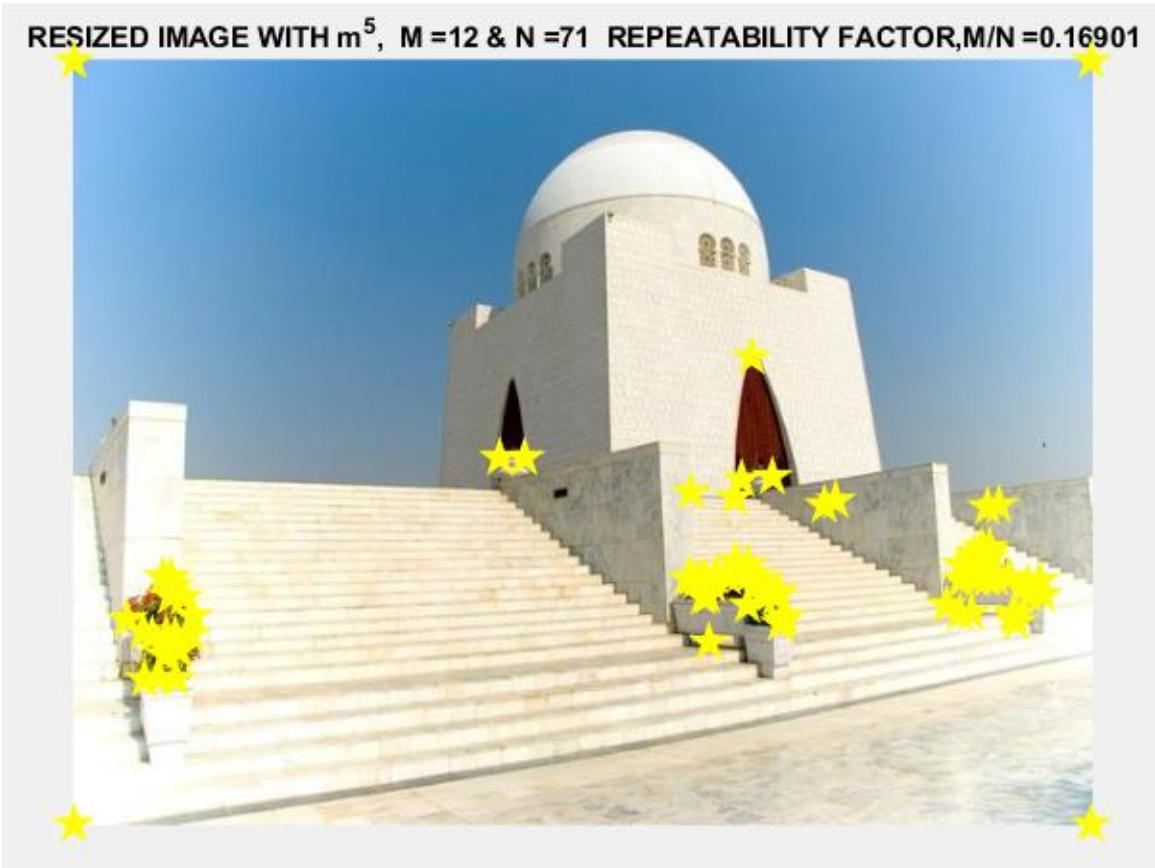
RESIZED IMAGE WITH m^3 , M =25 & N =71 REPEATABILITY FACTOR,M/N =0.35211



RESIZED IMAGE WITH m^4 , M =13 & N =71 REPEATABILITY FACTOR,M/N =0.1831



RESIZED IMAGE WITH m^5 , M =12 & N =71 REPEATABILITY FACTOR,M/N =0.16901



RESIZED IMAGE WITH m^6 , M = 0 & N = 71 REPEATABILITY FACTOR, M/N = 0

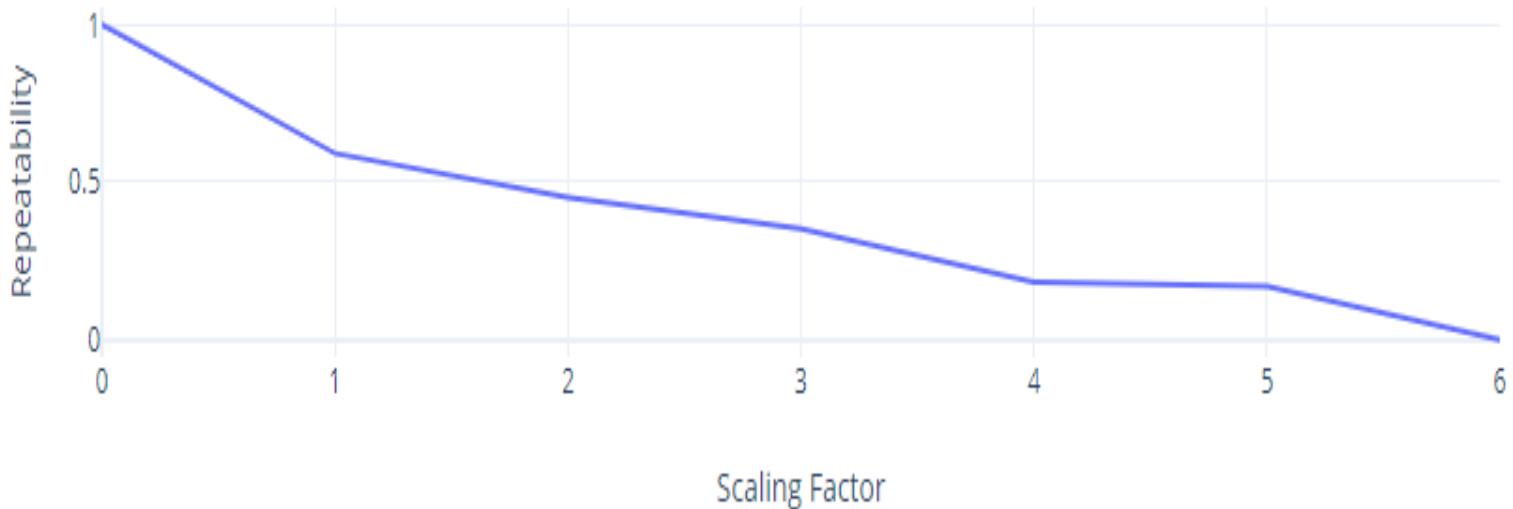


Note: As a result of facing some error while plotting repeatability vs scaling factor, an online tool has been used . The link for online tool is given below.

<https://chart-studio.plotly.com/create/>

Repeatability vs Scaling Factor Plot for the Image Mausoleum

Robustness of Harris Corner Detector to Scaling



Repeatability vs Scaling Factor Plot for the Image Famous Five

Robustness of Harris Corner Detector to Scaling



Conclusion

Repeatability plot for both images depicts that as we increase the scaling factor, the number of matched keypoints decreases which clearly shows that Harris corner detector is not robust to Scaling. In other words, Harris corner detector is not scale invariant.

Harris Corner Detector is invariant to Rotation but it is not invariant to Scaling. To some extent it has been shown in our results but not completely as it needs further work.

***** *The End* *****
