
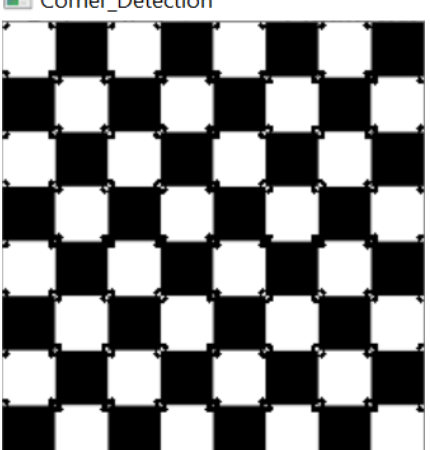

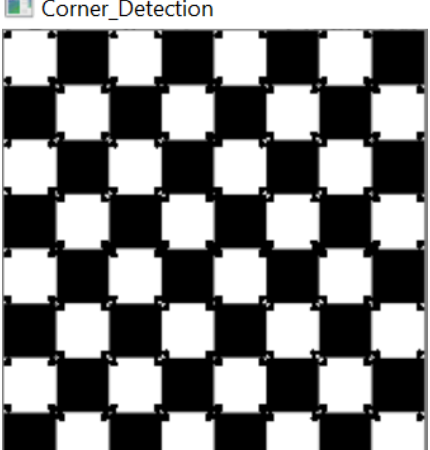

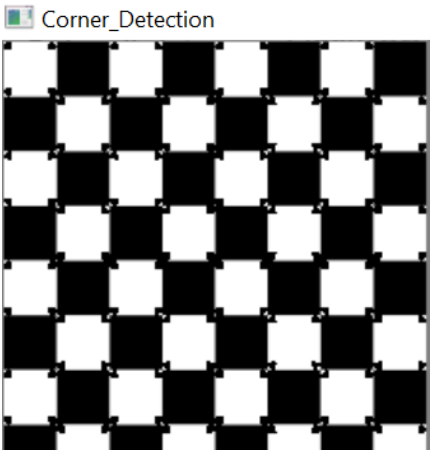





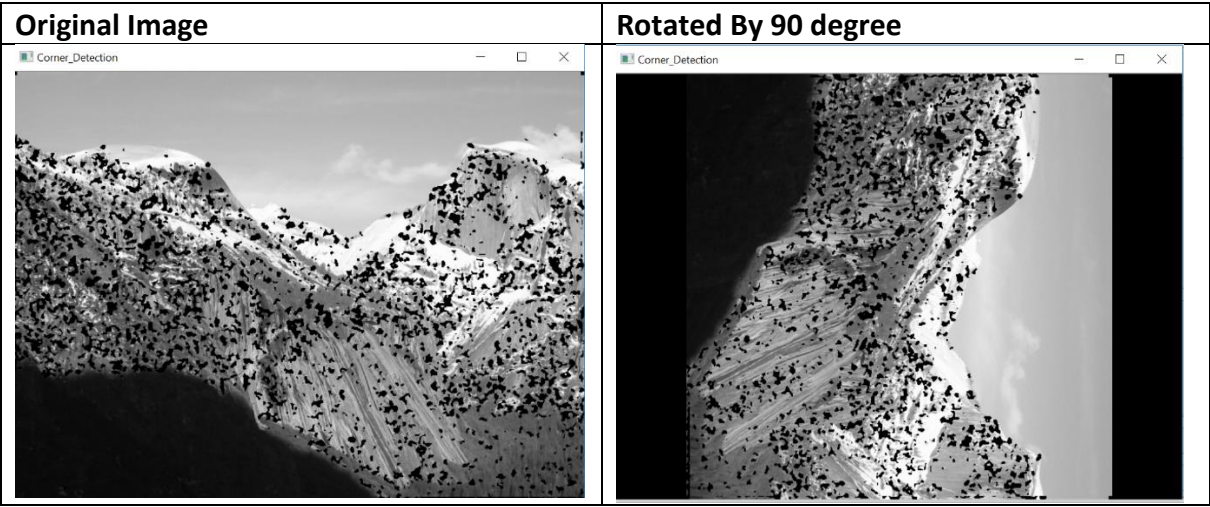
## Harris Corner Detection

Th=0.01		
Th=0.001		
Th=.0001		

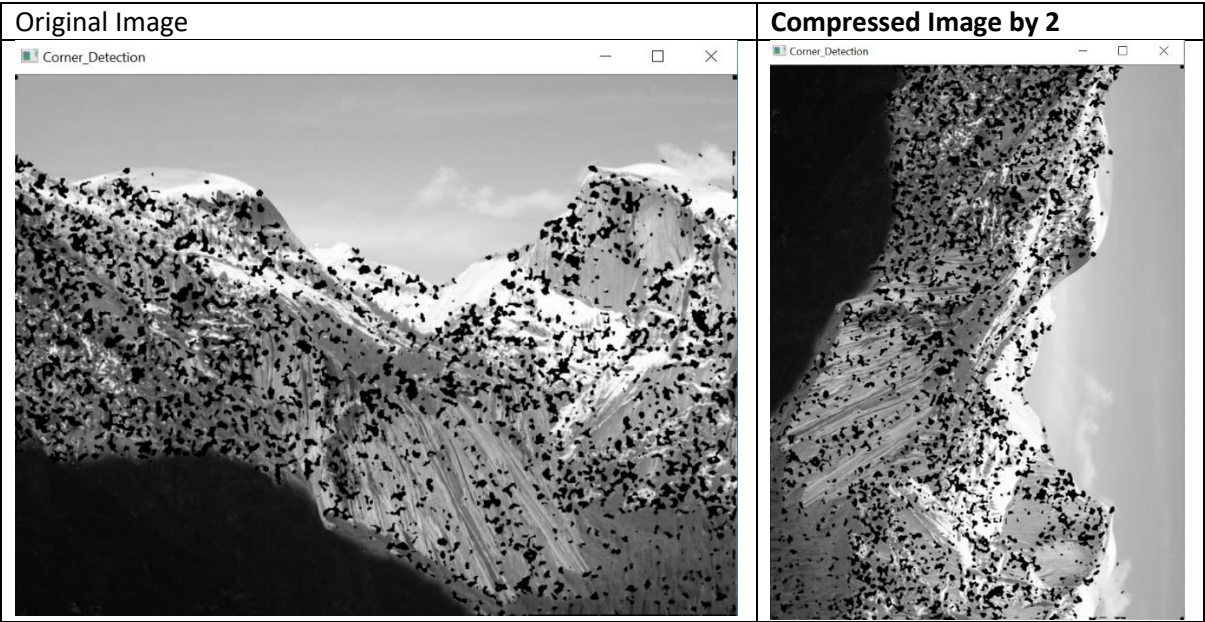
Th=0.01	 <p>A grayscale image of a flower with very few detected corner points, indicating a high threshold.</p>
Th=0.001	 <p>A grayscale image of a flower with a moderate number of detected corner points, indicating a medium threshold.</p>
Th=.0001	 <p>A grayscale image of a flower with many detected corner points, indicating a low threshold.</p>

a). Rotate the image clockwise 90 degrees.

For threshold = 0.0001(using cv2.getRotationMatrix2D())

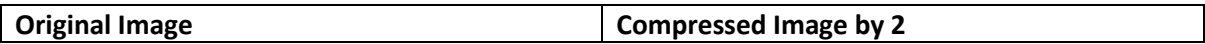


For threshold=0.0001(using np.rot90())



b). Compress the image by a factor of 2.

For threshold = 0.0001(using cv2.getRotationMatrix2D())







Observations : 1) As  $R > 0$  then corner is detected in given image where  $R$  is  $(\det(A) - k * (\text{trace}(A)^2))$  and  $A$  is determinant. If we increase the threshold then the number of corner decreases.

2). If we rotate the image by 90 degree then there will be no change in number of corners, it will be detected as same as in original image. Number of corners will not be affected if orientation of image is changed.

3). If we perform compression of image, the number of corners will be decreased because if we are resizing the image by performing down-sampling, the four pixel is converted to one pixel the corners will be accumulated. After down-sampling, since corner points will come more close to each other, and so x-gradient and y-gradient will now become more than in original image. Then, the neighbourhood pixels of corner points are more likely to be corner points in down-sampled image.