

Harris Corner detection Detection

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Abstract

Harris Corner Detector is the base of many other fast and efficient corner detection algorithms. It is used in a range of applications. Here in this report shows the implementation of Harris corner detection algorithm.

1 Introduction

Corners are unique and do not suffer from aperture problem, that may afflict edges. Corner is a position where a slight shift in position will lead to a large change in intensity in both horizontal(X) and vertical(Y) directions. Although corner are in very less amount with respect to image, but still used in applications like motion detecting and object recognition. It was first introduced by Chris Harris and Mike Stephens in 1988 upon the improvement of Moravec's corner detector. [1].

2 Methodology and Algorithms

First I calculated the gradient of image by applying (sobel or differentiation of gauss).

Filters used are:

for x-direction

$$\begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

for y-direction

$$\begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

We have to maximize $\sum_{x,y} [I(x+u, y+v) - I(x, y)]^2$ [2].

$$\text{As, } E(u, v) = \sum_{x,y} w(x, y) [I(x+u, y+v) - I(x, y)]^2$$

$E(u, v)$ can be written in matrix form as.

$$E(u, v) \approx \begin{bmatrix} u & v \end{bmatrix} M \begin{bmatrix} u \\ v \end{bmatrix}$$

$$\text{where } M = \sum_{x,y} w(x, y) \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix}$$

$$R = \det(M) - k(\text{trace}(M))^2$$

Here is its implementation . See Algorithm1 harris corner detection.

Algorithm 1 harris corner detection**Input:** *img_cpy*, *gray_img*, *Harris_const*, *window***Output:** *img_cpy*

```

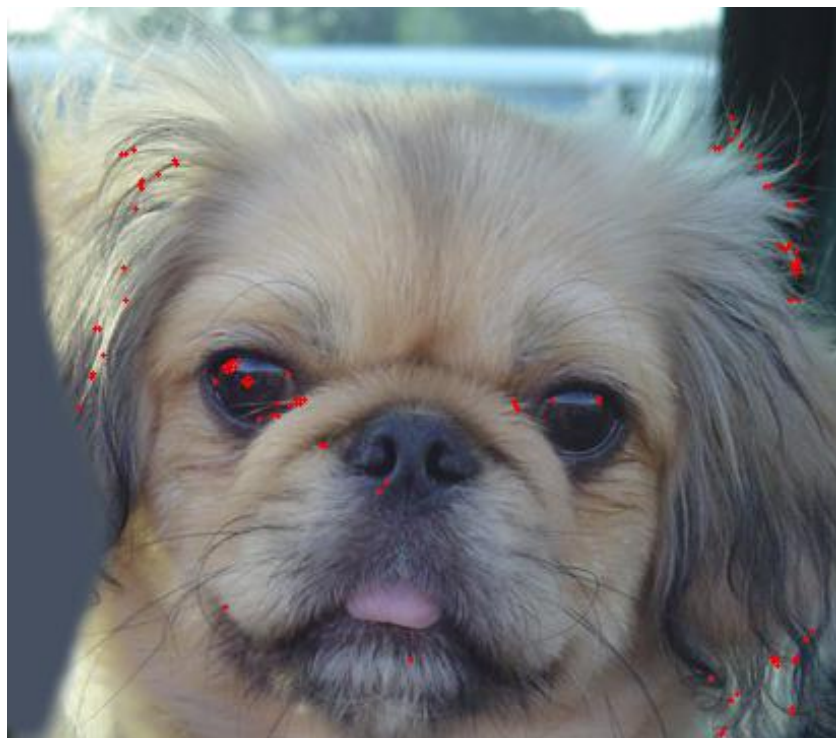
1:  $F_x, F_y \leftarrow \text{apply gradient on gray\_img}$ 
2:  $F_{xx}, F_{yy}, F_{xy} \leftarrow F_x^2, F_y^2, F_x * F_y$ 
3:  $height, width \leftarrow \text{img\_cpy.shape}$ 
4:  $offset \leftarrow \text{int}(window/2)$ 
5: for  $y \leftarrow offset$  to  $height - offset$  do
6:   for  $x \leftarrow offset$  to  $width - offset$  do
7:      $S_{xx} \leftarrow \sum(F_{xx}[y - offset : y + offset + 1, x - offset : x + offset + 1])$ 
8:      $S_{yy} \leftarrow \sum(F_{yy}[y - offset : y + offset + 1, x - offset : x + offset + 1])$ 
9:      $S_{xy} \leftarrow \sum(F_{xy}[y - offset : y + offset + 1, x - offset : x + offset + 1])$ 
10:     $det \leftarrow (S_{xx} * S_{yy}) - (S_{xy}^2)$ 
11:     $trace \leftarrow S_{xx} + S_{yy}$ 
12:     $r \leftarrow det - Harris\_const * (trace^2)$ 
13:    if  $r > Th$  then
14:       $cornerList.append([x, y, r])$ 
15:    end if
16:  end for
17: end for
18: sort cornerList with respect to r in decreasing order
19: suppress 8 neighbour pixels if thier r value is less value of r.
20: for top 100 corner pixels do
21:    $img\_cpy[Y1, X1] \leftarrow [255, 0, 0]$ 
22: end for

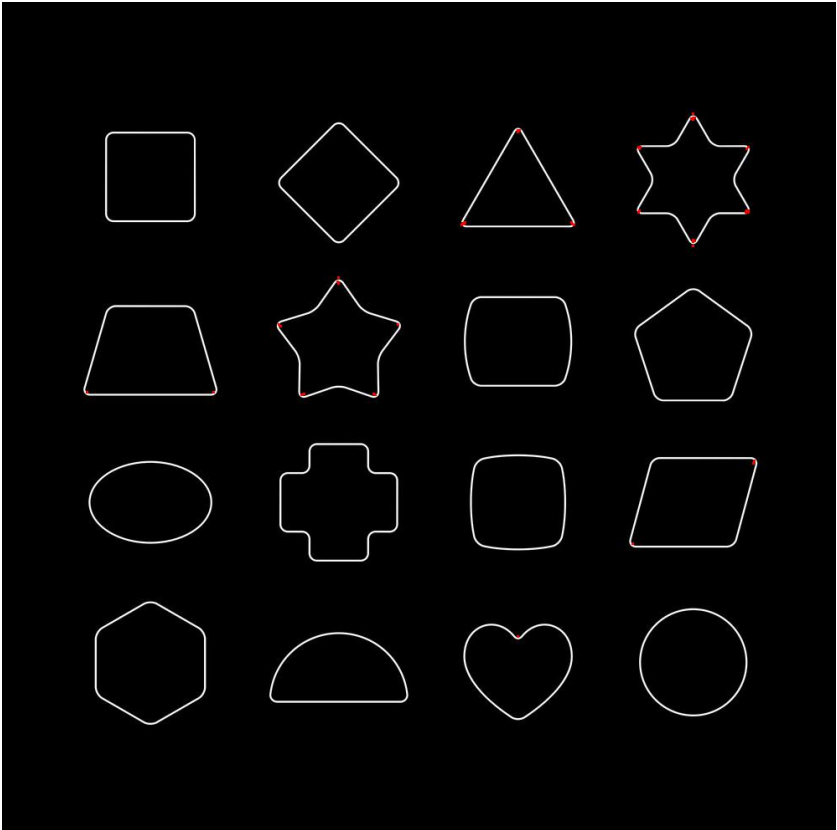
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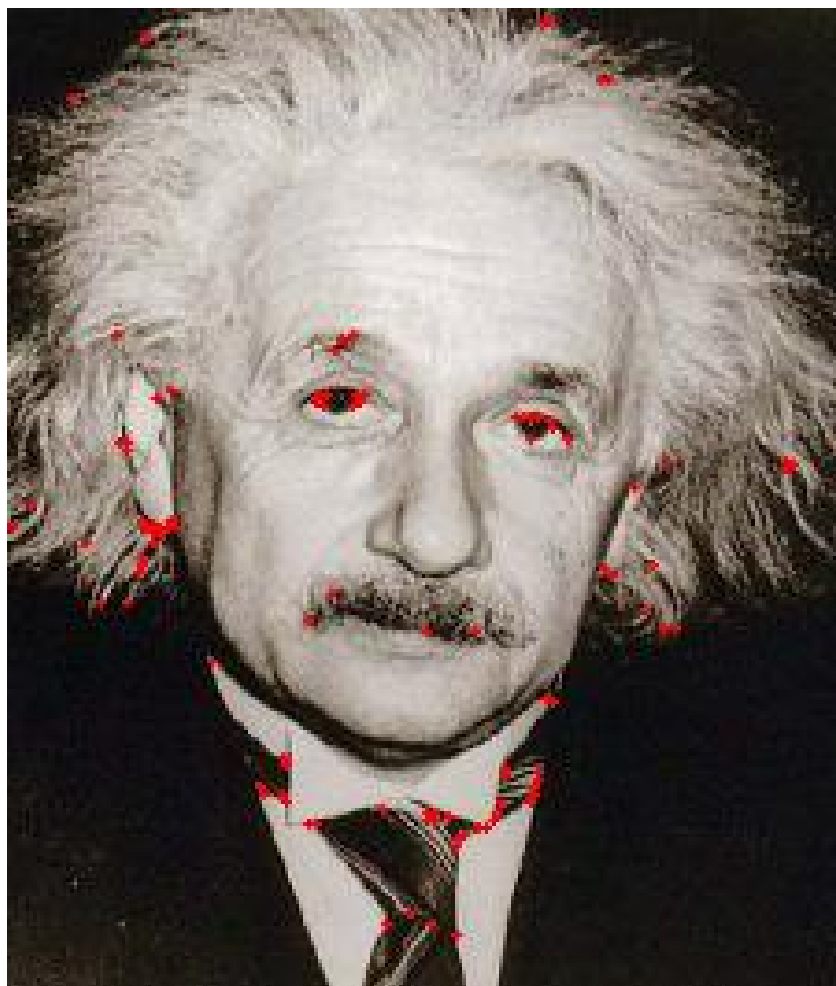
3 Results and Observations

- Generally, Harris constant belongs to [0.04,0.06] for all images. Lowering the threshold value increases the number of corners detected and also increases code's time complexity.
- In order to pick up the optimal values to indicate corners, we find the local maxima as corners within the window which is a 3 by 3 filter.
- Intersection of blur edges are not showing corners.
- A large value of R indicates a corner, a negative value indicates an edge.
- This works only for sharp corners, as in *toy_harris* image triangle and star image shows corners.









4 Conclusion and Takeaway

- Works best on binary image in comparison to coloured images.
- Lowering the threshold increases number of corners as well as time complexity of algorithm.

References

- [1] openCV. Harris corner detection. https://docs.opencv.org/3.4/d4/d7d/tutorial_harris_detector.html
- [2] Wikipedia. Harris corner detection. https://en.wikipedia.org/wiki/Harris_Corner_Detector.