

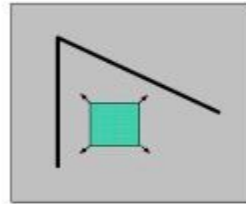


Harris Corner Detection

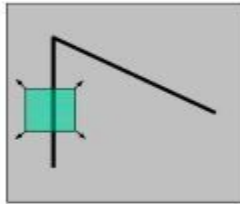
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Introduction:

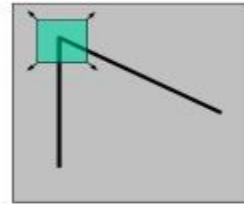
- A corner is a point whose local neighborhood stands in two dominant and different edge directions
- Corners are the important features in the image, and they are generally termed as interest points which are invariant to translation, rotation, and illumination.



"flat" region:
no change in all
directions



"edge":
no change along the
edge direction



"corner":
significant change in
all directions

Source:
<https://medium.com/data-breach/introduction-to-harris-corner-detector-32a88850b3f6>




Theory:

Algorithm basically finds the difference in intensity for a displacement of (u,v) in all directions. This is expressed as follows:

$$E(u, v) = \sum_{x,y} \underbrace{w(x, y)}_{\text{window function}} \underbrace{[I(x + u, y + v) - I(x, y)]}_{\text{shifted intensity} - \text{intensity}}^2$$

Window function is either a rectangular window or gaussian window which gives weights to pixels underneath.



We need to maximize function $E(u,v)$ for corner detection. We need to maximize second term in equation.


Applying Taylor Expansion to above equation and using some mathematical steps we get:

$$E(u, v) \approx [u \quad v] M \begin{bmatrix} u \\ v \end{bmatrix}$$

Where,

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

Here I_x and I_y are image derivatives in x and y directions respectively.



Below equation will determine if window can contain a corner or not.

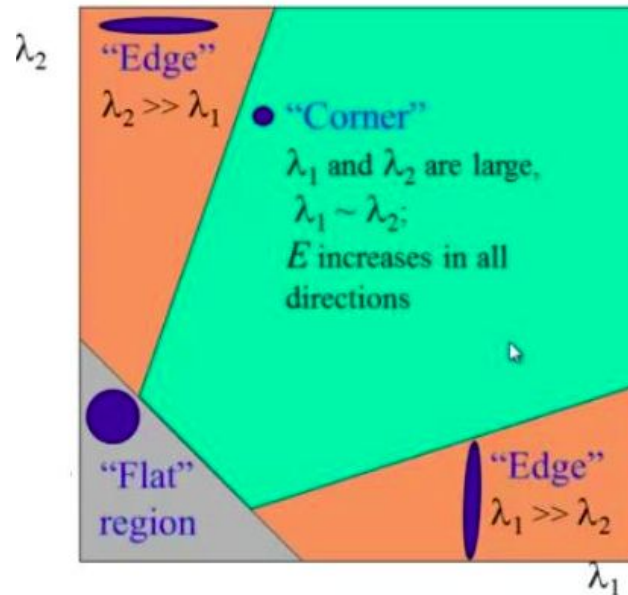
$$R = \det(M) - k(\text{trace}(M))^2$$

Where,

- $\det(M) = \lambda_1 \lambda_2$
- $\text{trace}(M) = \lambda_1 + \lambda_2$
- Eigenvalues of M are λ_1 and λ_2

So the values of these eigen values decide whether a region is corner, edge or flat.

- When $|R|$ is small, which happens when λ_1 and λ_2 are small, the region is flat
- When $R < 0$, which happens when $\lambda_1 \gg \lambda_2$ or vice versa, the region is edge
- When R is large, which happens when λ_1 and λ_2 are large and $\lambda_1 \sim \lambda_2$ the region is a corner.



Source:
https://opencv-python-tutroals.readthedocs.io/en/latest/_images/harris_region.jpg

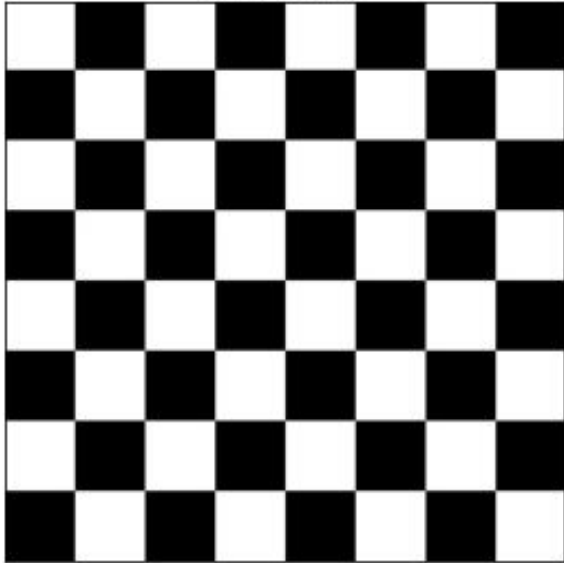


Pseudocode:

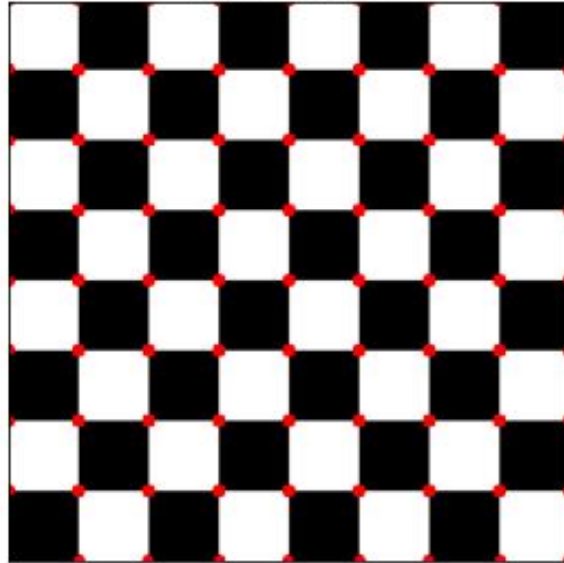
- Take the grayscale of the original image
- Apply a Gaussian filter to smooth out any noise
- Apply Sobel operator to find the x and y gradient values for every pixel in the grayscale image
- For each pixel p in the grayscale image, consider a 3×3 window around it and compute the corner strength function. Call this its Harris value.
- Find all pixels that exceed a certain threshold and are the local maxima within a certain window (to prevent redundant dupes of features)
- For each pixel that meets the criteria in 5, compute a feature descriptor.

Outputs:

Input Image

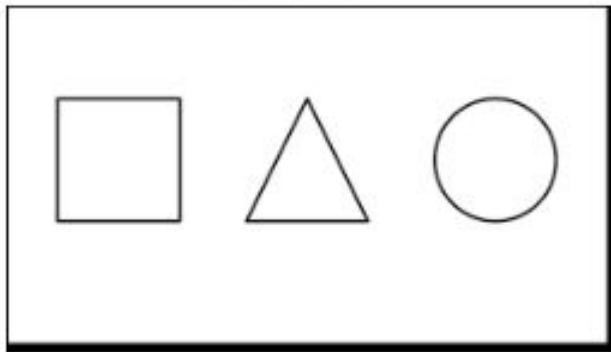


Harris Corners

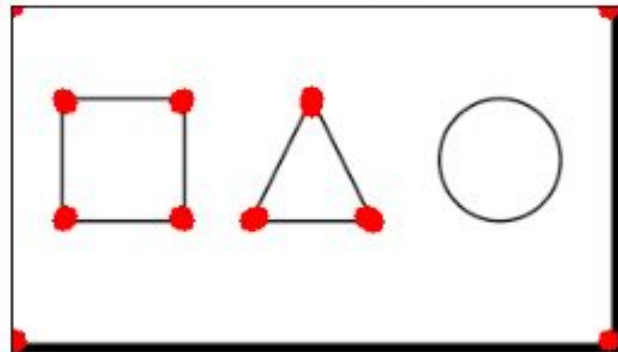




Input Image



Harris Corners





References:

- https://opencv-python-tutroals.readthedocs.io/en/latest/py_tutorials/py_feature2d/py_features_harris/py_features_harris.html
- <https://medium.com/data-breach/introduction-to-harris-corner-detector-32a88850b3f6>
- https://en.wikipedia.org/wiki/Harris_Corner_Detector