

Computer Vision

Interest Point detection
Corner Detection

Slides are created from the material By;
Oge Marques Practical Image and Video Processing Using Matlab
Mubarak Shah and Alper Yilmaz UCF, Fundamentals of Computer Vision
Richard Szeliski Computer Vision Algorithms and Applications

Interest Point Detection

Interest point

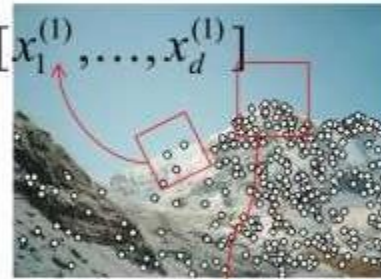
- Interest point or Feature Point is the point which is expressive in texture.
- Interest point is the point at which the direction of the boundary of the object changes abruptly or
- intersection point between two or more edge segments.

Local features: main components

- 1) Detection: Identify the interest points



- 2) Description :Extract feature vector descriptor surrounding each interest point.



- 3) Matching: Determine correspondence between descriptors in two views

$$\mathbf{x}_2 = [x_1^{(2)}, \dots, x_d^{(2)}]$$

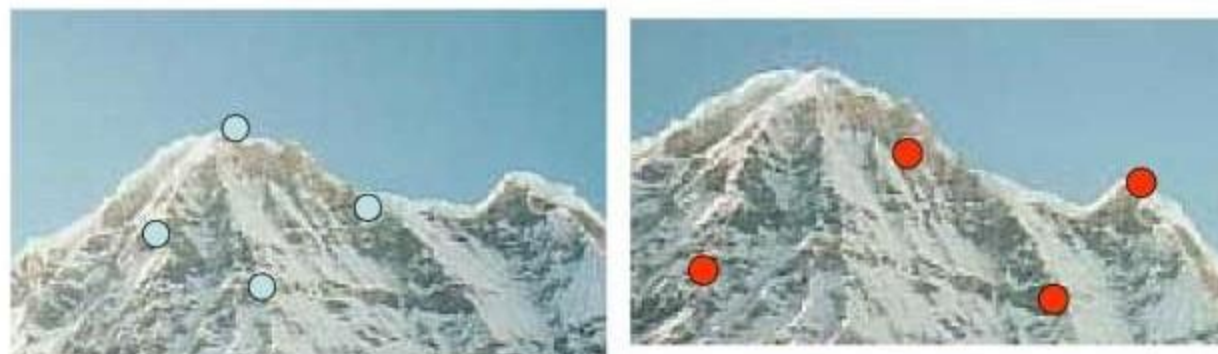


Where can we use it?

- Automate object tracking
- Point matching for computing disparity
- Stereo calibration
 - Estimation of fundamental matrix
- Motion based segmentation
- Recognition
- 3D object reconstruction
- Robot navigation
- Image retrieval and indexing

Goal: interest operator repeatability

- We want to detect (at least some of) the same points in both images.

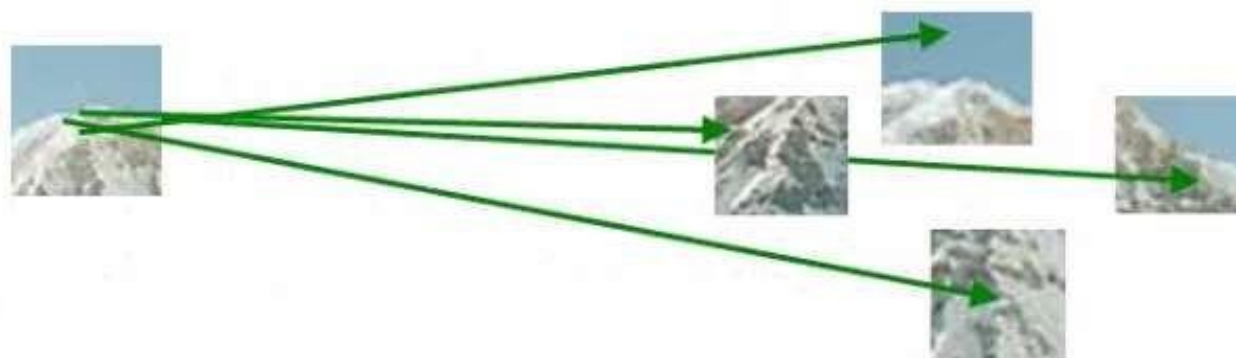


No chance to find true matches!

- Yet we have to be able to run the detection procedure *independently* per image.

Goal: descriptor distinctiveness

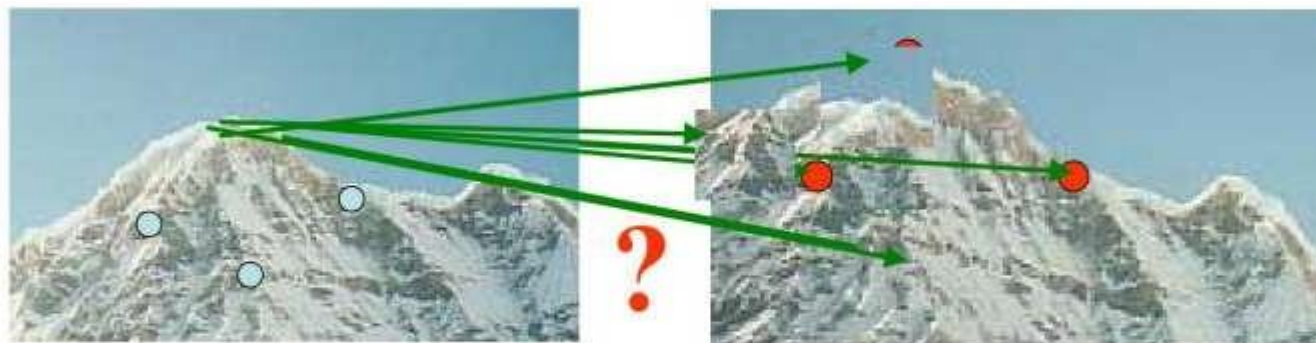
- We want to be able to reliably determine which point goes with which.



- Must provide some **invariance** to **geometric** and **photometric** differences between the two views.

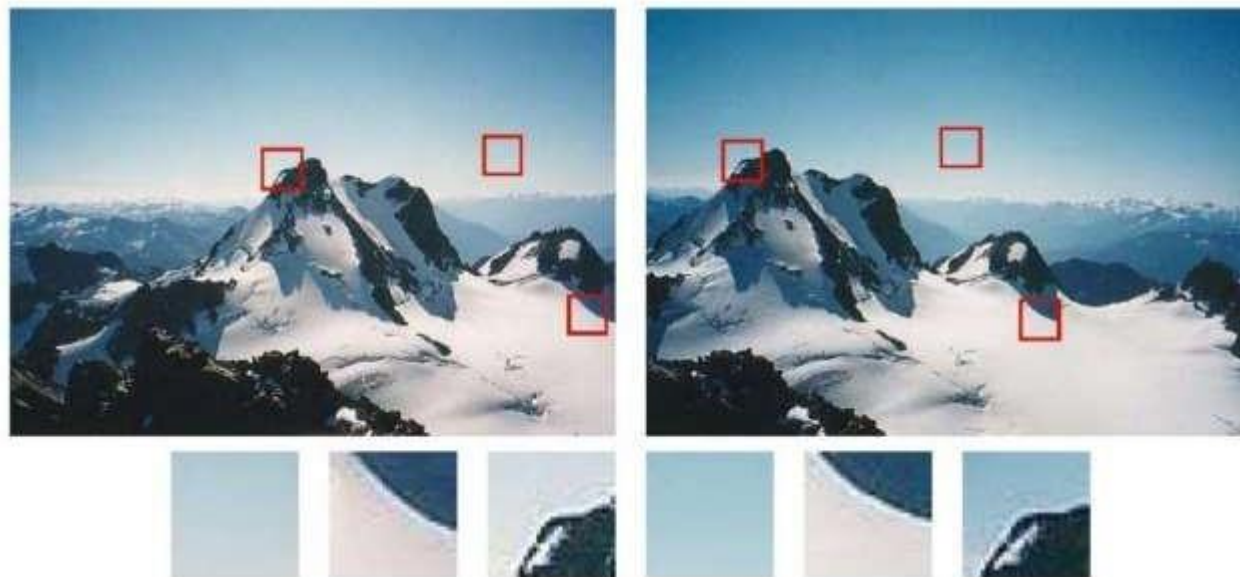
Goal: descriptor distinctiveness

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Some patches can be localized
or matched with higher accuracy than
others.



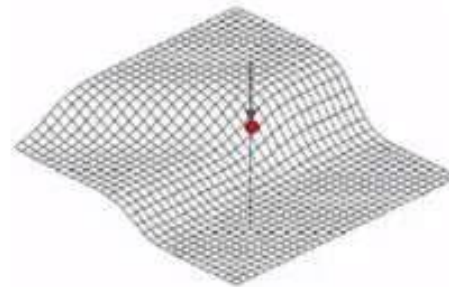
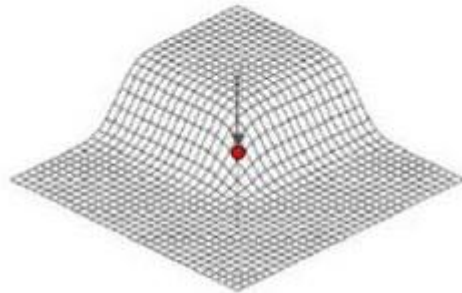
Local features: main components

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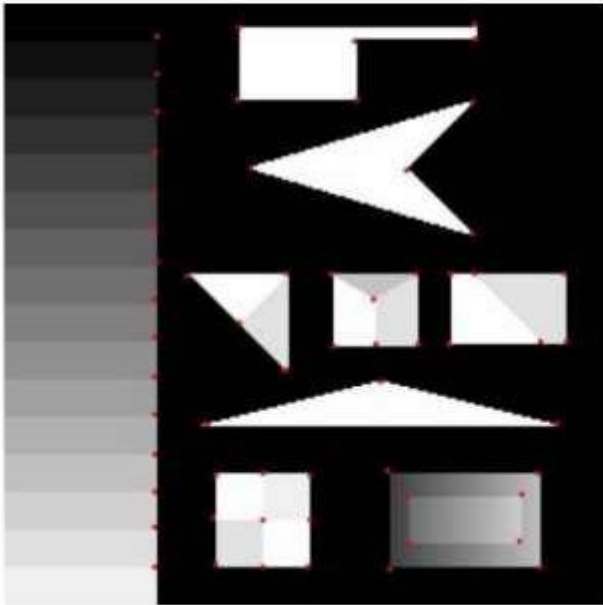


What is an interest point

- Expressive texture
 - The point at which the direction of the boundary of object changes abruptly
 - Intersection point between two or more edge segments



Synthetic & Real Interest Points



Corners are indicated in red



Properties of Interest Point Detectors

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- Detect all (or most) true interest points
- No false interest points
- Well localized.
- Robust with respect to noise.
- Efficient detection

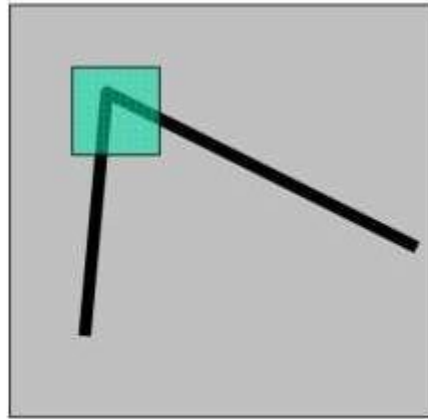
Possible Approaches to Corner Detection

- Based on brightness of images
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- Based on brightness of images
 - Usually image derivatives
- Based on boundary extraction
 - First step edge detection
 - Curvature analysis of edges

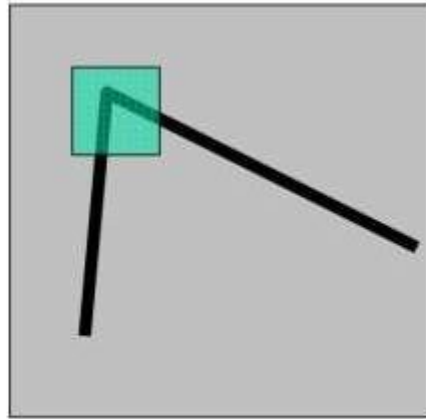
Harris Corner Detector



C.Harris, M.Stephens. "A Combined Corner and Edge Detector". 1988

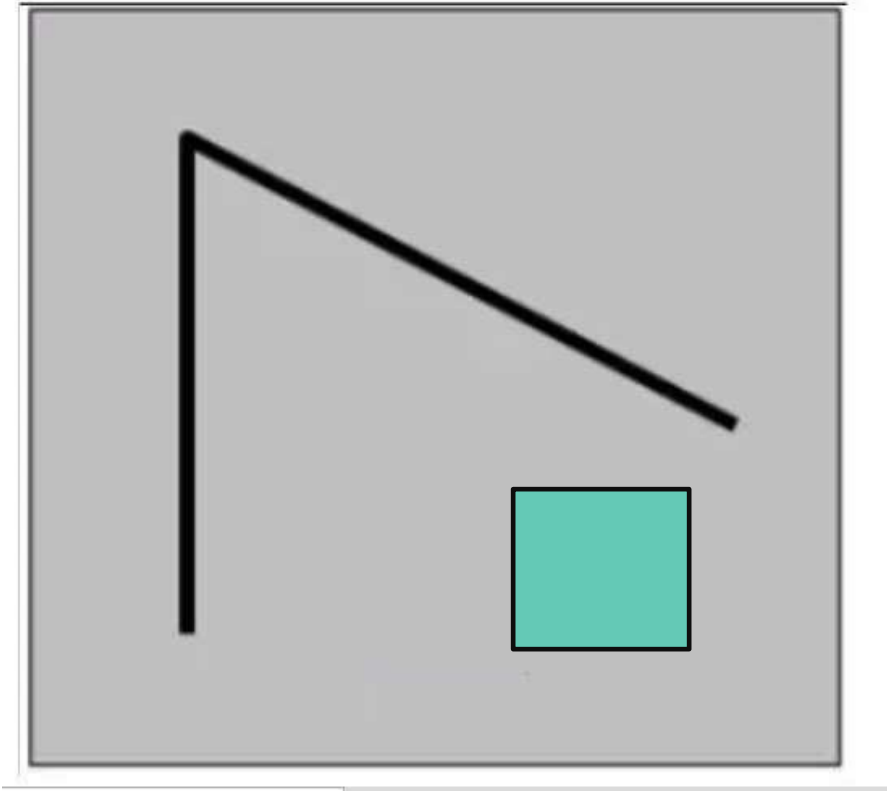
Harris Corner Detector

- Corner point can be recognized in a window

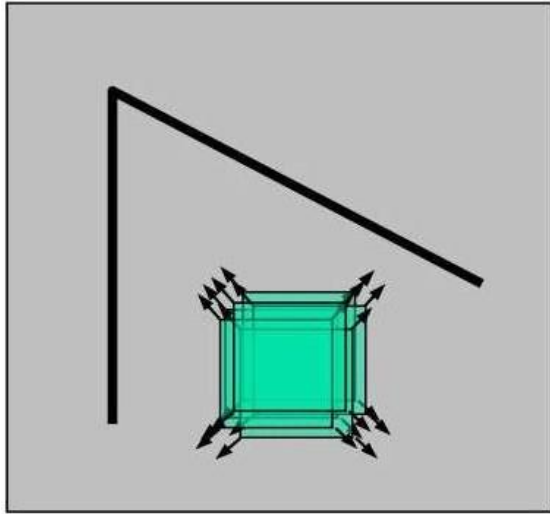


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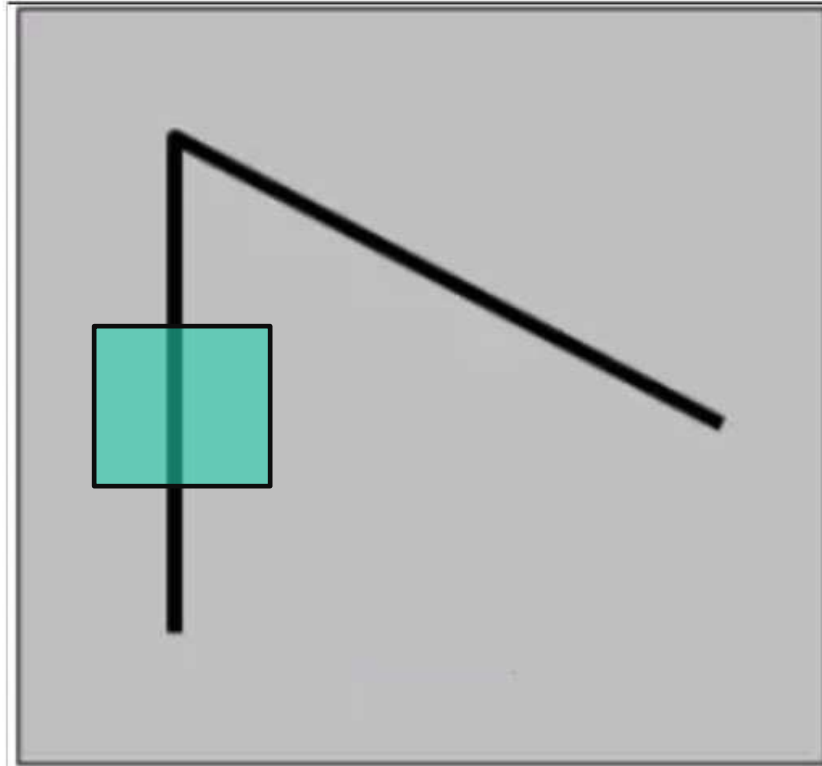
Basic Idea



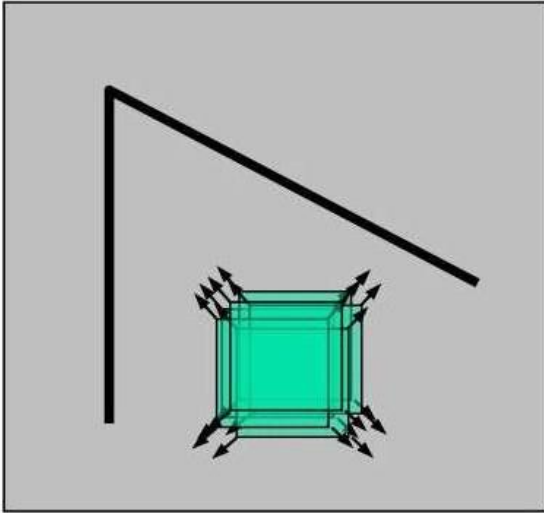
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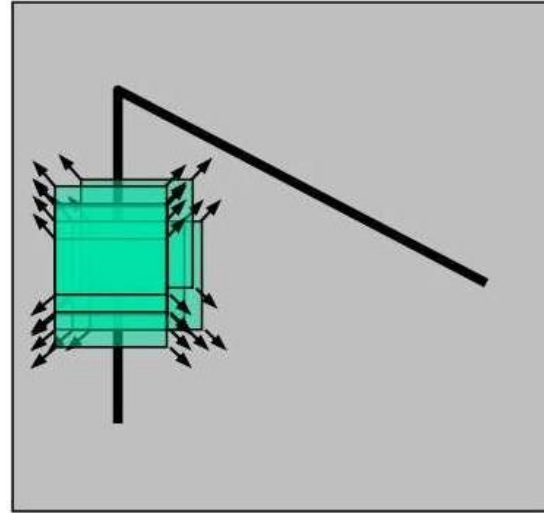
“flat” region:
no change in
all directions



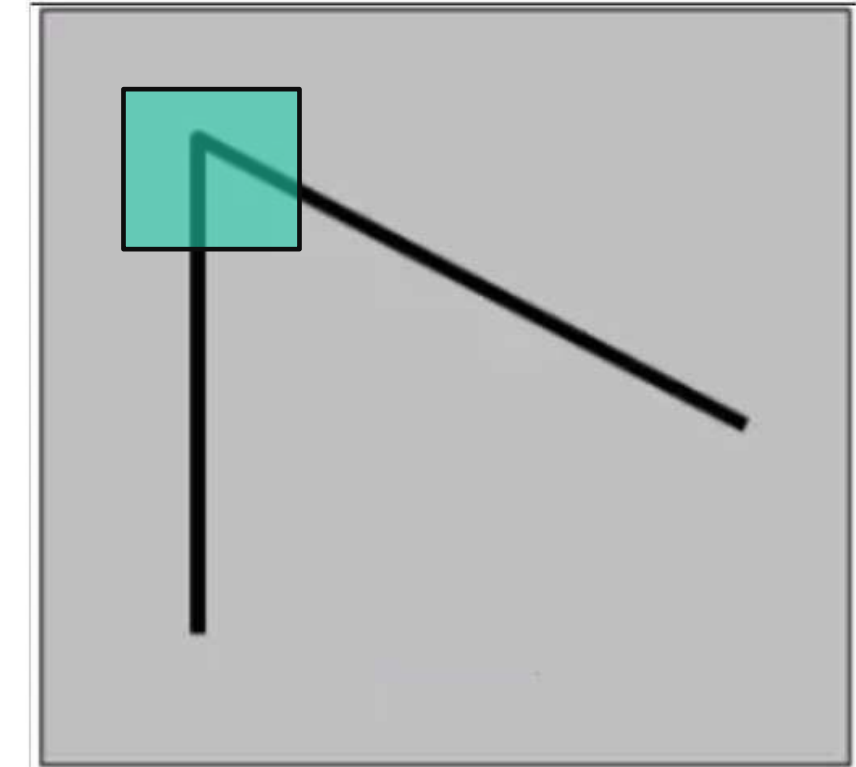
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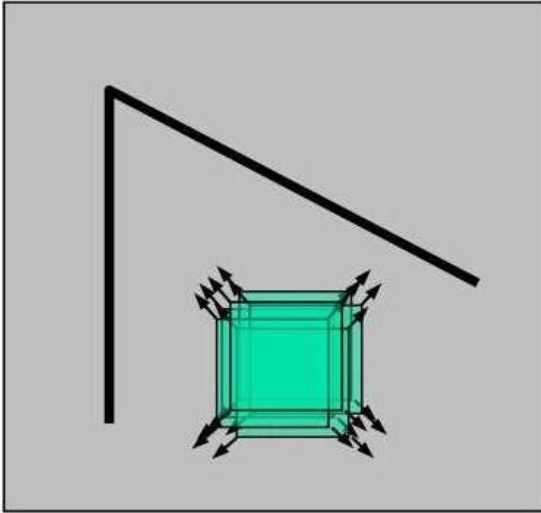
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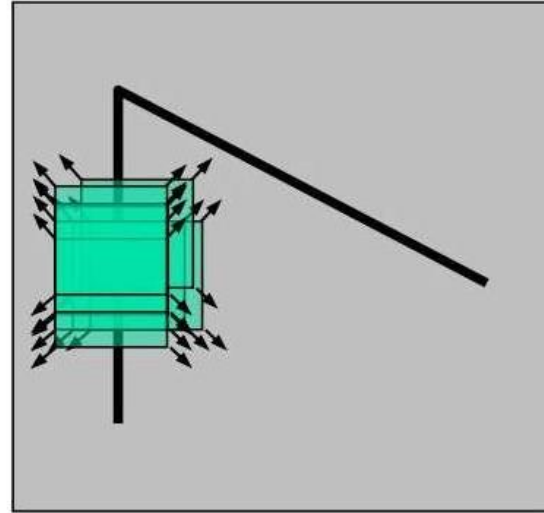
“edge”:
no change along
the edge direction



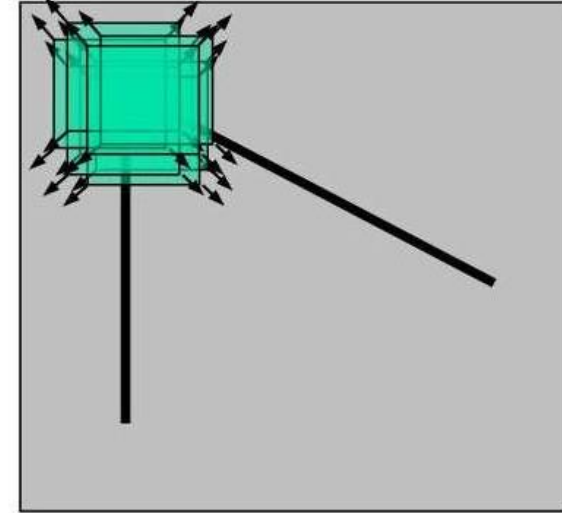
Basic Idea



“flat” region:
no change in
all directions



“edge”:
no change along
the edge direction



“corner”:
significant change
in all directions

Mathematics of Harris Detector

$$E(u,v) = \begin{pmatrix} u & v \end{pmatrix} M \begin{pmatrix} u \\ v \end{pmatrix}$$

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

Mathematics of Harris Detector

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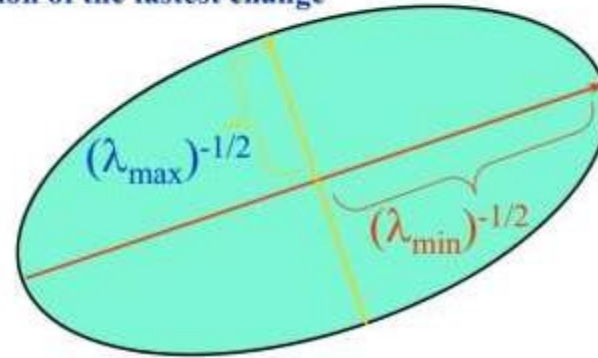
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direction of the fastest change



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Eigen Vectors and Eigen Values

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To find eigen values of a matrix A first find the roots of:

$$\det(A - \lambda I) = 0$$

Then solve the following linear system for each eigen value to find corresponding eigen vector

$$(A - \lambda I)x = 0$$

Example

$$A = \begin{bmatrix} -1 & 2 & 0 \\ 0 & 3 & 4 \\ 0 & 0 & 7 \end{bmatrix}$$

Eigen Values

$$\det(A - \lambda I) = 0$$

$$\lambda_1 = 7, \lambda_2 = 3, \lambda_3 = -1$$

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$$\mathbf{x}_1 = \begin{bmatrix} 1 \\ 4 \\ 4 \end{bmatrix}, \mathbf{x}_2 = \quad \mathbf{x}_3 =$$

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$$\det\left(\begin{bmatrix} -1-\lambda & 2 & 0 \\ 0 & 3-\lambda & 4 \\ 0 & 0 & 7-\lambda \end{bmatrix}\right) = 0$$

$$(-1-\lambda)((3-\lambda)(7-\lambda)-0) = 0$$

$$(-1-\lambda)(3-\lambda)(7-\lambda) = 0$$

$$\lambda = -1, \quad \lambda = 3, \quad \lambda = 7$$

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$$\left(\begin{bmatrix} -1 & 2 & 0 \\ 0 & 3 & 4 \\ 0 & 0 & 7 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right) \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

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$$0 + 2x_2 + 0 = 0$$

$$0 + 4x_2 + 4x_3 = 0$$

$$0 + 0 + 8x_3 = 0$$

$$x_1 = 1, \quad x_2 = 0, \quad x_3 = 0$$

$$\mathbf{x}_1 = \begin{bmatrix} 1 \\ 4 \\ 4 \end{bmatrix}, \quad \mathbf{x}_2 =$$

$$\mathbf{x}_3 =$$

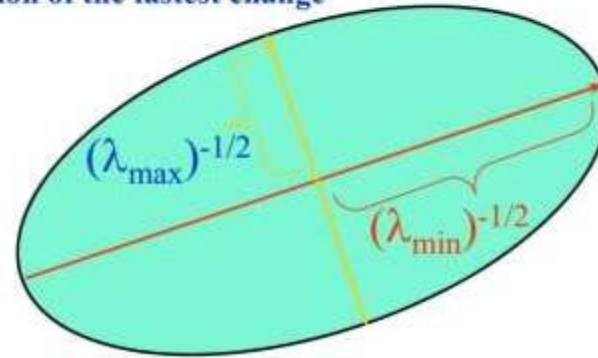
Eigen Vectors

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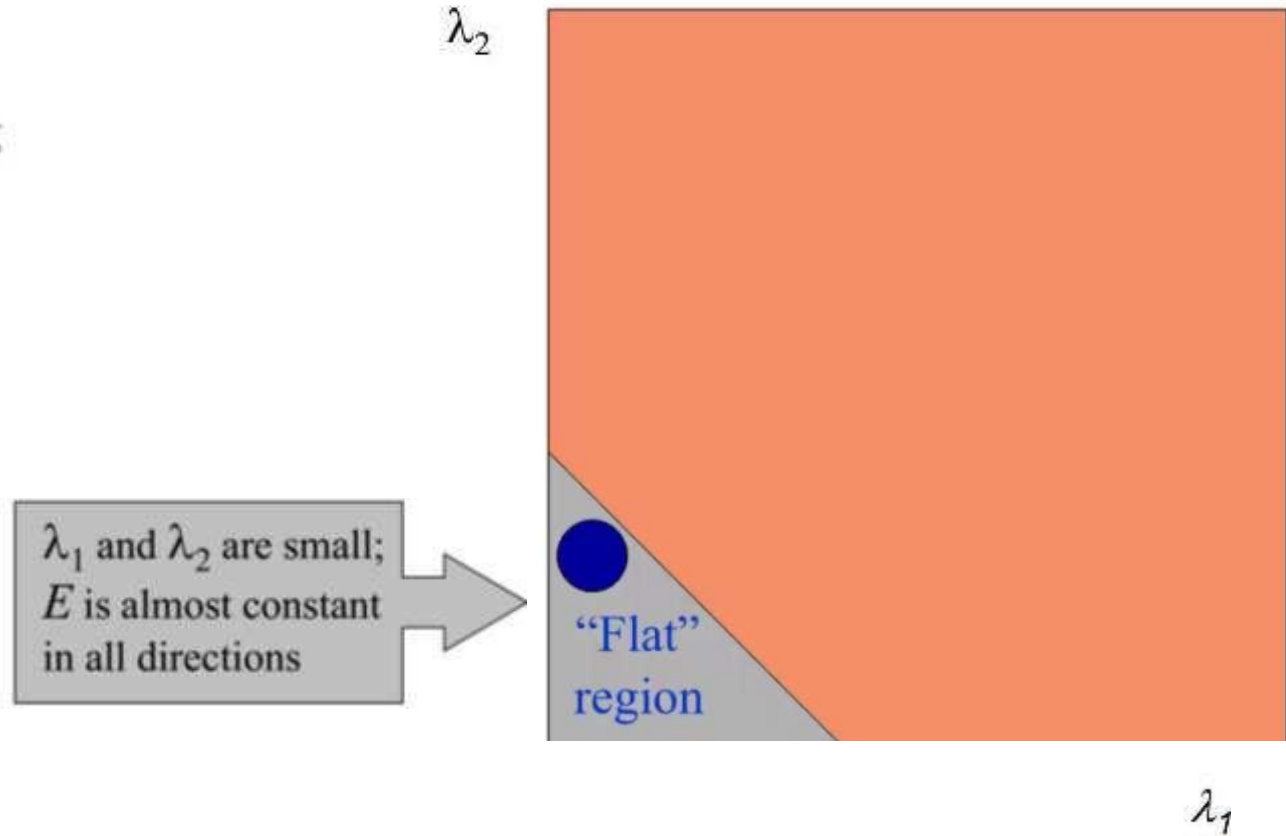
Mathematics of Harris Detector

Classification of
image points using
eigenvalues of M :

Now we will look into these eigen values and decide whether that point is interest point/corner or not which are actually the Derivative of I_x and I_y and their squares....

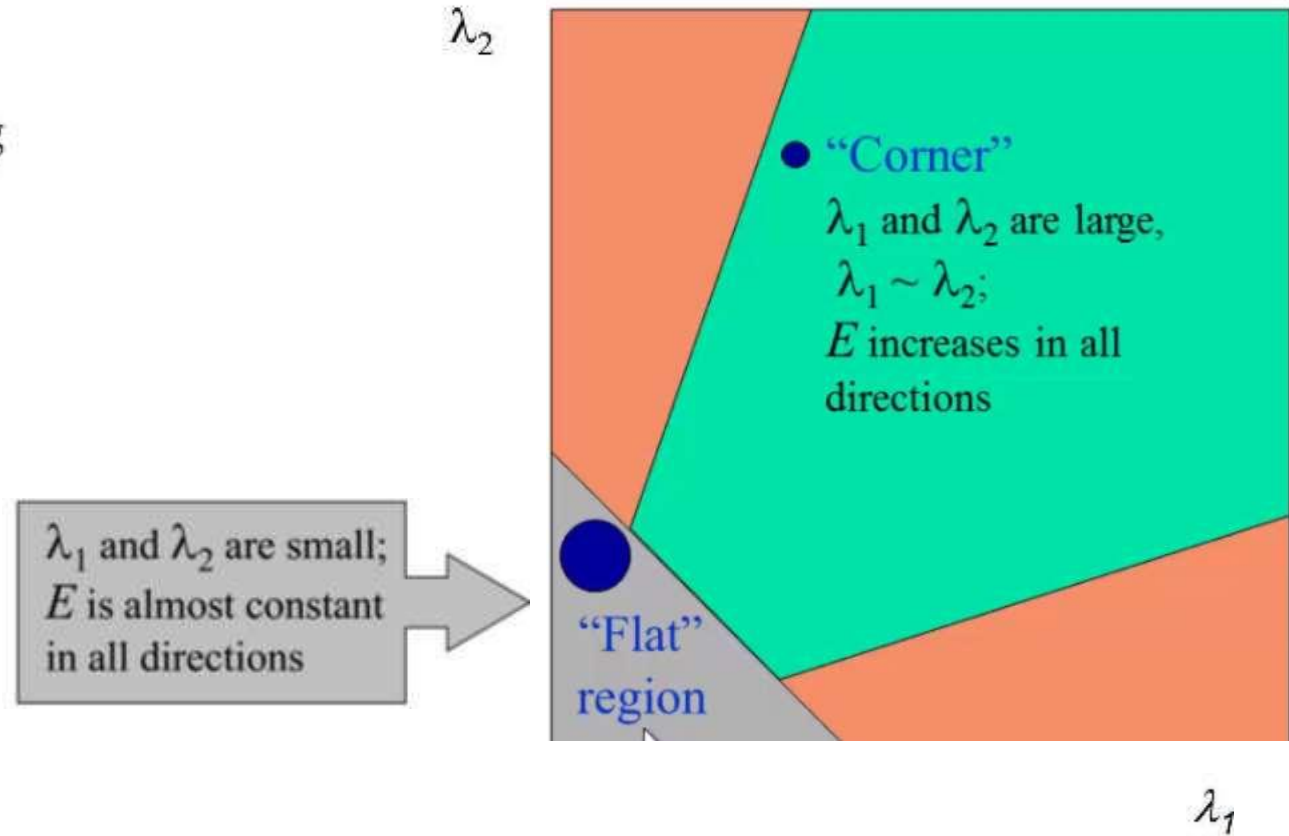
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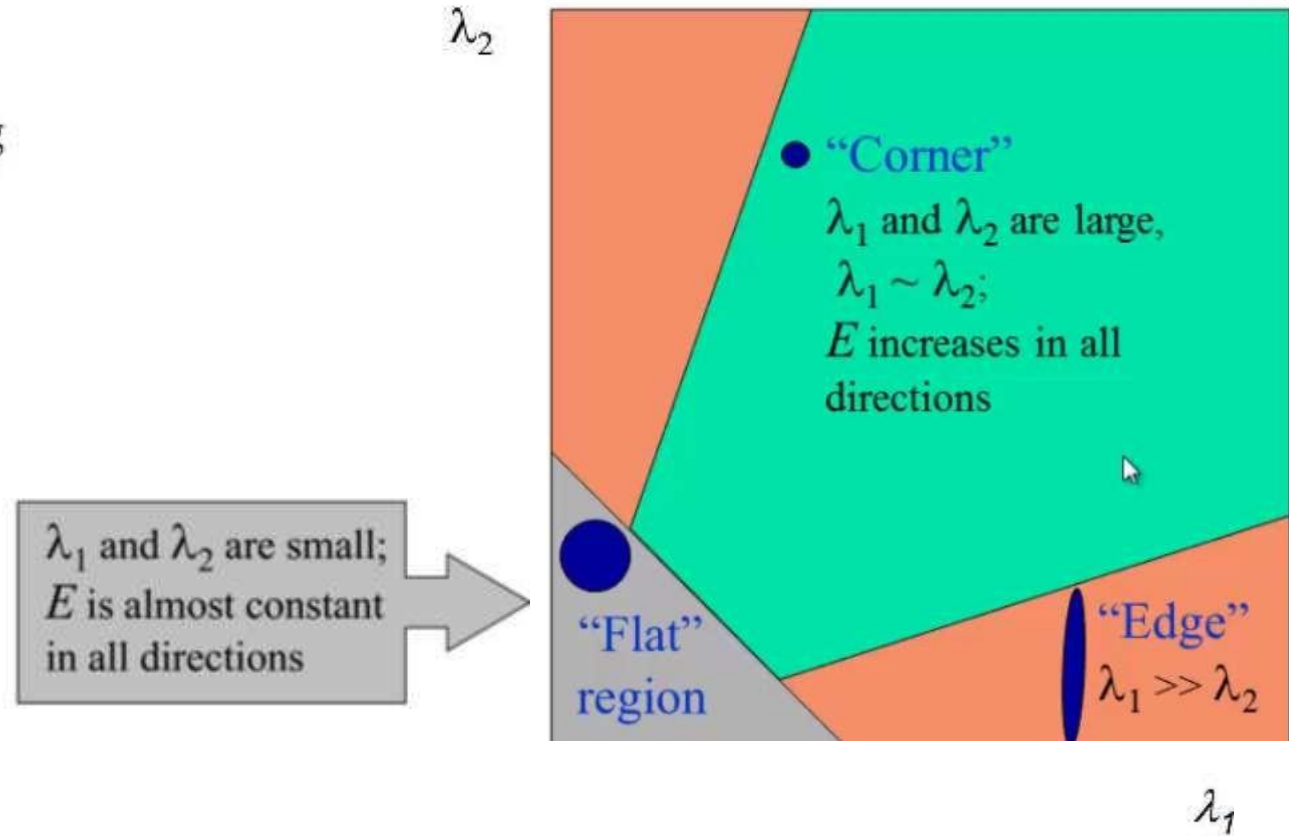
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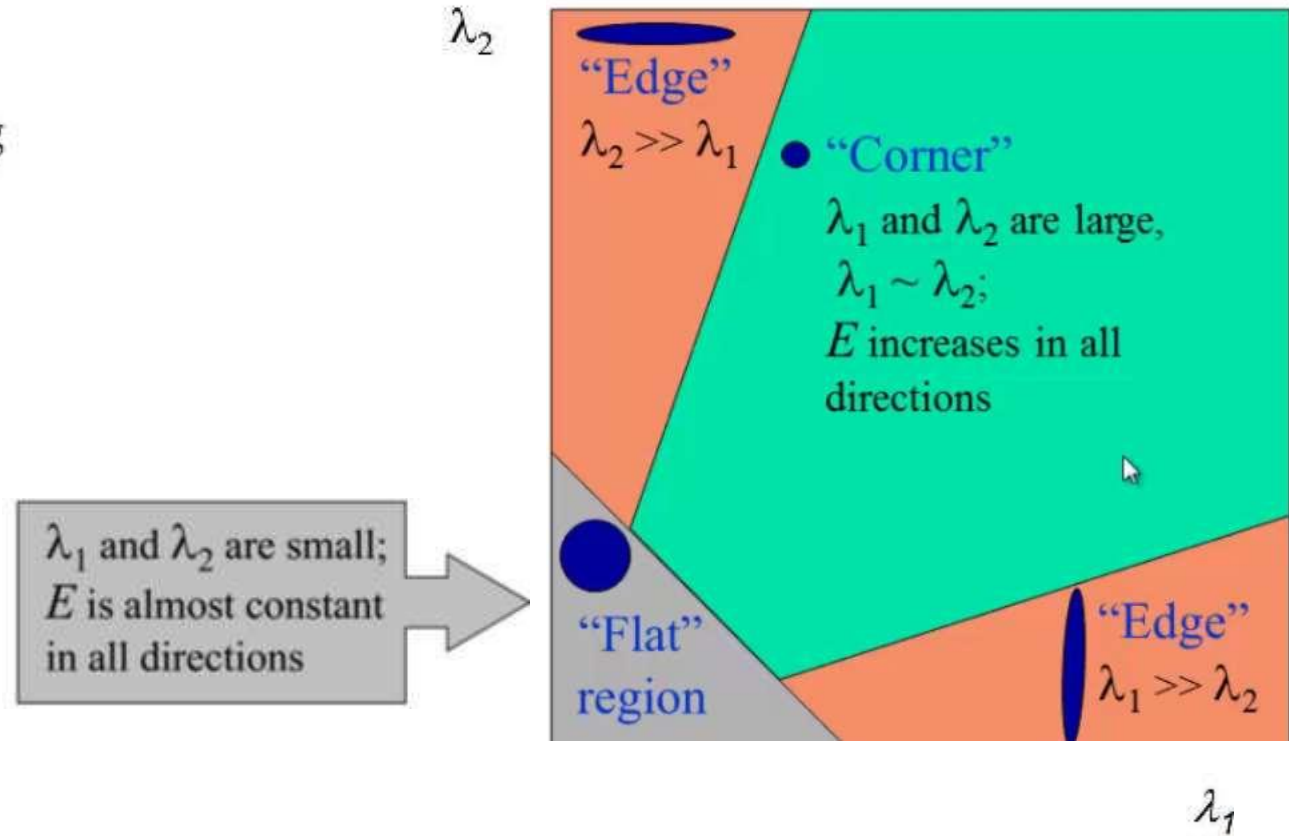
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Mathematics of Harris Detector

Classification of image points using eigenvalues of M :



Mathematics of Harris Detector

- Measure of cornerness in terms of λ_1, λ_2

$$M = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

$$R = \det M - k(\text{trace} M)^2 \qquad R = \lambda_1 \lambda_2 - k(\lambda_1 + \lambda_2)^2$$

Mathematics of Harris Detector

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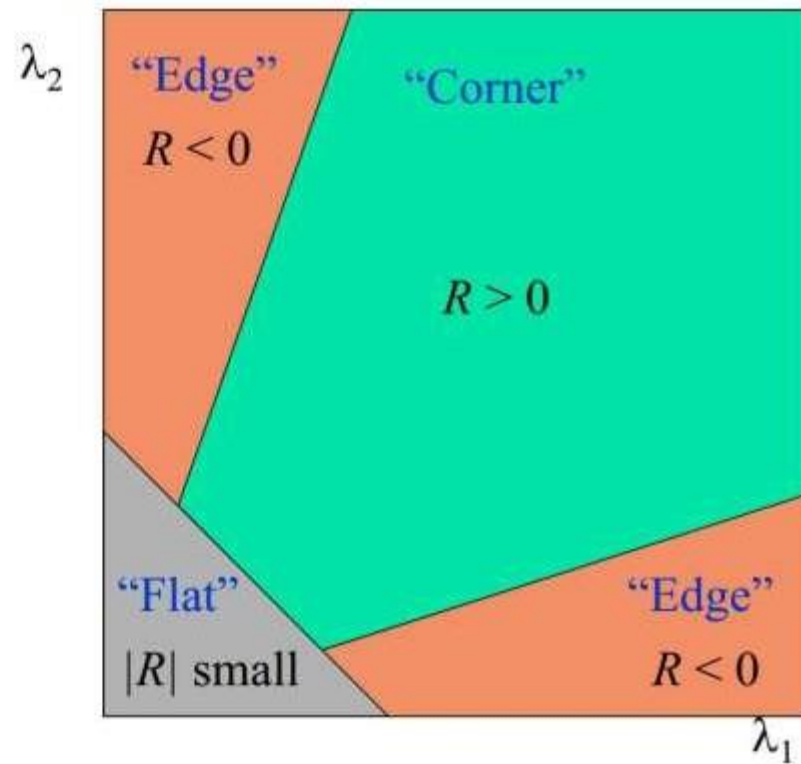
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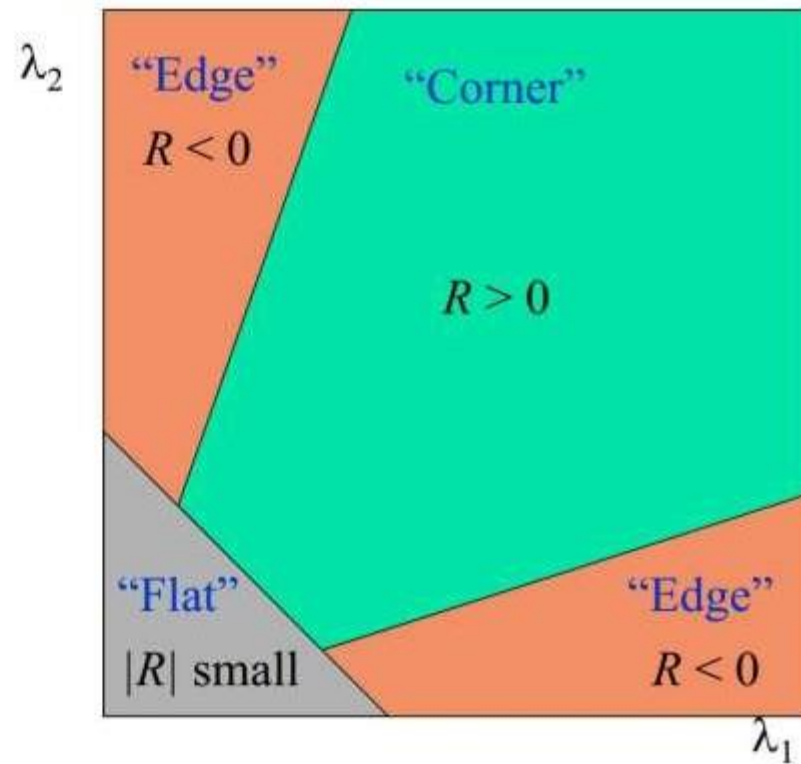
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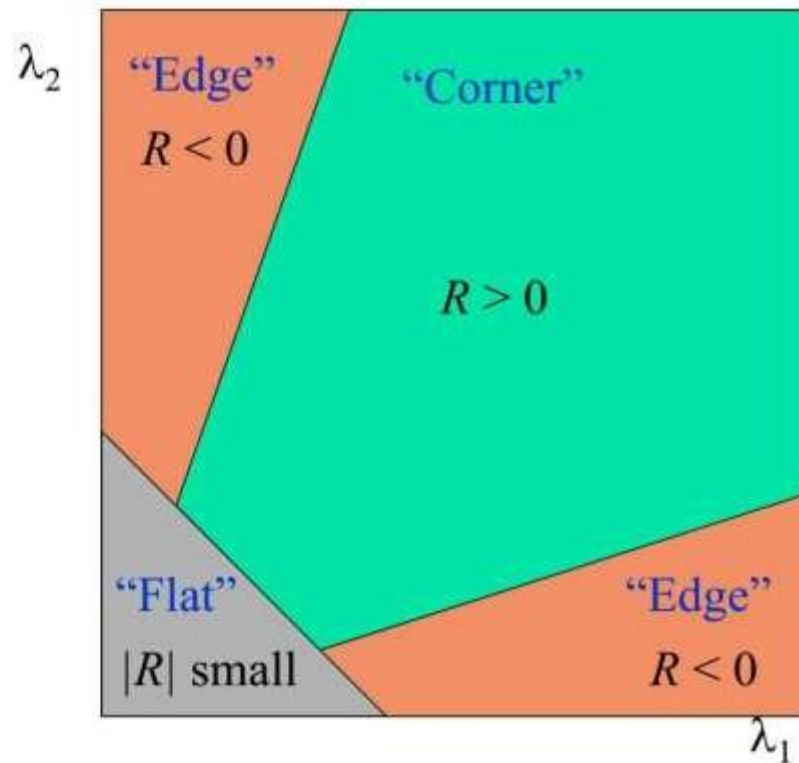
Mathematics of Harris Detector

- R depends only on eigenvalues of M



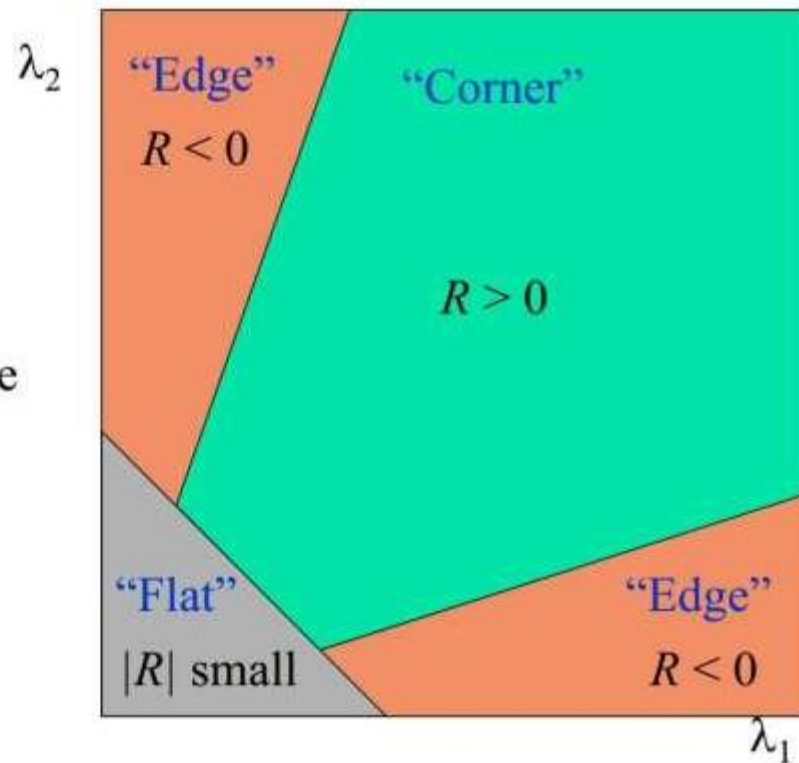
Mathematics of Harris Detector

- R depends only on eigenvalues of M
- R is large for a corner



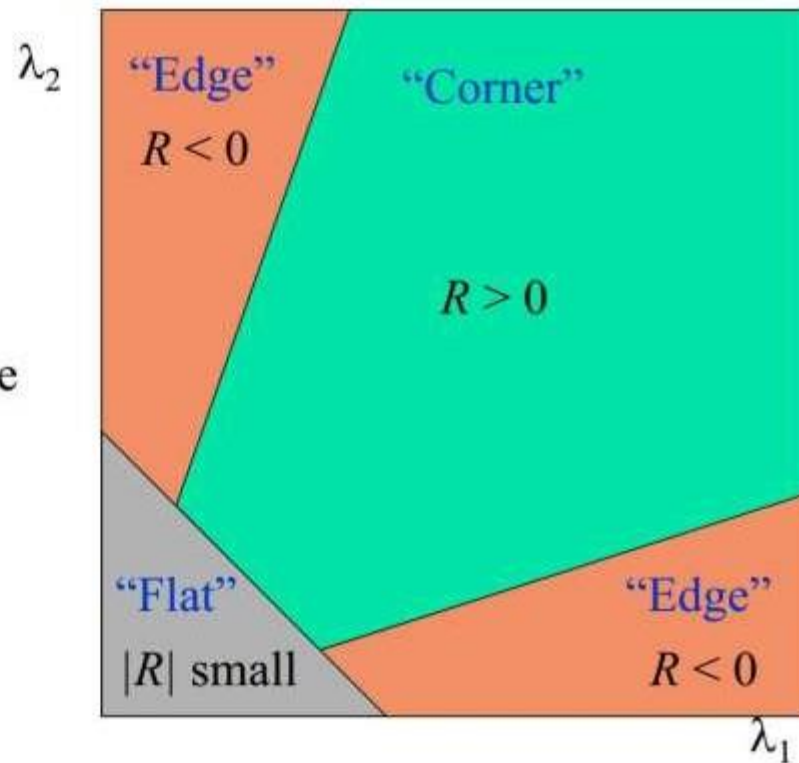
Mathematics of Harris Detector

- R depends only on eigenvalues of M
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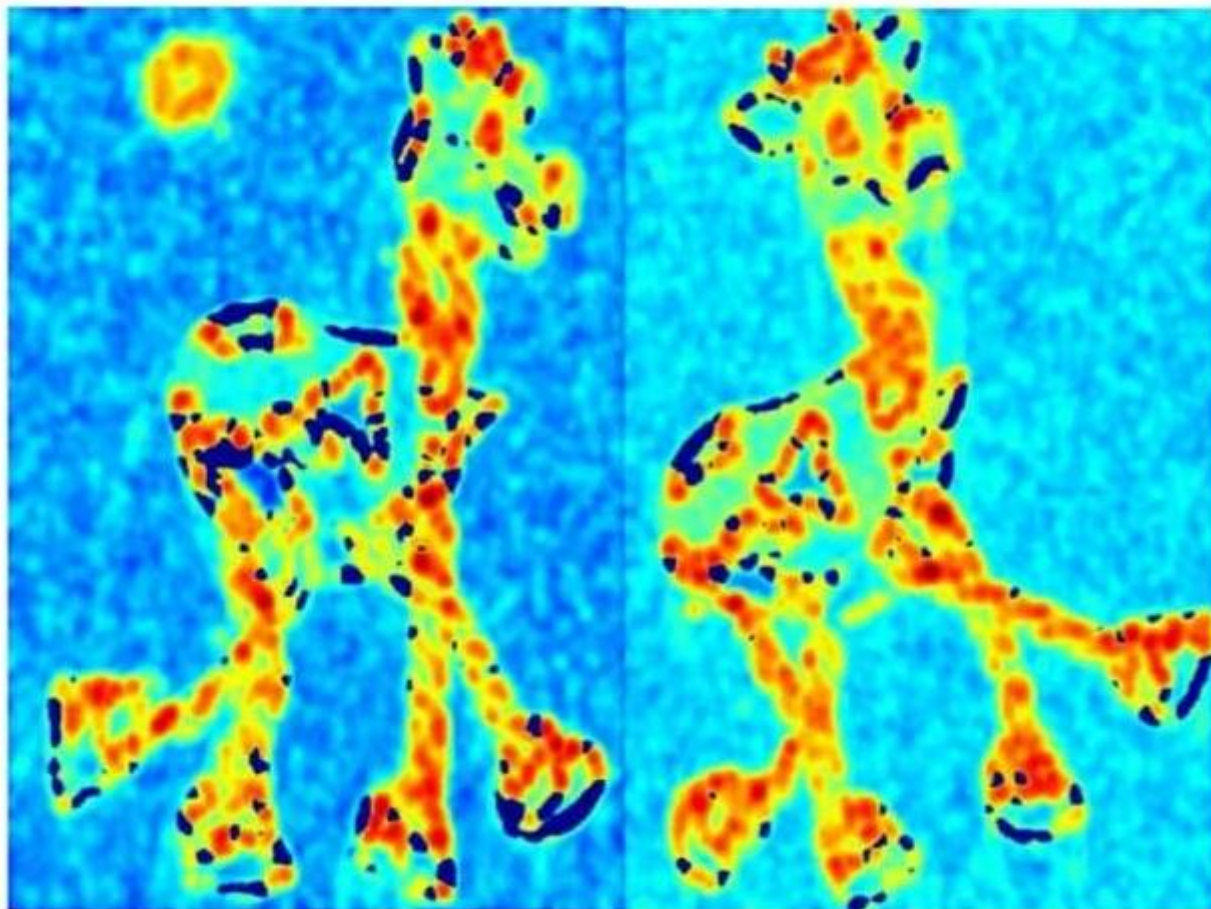
Mathematics of Harris Detector

- R depends only on eigenvalues of M
- R is large for a corner
- R is negative with large magnitude for an edge
- $|R|$ is small for a flat region

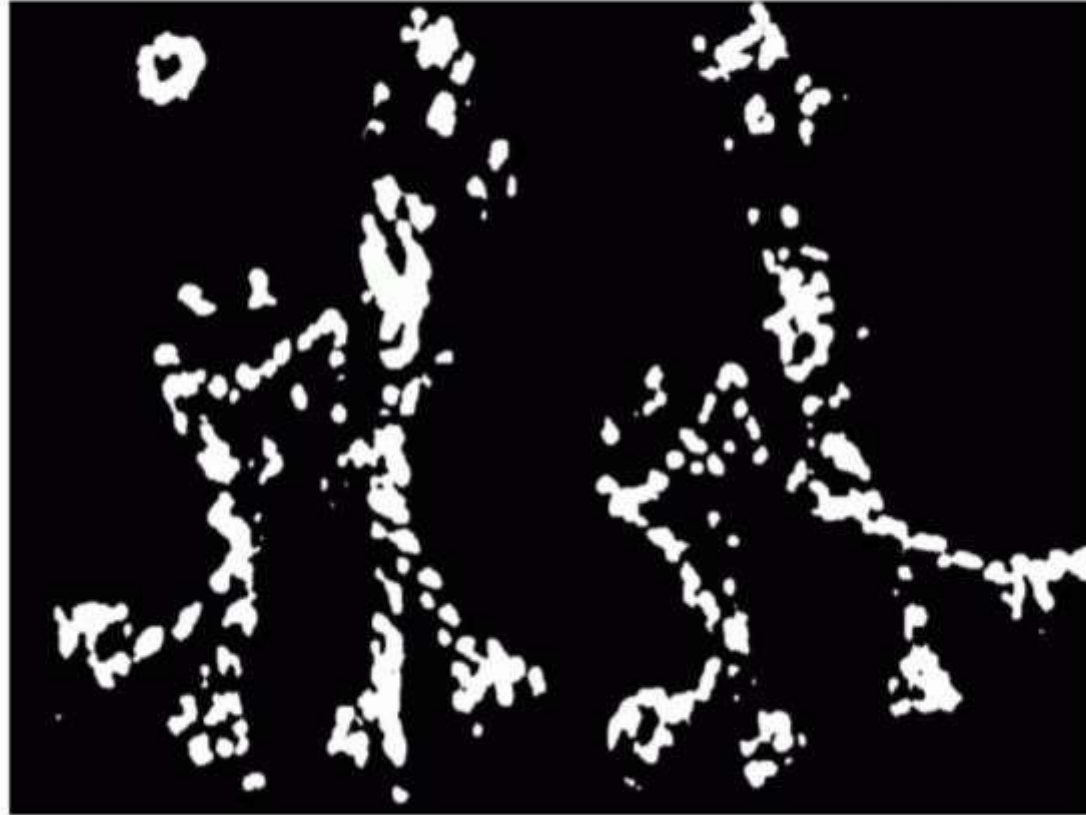




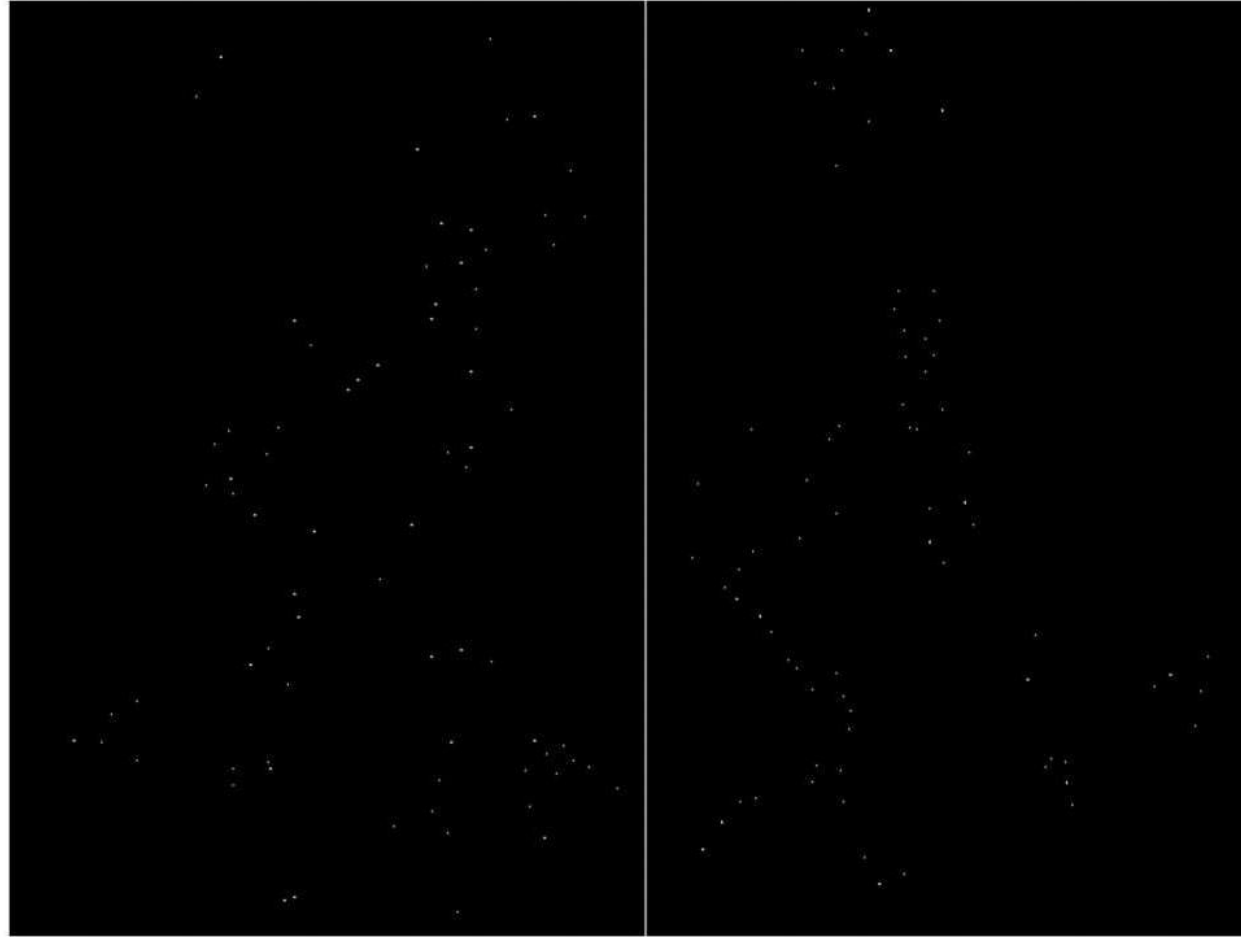
Compute corner response



Find points with large corner response: $R > \text{threshold}$



Take only the points of local maxima of R



If pixel value is greater than its neighbors then it is a local maxima.



Other Version of Harris Detectors

$$R = \lambda_1 - \alpha \lambda_2$$

Triggs

$$R = \frac{\det(M)}{\text{trace}(M)_1} = \frac{\lambda_1 \lambda_2}{\lambda_1 + \lambda_2}$$

Szeliski (Harmonic mean)

$$R = \lambda_1$$

Shi-Tomasi

