

DMET 901: Computer Vision

Corner Detectors

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Corner Detectors

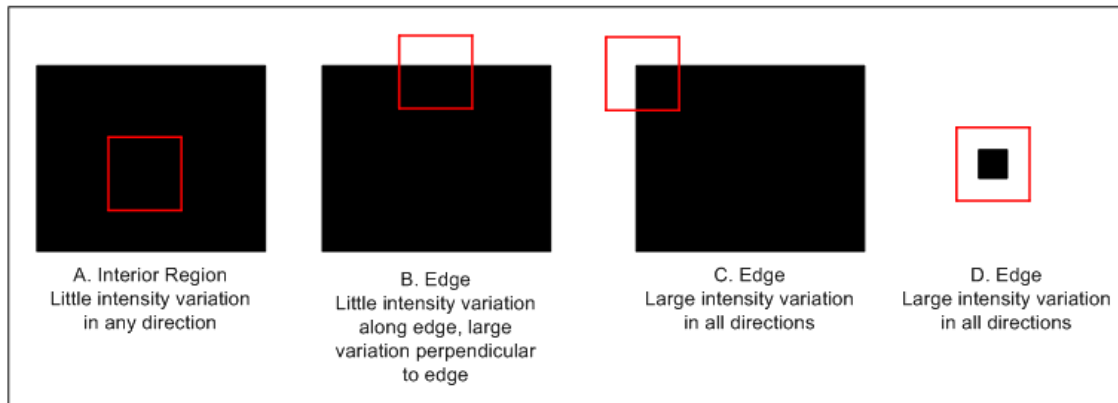
- What is a corner?
- It is the point where two lines meet forming an angular inclination.
 - Junction detection, JUDOCA.
- It is the area where any small shift, in perspective, in any direction causes a relatively large change in the scene.
 - Moravec Operator, Harris Corner Detector.
- It is the point with relatively small self esteem defined by similarity to neighborhoods.
 - SUSAN Corner Detector.

Moravec Operator

Moravec operator estimates the *cornerness* of a point by computing a measure of intensity variation in a given neighborhood w with a shift of u and v

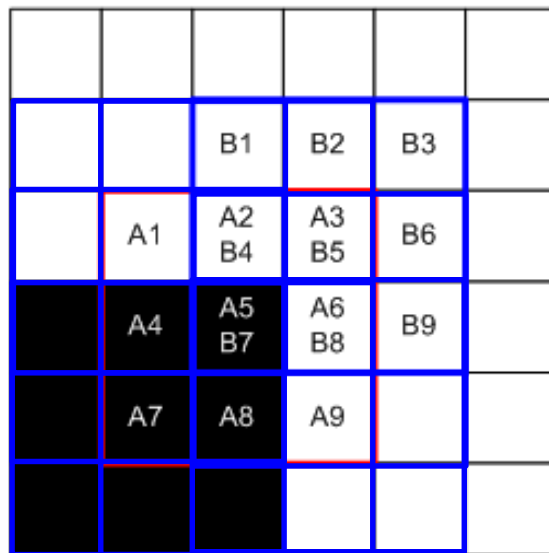
$$V(x, y)_{w(u,v)} = \sum_{i=-1}^1 \sum_{j=-1}^1 (f(x+i, y+j) - f(x+u+i, y+v+j))^2$$

Why would Moravec operator work?



Moravec Operator

- It measures the cornerness by shifting a window around the considered pixel by 1 pixel in each of the 8 principal directions and calculating the corresponding V



$$V = \sum_{i=1}^9 (A_i - B_i)^2 = 3 * 255^2$$

<http://kiwi.cs.dal.ca/~dparks/CornerDetection/algorithms.htm>

Window B Centered at

$$V = \begin{bmatrix} 3 * 255^2 \\ 2 * 255^2 \\ 3 * 255^2 \\ 2 * 255^2 \\ 5 * 255^2 \\ 2 * 255^2 \\ 3 * 255^2 \\ 2 * 255^2 \end{bmatrix} \begin{matrix} A3 \\ A2 \\ A1 \\ A4 \\ A7 \\ A8 \\ A9 \\ A6 \end{matrix}$$

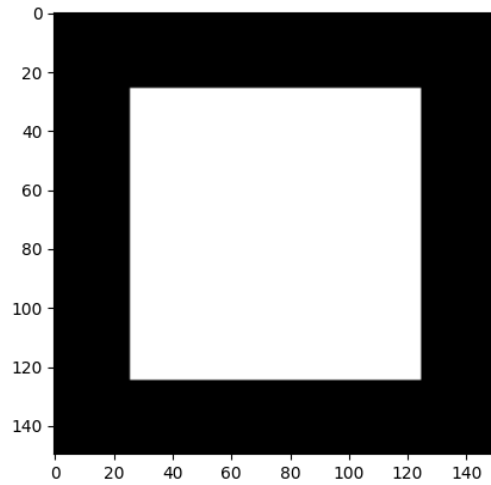
- The cornerness at a pixel (x, y)

$$C(x, y) = \min_{u, v} V(x, y)_{w(u, v)}$$

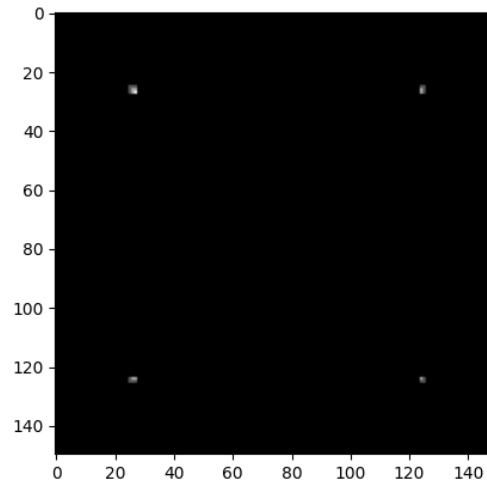
- Or Average?

Moravec Operator

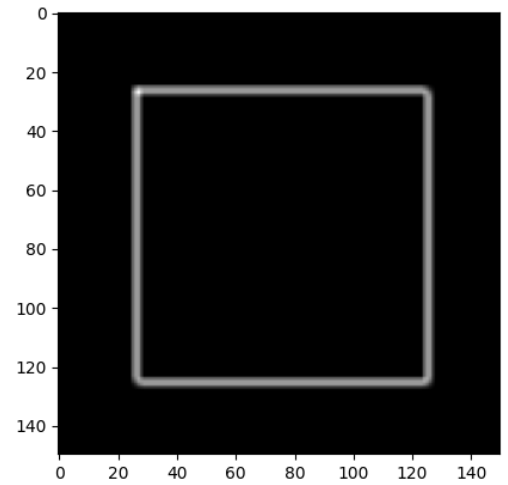
Input



Minimum



Average



Harris Corner Detector

- Improved upon Moravec operator

$$S_W(\Delta x, \Delta y) = \sum_{x_i \in W} \sum_{y_i \in W} (f(x_i, y_i) - f(x_i - \Delta x, y_i - \Delta y))^2$$

- Similar to the goal of Moravec operator, we try to find the minimum value of S_W
- This could be found analytically if the shifted image patch is approximated by the first-order Taylor expansion

$$f(x_i - \Delta x, y_i - \Delta y) \approx f(x_i, y_i) + \left[\frac{\partial f(x_i, y_i)}{\partial x}, \frac{\partial f(x_i, y_i)}{\partial y} \right] \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}$$

- Substituting in the expression for S_W

$$S_W(\Delta x, \Delta y) = \sum_{x_i \in W} \sum_{y_i \in W} \left(f(x_i, y_i) - f(x_i, y_i) - \left[\frac{\partial f(x_i, y_i)}{\partial x}, \frac{\partial f(x_i, y_i)}{\partial y} \right] \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} \right)^2$$

Harris Corner Detector

$$\begin{aligned} S_W(\Delta x, \Delta y) &= \sum_{x_i \in W} \sum_{y_i \in W} \left(f(x_i, y_i) - f(x_i, y_i) - \left[\frac{\partial f(x_i, y_i)}{\partial x}, \frac{\partial f(x_i, y_i)}{\partial y} \right] \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} \right)^2 \\ &= \sum_{x_i \in W} \sum_{y_i \in W} \left(- \left[\frac{\partial f(x_i, y_i)}{\partial x}, \frac{\partial f(x_i, y_i)}{\partial y} \right] \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} \right)^2 \\ &= \sum_{x_i \in W} \sum_{y_i \in W} \left(\left[\frac{\partial f(x_i, y_i)}{\partial x}, \frac{\partial f(x_i, y_i)}{\partial y} \right] \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} \right)^2 \\ &= \sum_{x_i \in W} \sum_{y_i \in W} [\Delta x, \Delta y] \left(\begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix} \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{bmatrix} \right) \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} \\ &= [\Delta x, \Delta y] \left(\sum_{x_i \in W} \sum_{y_i \in W} \begin{bmatrix} \frac{\partial f}{\partial x} \\ \frac{\partial f}{\partial y} \end{bmatrix} \begin{bmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{bmatrix} \right) \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} \\ &= [\Delta x, \Delta y] A_W(x, y) \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}, \end{aligned}$$

Harris Corner Detector

- The goal now is to minimize

$$S_W(\Delta x, \Delta y) = [\Delta x, \Delta y] A_W(x, y) \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}$$

- This is equivalent to finding the eigenvector of A_w corresponding to the minimum eigenvalue
- What are the eigenvectors and eigenvalues?

For any matrix L , the eigenvectors and eigenvalues are defined as

$$Lf = \lambda f \rightarrow f^T Lf = \lambda$$

f is an eigenvector of L

λ is the eigenvalue corresponding to f

- For A_w , the eigenvector is $\begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix}$ and the eigenvalue is the corresponding S_w

Harris Corner Detector

- Therefore, finding the eigenvalues of A_w would be sufficient to solve the minimization problem where

$$A(x, y) = \begin{bmatrix} \sum_{x_i \in W} \sum_{y_i \in W} \frac{\partial^2 f(x_i, y_i)}{\partial x^2} & \sum_{x_i \in W} \sum_{y_i \in W} \frac{\partial f(x_i, y_i)}{\partial x} \frac{\partial f(x_i, y_i)}{\partial y} \\ \sum_{x_i \in W} \sum_{y_i \in W} \frac{\partial f(x_i, y_i)}{\partial x} \frac{\partial f(x_i, y_i)}{\partial y} & \sum_{x_i \in W} \sum_{y_i \in W} \frac{\partial^2 f(x_i, y_i)}{\partial y^2} \end{bmatrix}$$

- Instead of computing the eigenvalues, Harris suggested using the following approximation

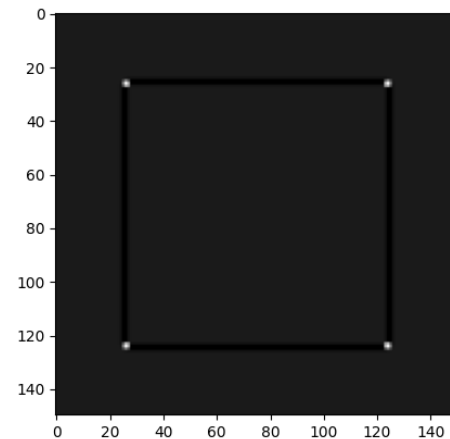
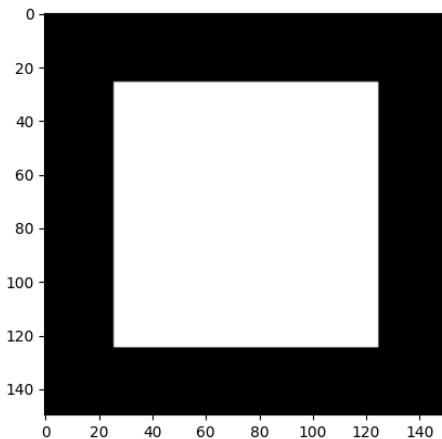
$$R(A) = \det(A) - \kappa \text{trace}^2(A)$$

where $\det(A)$ is the determinant of the local structure matrix A

$\text{trace}(A)$ is the trace of matrix A (sum of elements on the diagonal)
 κ is a tunable parameter

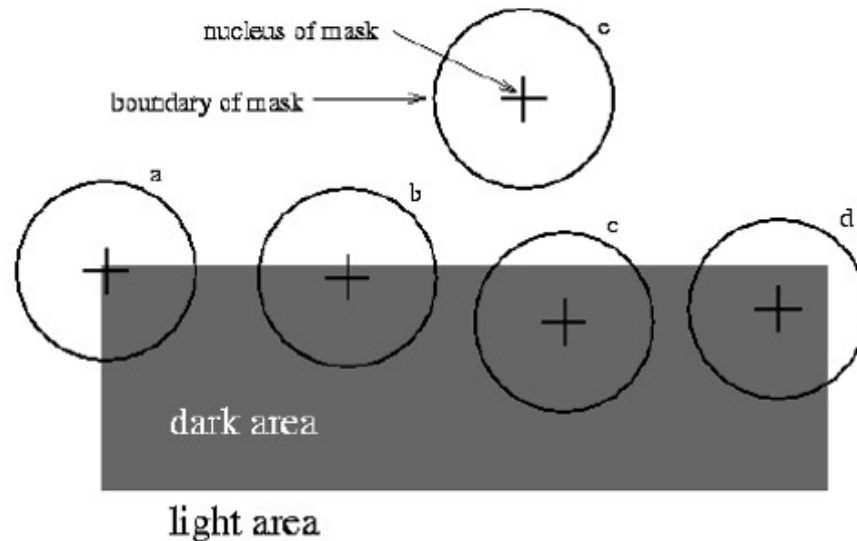
Harris Corner Detector

- Algorithm
 - Estimate intensity gradient in 2 perpendicular directions for each pixel
$$\frac{\partial f(x, y)}{\partial x} \text{ and } \frac{\partial f(x, y)}{\partial y}$$
 - For each pixel and a given neighborhood window
 - Calculate the local structure matrix A
 - Evaluate the response function $R(A)$



SUSAN Corner Detector

- **SUSAN** stands for Smallest Univalence Segment Assymilating Nucleus.
- A circular mask (having a center pixel which shall be known as the “nucleus”) is shown at five image positions.
- If the brightness of each pixel within a mask is compared with the brightness of that mask's nucleus then an area of the mask can be defined which has the same (or similar) brightness as the nucleus.
- This area of the mask shall be known as the “USAN”.
- Note that the USAN area is at a maximum when the nucleus lies in a flat homogeneous region. It falls to half of this maximum very near a straight edge, and falls even further when inside a corner.



SUSAN Corner Detector

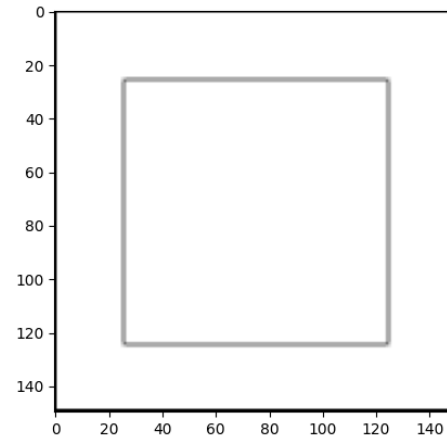
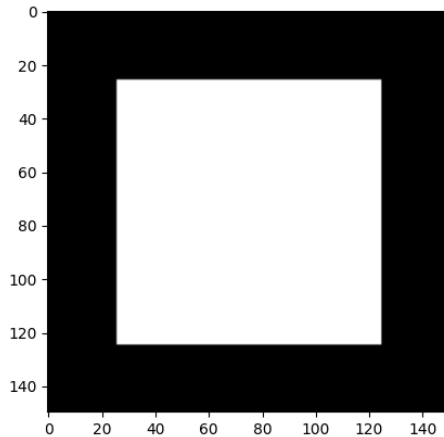
- Every pixel is compared to the nucleus using the comparison function:

$$c(\vec{m}) = e^{-\left(\frac{I(\vec{m}) - I(\vec{m}_0)}{t}\right)^6}$$

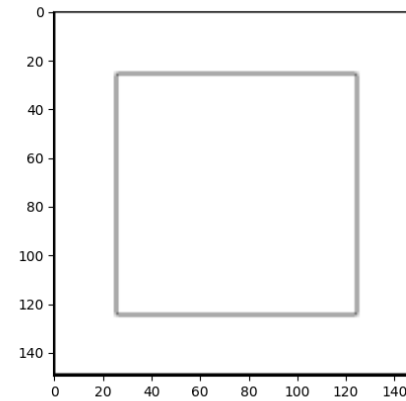
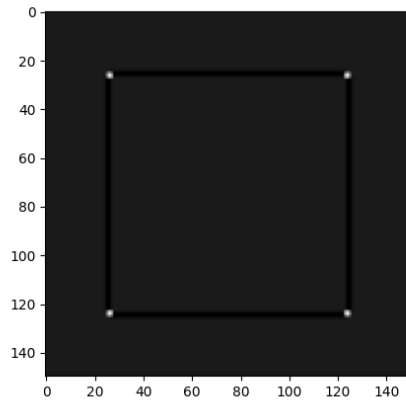
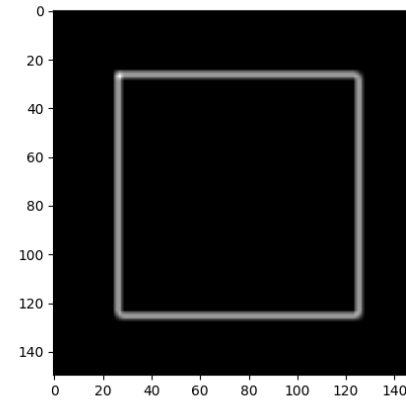
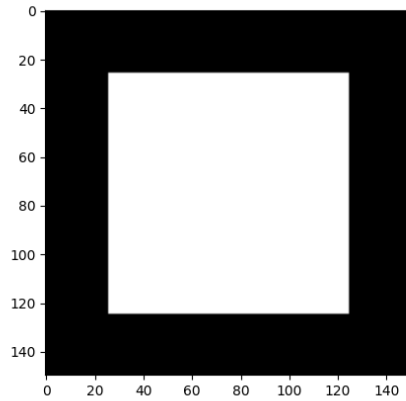
Where t is the brightness difference threshold.

- For all \vec{m} in the neighborhood of \vec{m}_0 :

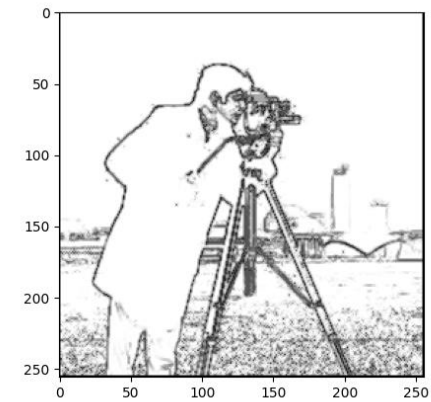
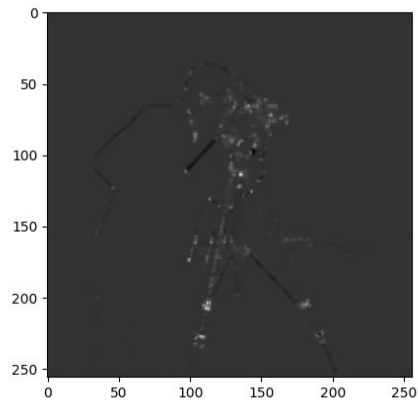
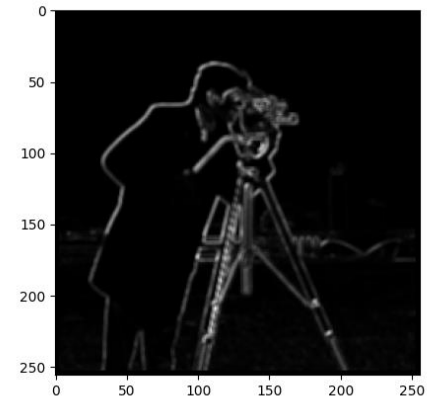
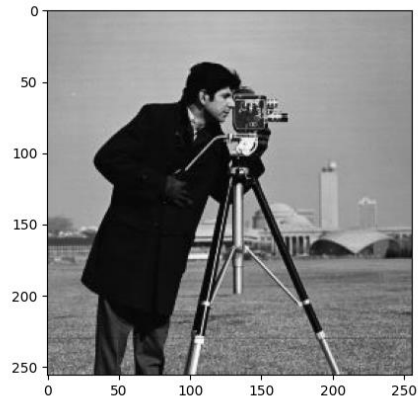
$$n(\vec{m}_0) = \sum_{\vec{m}} c(\vec{m})$$



Corner Detectors Examples



Corner Detectors Examples



Corner Detectors Examples

