

# lecture 11

Harris corners

image registration

Recall from last lecture

$$\sum_{(x,y) \in N_{\Delta}(x_0, y_0)} (I(x,y) - I(x+\Delta x, y+\Delta y))^2$$

$$\approx \sum_{(x,y) \in N_{\Delta}(x_0, y_0)} \left( \frac{\partial I(x,y)}{\partial x} \Delta x + \frac{\partial I(x,y)}{\partial y} \Delta y \right)^2$$

$$= (\Delta x, \Delta y) \underbrace{\sum_{(x,y) \in N_{\Delta}(x_0, y_0)} \begin{bmatrix} \left(\frac{\partial I}{\partial x}\right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left(\frac{\partial I}{\partial y}\right)^2 \end{bmatrix}}_M \begin{pmatrix} \Delta x \\ \Delta y \end{pmatrix}$$

$$(\Delta x \ \Delta y) M \begin{pmatrix} \Delta x \\ \Delta y \end{pmatrix} \geq 0$$

If  $\begin{pmatrix} \Delta x \\ \Delta y \end{pmatrix}$  is an unit length eigenvector of  $M$ ,

then

$$(\Delta x \ \Delta y) M \begin{pmatrix} \Delta x \\ \Delta y \end{pmatrix} = \lambda \geq 0$$

eigenvalue

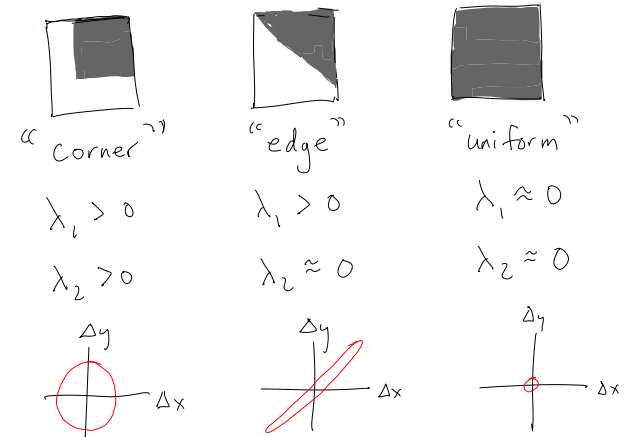
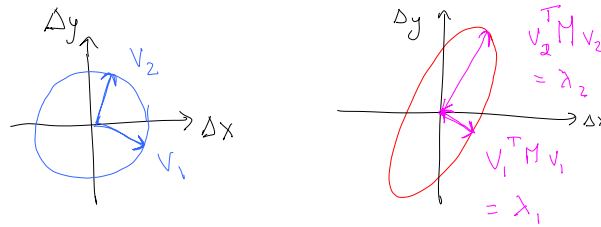
$M$  is symmetric  
 $\therefore M$  has 2 orthogonal eigenvectors.

$$v_1^T M v_2 = v_1^T M^T v_2$$

$$\begin{aligned} \leftarrow & = \lambda_2 v_1^T v_2 & \rightarrow & = v_2^T M v_1 \\ & & & = \lambda_1 v_2^T v_1 \end{aligned}$$

$$\Rightarrow \lambda_1 = \lambda_2 \text{ or } v_1^T v_2 = 0$$

How does quadratic form  $(\Delta x \ \Delta y) M \begin{pmatrix} \Delta x \\ \Delta y \end{pmatrix}$  vary with  $(\Delta x, \Delta y)$  on unit circle?



To distinguish "corner" vs. "edge" vs. "uniform" at each  $(x,y)$ .

- compute  $M = \sum_{(x,y) \in N_{\Delta}(x_0, y_0)} \begin{bmatrix} \left(\frac{\partial I}{\partial x}\right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left(\frac{\partial I}{\partial y}\right)^2 \end{bmatrix}$

- compute  $\lambda_1, \lambda_2, v_1, v_2$

$$\det M = M_{11}M_{22} - M_{12}M_{21} = \lambda_1 \lambda_2$$

$$\text{tr } M = M_{11} + M_{22} = \lambda_1 + \lambda_2$$

Interesting Cases?

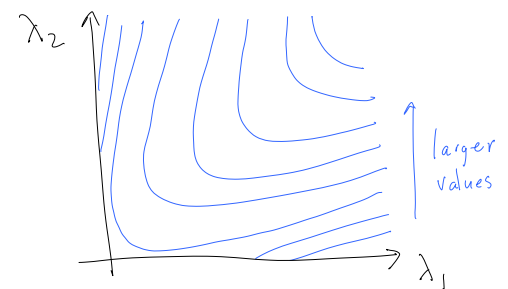
$$\det M \neq 0$$

$$\det M = 0 \text{ and } \text{tr } M \neq 0$$

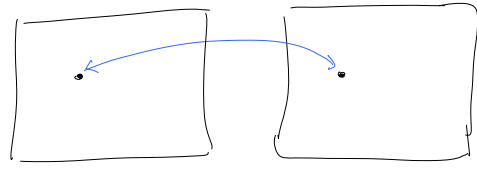
$$\det M = 0 \text{ and } \text{tr } M = 0$$

Harris (and Stevens) corner detector

$$\lambda_1 \lambda_2 - k (\lambda_1 + \lambda_2)^2 \quad k = \frac{1}{10}$$



# Image Registration

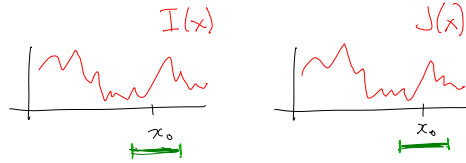


$I(x,y)$

$J(x,y)$

Find corresponding points in two images

# Image Registration in 1D



For each  $x_0$ , find  $h$  that minimizes

$$\sum_{x \in N_{gd}(x_0)} (I(x+h) - J(x))^2$$

$$\approx \sum_{x \in N_{gd}(x_0)} (I(x) + \frac{\partial I}{\partial x} h - J(x))^2$$

$$\sum_{x \in N_{gd}(x_0)} (I(x+h) - J(x))^2 \approx \sum_{x \in N_{gd}(x_0)} (I(x) + \frac{\partial I}{\partial x} h - J(x))^2$$

To minimize this, take derivative w.r.t.  $h$  and set to 0.

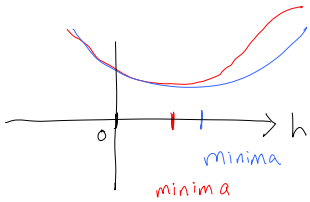
$$0 = \frac{d}{dh} \sum_{x \in N_{gd}(x_0)} (I(x) + \frac{\partial I}{\partial x} h - J(x))^2$$

$$= 2 \sum_{x \in N_{gd}(x_0)} (I(x) - J(x) + \frac{\partial I}{\partial x} h) \frac{\partial I}{\partial x}$$

$$\therefore h = - \frac{\sum_{x \in N_{gd}(x_0)} (I(x) - J(x)) \frac{\partial I}{\partial x}}{\sum_{x \in N_{gd}(x_0)} (\frac{\partial I}{\partial x})^2}$$

$$\sum_{x \in N_{gd}(x_0)} (I(x+h) - J(x))^2$$

$$\sum_{x \in N_{gd}(x_0)} (I(x) - J(x) + \frac{\partial I}{\partial x} h)^2$$



Linear model is good fit near  $h \approx 0$  but may be bad fit for  $h \neq 0$ .

# Iterative Method

Suppose we have estimated  $h_k$  such that

$$\sum_{x \in N_{gd}(x_0)} (I(x) + \frac{\partial I}{\partial x} |_{h_k} - J(x))^2$$

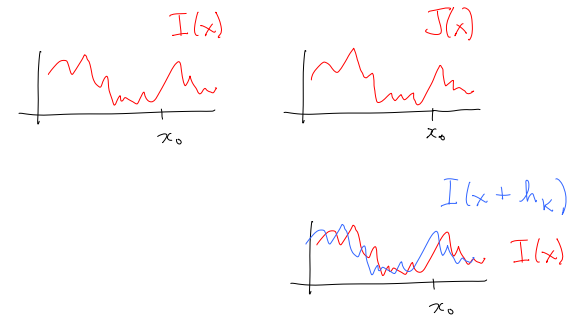
$x_0$  is minimized.

We want to find  $h_{k+1}$  such that

$$\sum_{x \in N_{gd}(x_0)} (I(x+h_{k+1}) - J(x))^2$$

is minimized.

Given  $h_k$ , find  $h_{k+1}$  that minimizes



Given  $h_k$ , find  $h_{k+1}$  that minimizes

$$\sum_{x \in N_{gd}(x_0)} (I(x+h_{k+1}) - J(x))^2 \equiv \sum_{x \in N_{gd}(x_0)} (I(x+h_k+h) - J(x))^2$$

where  $h_{k+1} \equiv h_k + h$

$$h = - \frac{\sum_{x \in N_{gd}(x_0)} (I(x+h_k) - J(x)) \frac{\partial I}{\partial x}}{\sum_{x \in N_{gd}(x_0)} (\frac{\partial I}{\partial x})^2}$$

← evaluated at  $x+h_k$

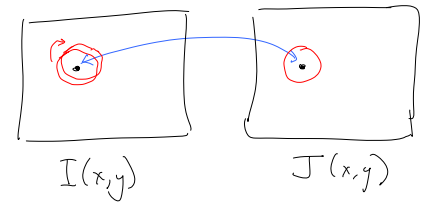
# Iterative Method

Repeat until convergence :

$$\sum_{x \in N_{gd}(x_0)} (I(x+h_k) - J(x))^2$$

$$\approx \sum_{x \in N_{gd}(x_0)} (I(x+h_{k+1}) - J(x))^2$$

# Image Registration



$I(x,y)$

$J(x,y)$

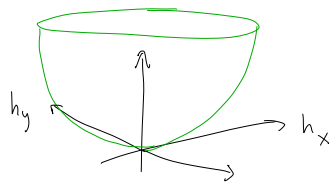
For each  $(x_0, y_0)$ , find  $(h_x, h_y)$  that minimizes

$$\sum_{(x,y) \in N_{gd}(x_0, y_0)} (I(x+h_x, y+h_y) - J(x,y))^2$$

$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} (I(x+h_x, y+h_y) - J(x,y))^2$$

$$\approx \sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \left( I(x,y) + \frac{\partial I}{\partial x} h_x + \frac{\partial I}{\partial y} h_y - J(x,y) \right)^2$$

evaluated at  $(x_0, y_0)$



To minimize, take  $\frac{\partial}{\partial h_x}$ ,  $\frac{\partial}{\partial h_y}$  of

$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \left( I(x,y) + \frac{\partial I}{\partial x} h_x + \frac{\partial I}{\partial y} h_y - J(x,y) \right)^2$$

and set to zero.

$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} 2 \left( I(x,y) + \frac{\partial I}{\partial x} h_x + \frac{\partial I}{\partial y} h_y - J(x,y) \right) \frac{\partial I}{\partial x} = 0$$

$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \left( I(x,y) + \frac{\partial I}{\partial x} h_x + \frac{\partial I}{\partial y} h_y - J(x,y) \right) \frac{\partial I}{\partial y} = 0$$

$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} 2 \left( I(x,y) + \frac{\partial I}{\partial x} h_x + \frac{\partial I}{\partial y} h_y - J(x,y) \right) \frac{\partial I}{\partial x} = 0$$

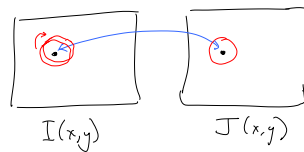
$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \left( I(x,y) + \frac{\partial I}{\partial x} h_x + \frac{\partial I}{\partial y} h_y - J(x,y) \right) \frac{\partial I}{\partial y} = 0$$

$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \begin{bmatrix} \left(\frac{\partial I}{\partial x}\right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left(\frac{\partial I}{\partial y}\right)^2 \end{bmatrix} \begin{bmatrix} h_x \\ h_y \end{bmatrix} = - \sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \begin{bmatrix} (I-J) \frac{\partial I}{\partial x} \\ (I-J) \frac{\partial I}{\partial y} \end{bmatrix}$$

Solve for  $(h_x, h_y)$ .

$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \begin{bmatrix} \left(\frac{\partial I}{\partial x}\right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left(\frac{\partial I}{\partial y}\right)^2 \end{bmatrix} \begin{bmatrix} h_x \\ h_y \end{bmatrix} = - \sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \begin{bmatrix} (I-J) \frac{\partial I}{\partial x} \\ (I-J) \frac{\partial I}{\partial y} \end{bmatrix}$$

We can solve (uniquely) for  $(h_x, h_y)$  if  $M$  is invertible, namely  $\lambda_1, \lambda_2$  both non-zero.



"corner"

$$\lambda_1 > 0$$

$$\lambda_2 > 0$$



"edge"

$$\lambda_1 > 0$$

$$\lambda_2 \approx 0$$

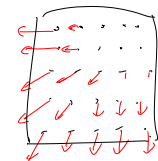


"uniform"

$$\lambda_1 \approx 0$$

$$\lambda_2 \approx 0$$

Notation



$\text{Ngd}(x_0, y_0)$

$$A = \begin{bmatrix} \uparrow & \uparrow \\ \frac{\partial I}{\partial x} & \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} & \frac{\partial I}{\partial y} \\ \downarrow & \downarrow \end{bmatrix}_{n \times 2}$$

$$M = A^T A = \sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \begin{bmatrix} \left(\frac{\partial I}{\partial x}\right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left(\frac{\partial I}{\partial y}\right)^2 \end{bmatrix}$$

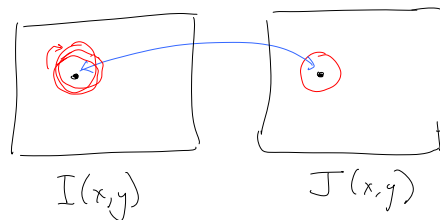
$$\sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \begin{bmatrix} \left(\frac{\partial I}{\partial x}\right)^2 & \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} \\ \frac{\partial I}{\partial x} \frac{\partial I}{\partial y} & \left(\frac{\partial I}{\partial y}\right)^2 \end{bmatrix} \begin{bmatrix} h_x \\ h_y \end{bmatrix} = - \sum_{(x,y) \in \text{Ngd}(x_0,y_0)} \begin{bmatrix} (I-J) \frac{\partial I}{\partial x} \\ (I-J) \frac{\partial I}{\partial y} \end{bmatrix}$$

$$A^T A \begin{bmatrix} h_x \\ h_y \end{bmatrix} = -A^T (\vec{I} - \vec{J})$$

Iterative Method

$$(h_x^0, h_y^0) = (0, 0)$$

$$(h_x^{k+1}, h_y^{k+1}) = (h_x^k, h_y^k) + (h_x, h_y)$$



Iterative Method

Repeat until convergence:

$$\sum_{x \in \text{Ngd}(x_0)} (I(x+h_x^k, y+h_y^k) - J(x,y))^2$$

$$\approx \sum_{x \in \text{Ngd}(x_0)} (I(x+h_x^{k+1}, y+h_y^{k+1}) - J(x,y))^2$$