

EECS332 Digital Image Analysis

# Motion Estimation

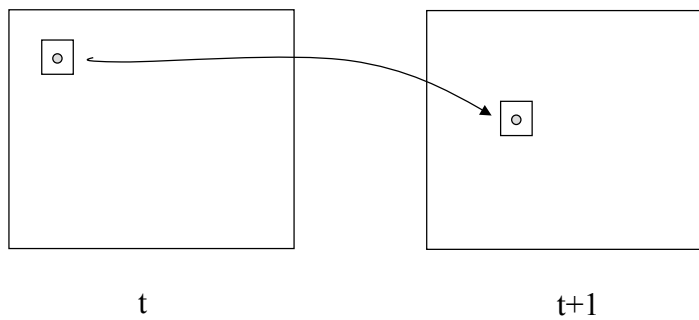
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## Motivations

- If something is moving in video, can you keep tracking its movements?
- The problem:



## Outline

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- Motivation
- Basic questions
- Exhaustive search
- Flow-constraint equation
- Gradient-based search

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## Basic questions

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- Matching ← what are the criteria for matching?
- Searching ← how to find the best match?

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## Matching Criterion

- SSD (sum of squared difference)

$$D = \sum_x \sum_y [I(x, y) - T(x, y)]^2$$

- Cross-correlation

$$C = \sum_x \sum_y I(x, y)T(x, y)$$

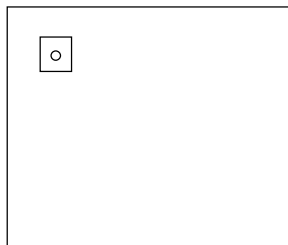
- Normalized cross-correlation

$$N = \frac{\sum_x \sum_y [I(x, y) - \bar{I}][T(x, y) - \bar{T}]}{\sqrt{[\sum_x \sum_y (I(x, y) - \bar{I})^2][\sum_x \sum_y (T(x, y) - \bar{T})^2]}}$$

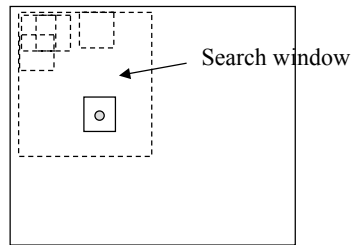
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## Exhaustive Search

- Search all locations nearby (search window)



t



t+1

Pros:

- ✓ easy to implement

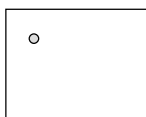
Cons:

- ✓ computationally intensive
- ✓ can not handle rotation

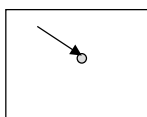
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## Gradient-based search

- Let  $I(x, y, \tau) \rightarrow$  current image
- Let  $I(x, y, 0) \rightarrow$  reference image or template
- For now, we assume a pure translational motion
- Constant Brightness Constraint:
  - i.e.,  $I(x, y, 0) = I(x+u, y+v, \tau) \quad \forall (x, y) \in \mathbb{R}$
  - where  $(u, v)$  is the displacement



$I(x, y, 0)$



$I(x+u, y+v, \tau)$

$$\begin{aligned} (u^*, v^*) &= \arg \min_{(u, v)} D(u, v) \\ &= \arg \min_{(u, v)} \sum_x \sum_y [I(x+u, y+v, \tau) - I(x, y, 0)]^2 \end{aligned}$$

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## Flow Constraint Equation

- We perform Taylor expansion of  $I(x+u, y+v, t)$  with respect to  $(x, y, 0)$ :

$$I(x+u, y+v, \tau) = I(x, y, 0) + \frac{\partial I(x, y, 0)}{\partial x} u + \frac{\partial I(x, y, 0)}{\partial y} v + \frac{\partial I(x, y, 0)}{\partial t} \tau + O(t^2)$$

- denote

$$\frac{\partial I(x, y, 0)}{\partial x} = I_x, \quad \frac{\partial I(x, y, 0)}{\partial y} = I_y, \quad \frac{\partial I(x, y, 0)}{\partial t} = I_t$$

- Since  $I(x+u, y+v, \tau) = I(x, y, 0)$



$$I_x u + I_y v + I_t \tau = 0$$

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## Solution



$$D(u, v) = \sum_x \sum_y (I_x u + I_y v + I_t \tau)^2$$

We have

$$\nabla D(u, v) = \begin{bmatrix} \sum_x \sum_y (I_x u + I_y v + I_t \tau) I_x \\ \sum_x \sum_y (I_x u + I_y v + I_t \tau) I_y \end{bmatrix} = 0$$

Easy to see

$$\sum_x \sum_y \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = -\tau \sum_x \sum_y \begin{bmatrix} I_x I_t \\ I_y I_t \end{bmatrix}$$



$$\begin{bmatrix} u \\ v \end{bmatrix} = -\tau \left( \sum_x \sum_y \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix} \right)^{-1} \left( \sum_x \sum_y \begin{bmatrix} I_x I_t \\ I_y I_t \end{bmatrix} \right)$$

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## Isn't it nice?

$$\begin{bmatrix} u \\ v \end{bmatrix} = -\tau \left( \sum_x \sum_y \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix} \right)^{-1} \left( \sum_x \sum_y \begin{bmatrix} I_x I_t \\ I_y I_t \end{bmatrix} \right)$$

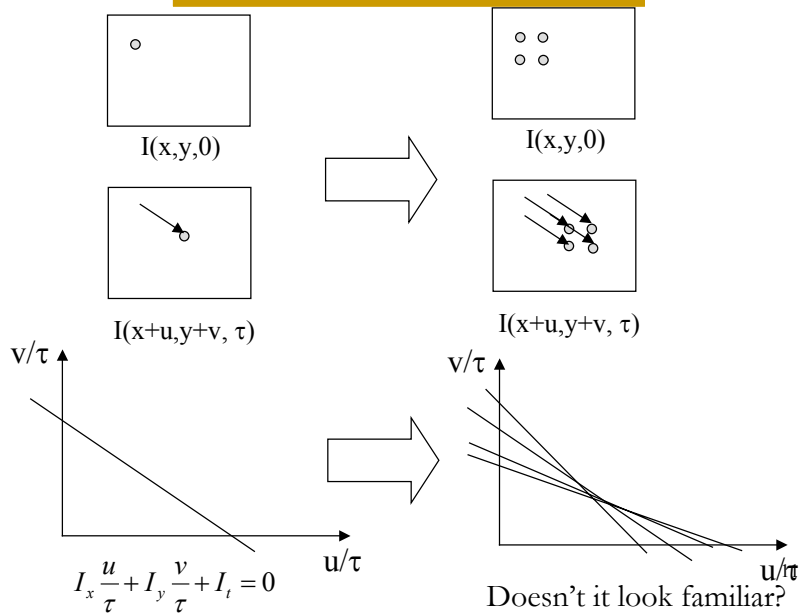
obtained from the template  
alone

$I_t$  is image  
difference

- This is a closed form solution.
- One important thing  $\rightarrow \tau$ 
  - We can not determine  $\tau$ !
  - Thus, we can only solve  $u/\tau$  and  $v/\tau \rightarrow$  velocities
  - i.e., this only provides a direction to search

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## Another explanation



## A familiar solution

$$\begin{cases} I_{x1}u + I_{y1}v + I_{t1}\tau = 0 \\ I_{x2}u + I_{y2}v + I_{t2}\tau = 0 \\ \vdots \\ I_{xN}u + I_{yN}v + I_{tN}\tau = 0 \end{cases} \Rightarrow \begin{bmatrix} I_{x1} & I_{y1} \\ I_{x2} & I_{y2} \\ \vdots & \vdots \\ I_{xN} & I_{yN} \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = -\tau \begin{bmatrix} I_{t1} \\ I_{t2} \\ \vdots \\ I_{tN} \end{bmatrix}$$

Sounds familiar?

$$Ax = b$$

You can easily figure out the solution now.

## Handling rotation?

- Assume a pure rotation

$$R(\theta) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

objective function

$$D(\theta) = \sum_x \sum_y \left[ I(R(\theta) \begin{bmatrix} x \\ y \end{bmatrix}, \tau) - I(x, y, 0) \right]^2$$

Taylor expansion

$$I(R(\theta) \begin{bmatrix} x \\ y \end{bmatrix}, \tau) = I(x, y, 0) + \frac{\partial I}{\partial \theta} \theta + \frac{\partial I}{\partial t} \tau + o(t^2)$$

where  $\frac{\partial I}{\partial \theta} = -\frac{\partial I}{\partial x} y + \frac{\partial I}{\partial y} x = I_\theta$

derivative

$$D(\theta) = \sum_x \sum_y (I_\theta \theta + I_t \tau)^2 \Rightarrow \nabla D(\theta) = \sum_x \sum_y (I_\theta \theta + I_t \tau) I_\theta$$

solution

$$\Rightarrow \theta = -\tau \frac{\sum_x \sum_y I_\theta I_t}{\sum_x \sum_y I_\theta^2}$$