

Optical Flow

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#13

Optical Flow

Where did each pixel in image 1 go to in image 2 ?

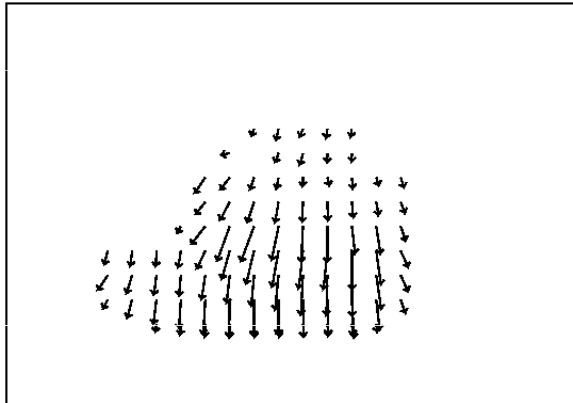


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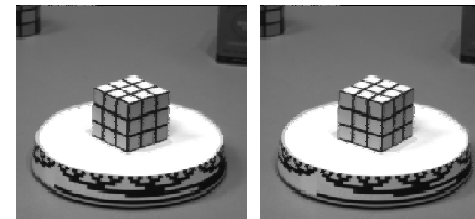
Pierre Kornprobst's Demo

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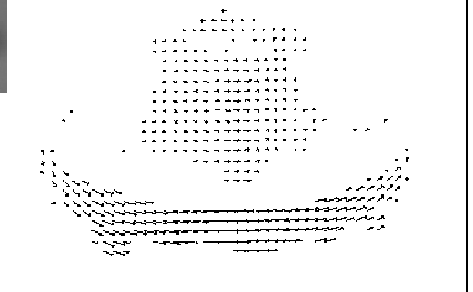
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▪ We are interested in finding the movement of scene objects from time-varying images (videos).



• Will start by estimating motion of *each pixel separately*
Then will consider motion of *entire* image

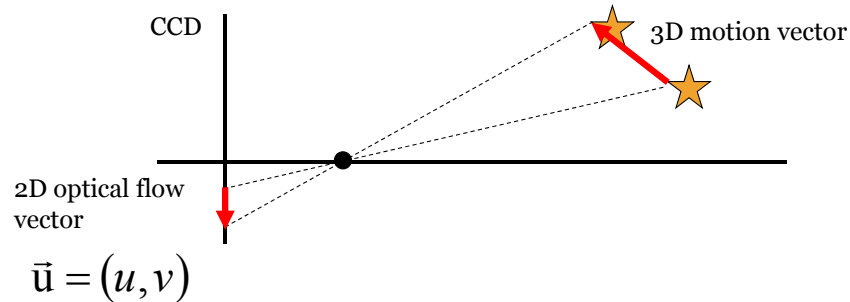
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Motion Field & Optical Flow Field

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- **Motion Field** = Real world 3D motion
- **Optical Flow Field** = Projection of the motion field onto the 2d image

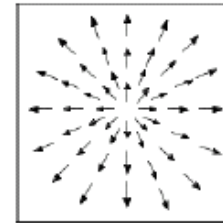


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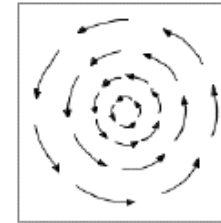
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Motion Field & Optical Flow Field

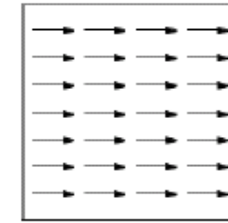
6



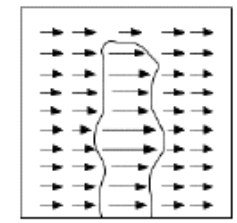
Forward motion



Rotation



Horizontal translation



Closer objects appear to move faster!!

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Scene Interpretation

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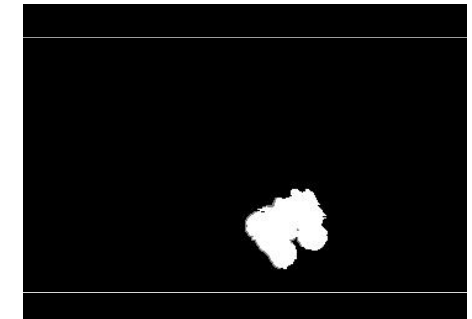
- Given a video sequence with camera/objects moving we can better understand the scene if we find the motions of the camera/objects.
 - How is the camera moving?
 - How many moving objects are there?
 - Which directions are they moving in?
 - How fast are they moving?
 - Can we recognize their type of motion (e.g. walking, running, etc.)?

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Motion Segmentation

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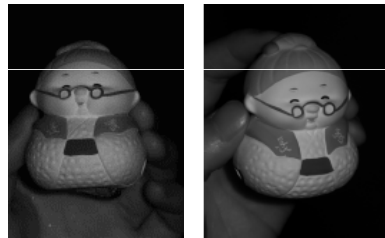


Result by: L.Zelnik-Manor, M.Machline, M.Irani
"Multi-body Segmentation: Revisiting Motion Consistency", IJCV 2006

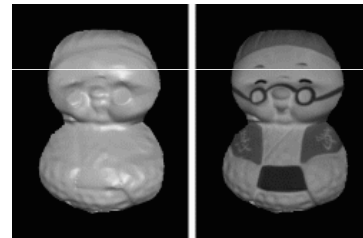
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3D Shape Reconstruction⁹



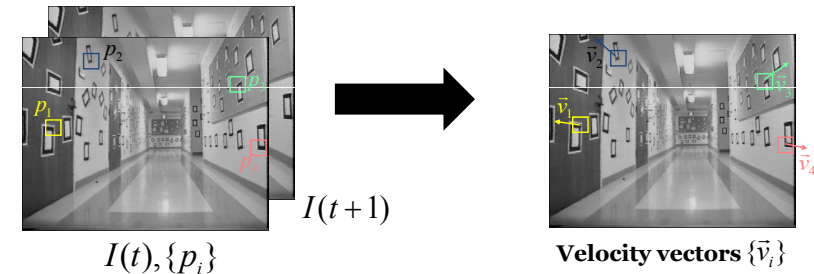
Input



Reconstructed shape

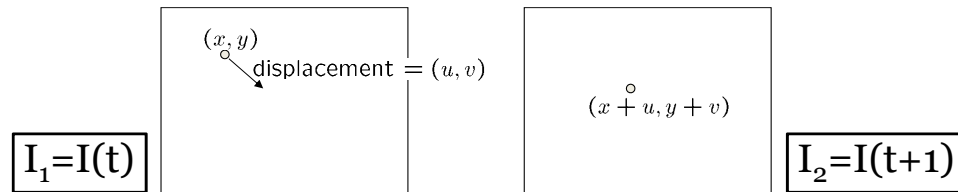
Result by: L. Zhang, B. Curless, A. Hertzmann, S.M. Seitz
 "Shape and motion under varying illumination: Unifying structure from motion, photometric stereo, and multi-view stereo" ICCV'03

Optical Flow¹⁰



Goal: Find for *each pixel* a velocity vector

Optical Flow¹¹



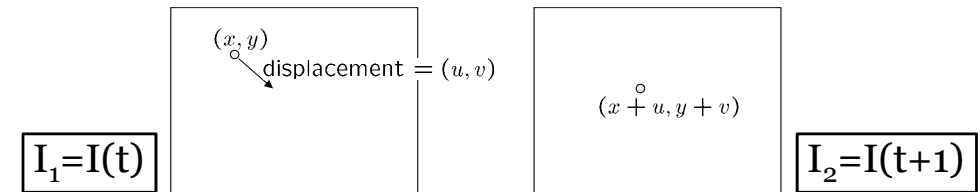
How to estimate pixel motion from image I_1 to image I_2 ?

- Solve pixel correspondence problem
 - given a pixel in I_1 , look for **nearby** pixels of the **same color** in I_2

Key assumptions

- color constancy:** a point in I_1 looks the same in I_2
 - For grayscale images, this is brightness constancy
- small motion:** points do not move very far

Optical Flow¹²



Color constancy

$$I_1(x, y) = I_2(x + u, y + v)$$

Small motion

$$I_2(x + u, y + v) \approx I_2(x, y) + \frac{\partial I_2}{\partial x} u + \frac{\partial I_2}{\partial y} v$$

Approximated by Taylor series expansion

Optical Flow

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$$I_1(x, y) \approx I_2(x, y) + \frac{\partial I_2}{\partial x} u + \frac{\partial I_2}{\partial y} v$$

$$0 \approx I_2(x, y) - I_1(x, y) + \frac{\partial I_2}{\partial x} u + \frac{\partial I_2}{\partial y} v$$

$$0 \approx I(t+1) - I(t) + \frac{\partial I_2}{\partial x} u + \frac{\partial I_2}{\partial y} v$$

$$0 \approx I_t + I_x \cdot u + I_y \cdot v$$

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We want to determine u and v for each pixel

$$0 \approx \underbrace{I_t + I_x \cdot u + I_y \cdot v}_{\text{This can be calculated from images}}$$

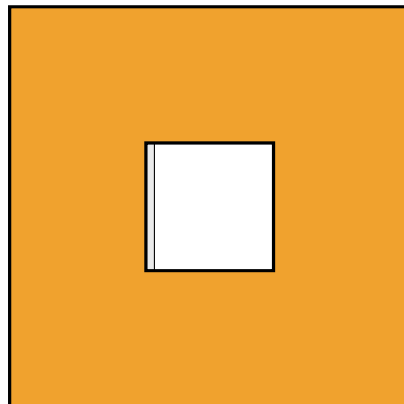
2 unknown
with only one equation

Need More
Constraints

Aperture Problem

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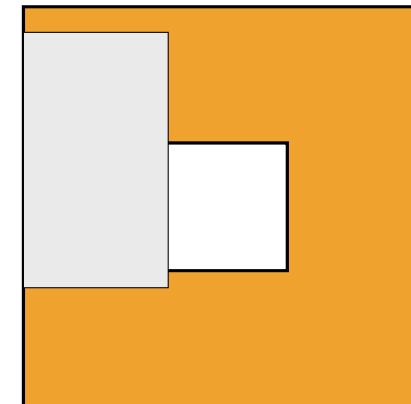
- How does this show up visually ?



Aperture Problem

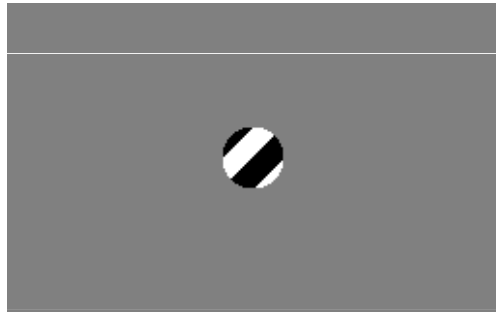
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- Motion along just an edge is ambiguous



The Barber's Pole Illusion

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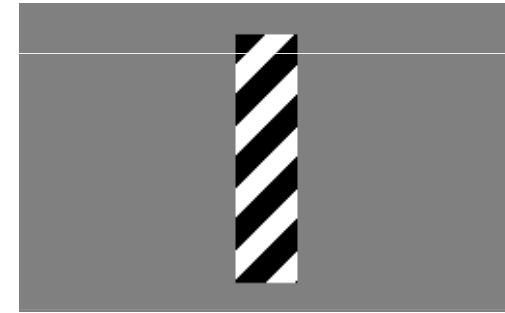
http://en.wikipedia.org/wiki/Barberpole_illusion

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The Barber's Pole Illusion

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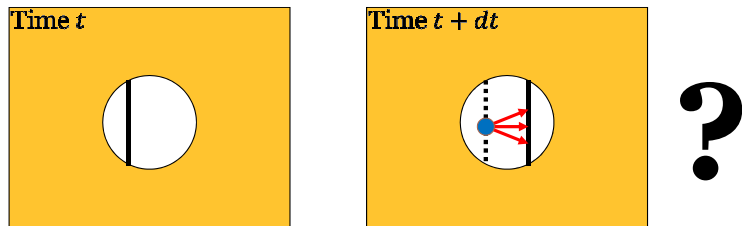
http://en.wikipedia.org/wiki/Barberpole_illusion

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Aperture Problem

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Where did the blue point move to?

We need additional constraints

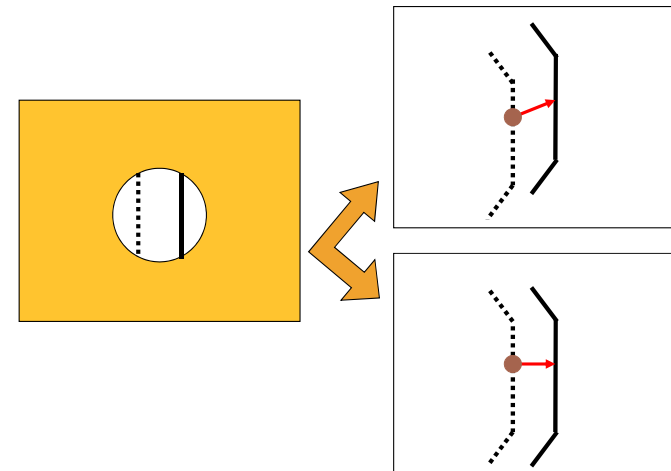
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Aperture Problem

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- Sometimes enlarging the aperture can help



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Horn-Schunck Method

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assume that the flow field is smooth locally

$$E_{HS}(u, v) = \sum_{x, y} \left[(I_t + I_x \cdot u + I_y \cdot v)^2 + \lambda \left(\left(\frac{\partial u}{\partial x} \right)^2 + \left(\frac{\partial u}{\partial y} \right)^2 + \left(\frac{\partial v}{\partial x} \right)^2 + \left(\frac{\partial v}{\partial y} \right)^2 \right) \right]$$

Regularization
parameter

smoothness in the flow
over the whole image

Lucas-Kanade Method

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pretend the pixel's neighbors have the same (u, v)

If we use a 5×5 window, that gives us 25 equations per pixel!

$$I_t(x_1, y_1) + I_x(x_1, y_1) \cdot u + I_y(x_1, y_1) \cdot v \approx 0$$

$$I_t(x_2, y_2) + I_x(x_2, y_2) \cdot u + I_y(x_2, y_2) \cdot v \approx 0$$

\vdots

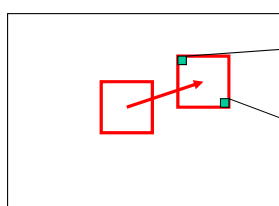
$$I_t(x_{25}, y_{25}) + I_x(x_{25}, y_{25}) \cdot u + I_y(x_{25}, y_{25}) \cdot v \approx 0$$

Find best u, v

Lucas-Kanade Method

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Assume constant (u, v) in small neighborhood



$$\begin{bmatrix} I_{x1} & I_{y1} \\ I_{x2} & I_{y2} \\ \vdots & \vdots \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = - \begin{bmatrix} I_{t1} \\ I_{t2} \\ \vdots \end{bmatrix}$$

$$\mathbf{A}\mathbf{u} = \mathbf{b}$$

Lucas-Kanade Method

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Find \mathbf{u} that minimize $\|\mathbf{A}\mathbf{u} - \mathbf{b}\|^2$



$$\mathbf{u}_{opt} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$$

Edge $\rightarrow \mathbf{A}^T \mathbf{A}$ becomes singular
Homogeneous $\rightarrow \mathbf{A}^T \mathbf{A} \approx 0$ (low gradient)
High texture $\rightarrow \mathbf{A}^T \mathbf{A} = \text{thumbs up}$

Coarse-to-fine Flow Estimation²⁵

