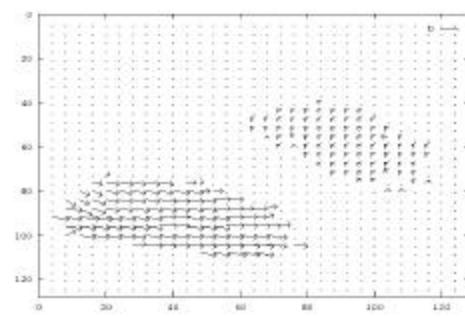
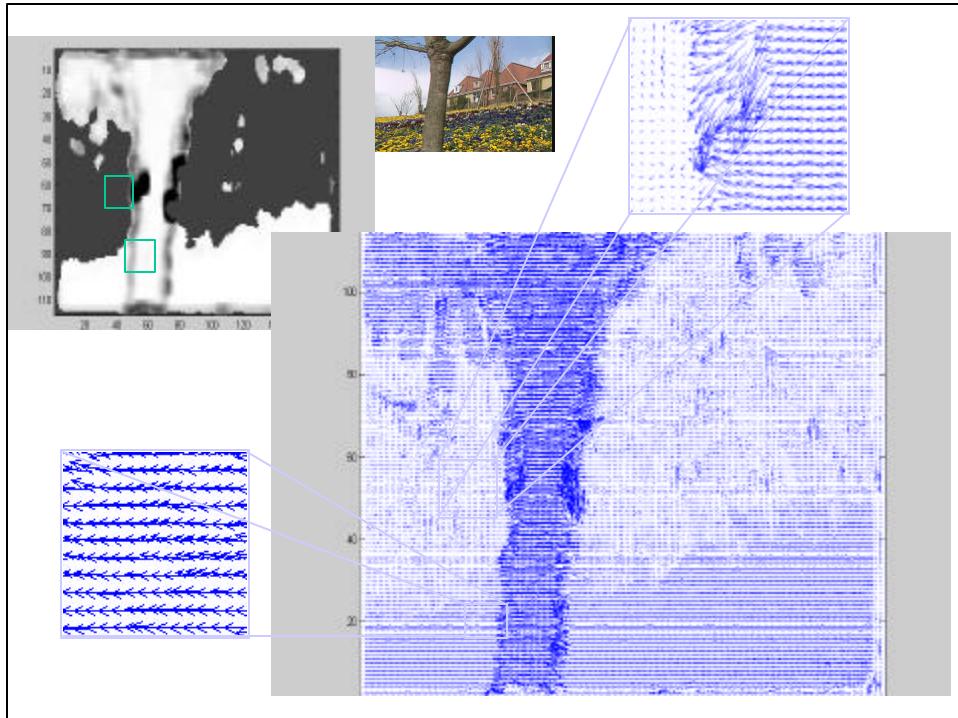
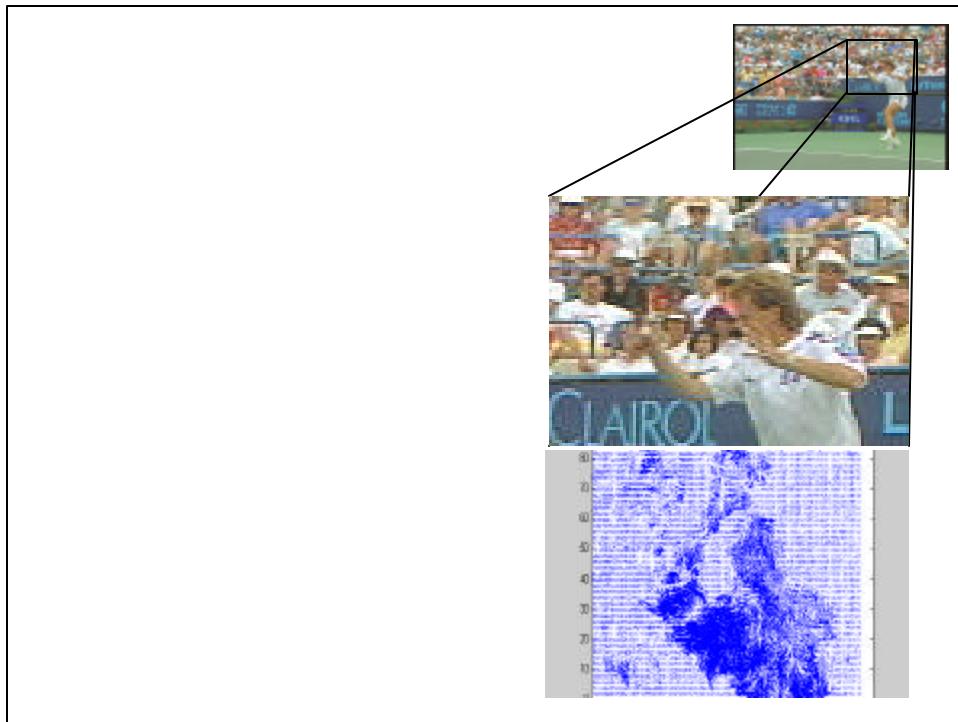


Lecture-3

Computing Optical Flow

Hamburg Taxi seq





Horn&Schunck Optical Flow

$$\frac{df(x, y, t)}{dt} = \frac{\partial f}{\partial x} \frac{dx}{dt} + \frac{\partial f}{\partial y} \frac{dy}{dt} + \frac{\partial f}{\partial t} = 0$$

brightness constancy eq

Horn&Schunck Optical Flow

$$f(x, y, t) = f(x + dx, y + dy, t + dt)$$

↓ Taylor Series

$$f(x, y, t) = f(x, y, t) + \frac{\partial f}{\partial x} dx + \frac{\partial f}{\partial y} dy + \frac{\partial f}{\partial t} dt$$

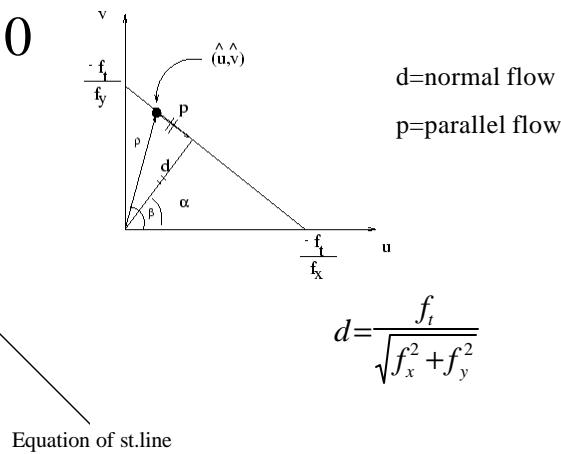
$$f_x dx + f_y dy + f_t dt = 0$$

brightness constancy eq

Interpretation of optical flow eq

$$f_x u + f_y v + f_t = 0$$

$$v = -\frac{f_x}{f_y}u - \frac{f_t}{f_y}$$



$$d = \sqrt{\frac{f_t^2}{f_x^2 + f_y^2}}$$

Horn&Schunck (contd)

$$\iint \{(f_x u + f_y v + f_t)^2 + I(u_x^2 + u_y^2 + v_x^2 + v_y^2)\} dx dy$$

\downarrow min variational calculus

$$(f_x u + f_y v + f_t) f_x + I(\Delta^2 u) = 0 \quad u = u_{av} - f_x \frac{P}{D}$$

$$(f_x u + f_y v + f_t) f_y + I(\Delta^2 v) = 0 \quad v = v_{av} - f_y \frac{P}{D}$$

\downarrow discrete version

$$(f_x u + f_y v + f_t) f_x + I(u - u_{av}) = 0 \quad P = f_x u_{av} + f_y v_{av} + f_t$$

$$(f_x u + f_y v + f_t) f_y + I(v - v_{av}) = 0 \quad D = I + f_x^2 + f_y^2$$

Algorithm-1

- k=0
- Initialize $u^K \quad v^K$
- Repeat until some error measure is satisfied

$$u^K = u_{av}^{k-1} - f_x \frac{P}{D} \quad P = f_x u_{av} + f_y v_{av} + f_t$$
$$v = v_{av}^{K-1} - f_y \frac{P}{D} \quad D = I + f_x^2 + f_y^2$$

Derivatives

- Derivative: Rate of change of some quantity
 - Speed is a rate of change of a distance
 - Acceleration is a rate of change of speed

Derivative

$$\frac{df}{dx} = \lim_{\Delta x \rightarrow 0} \frac{f(x) - f(x - \Delta x)}{\Delta x} = f'(x) = f_x$$

$$v = \frac{ds}{dt} \quad \text{speed}$$

$$a = \frac{dv}{dt} \quad \text{acceleration}$$

Examples

$$y = x^2 + x^4 \quad y = \sin x + e^{-x}$$

$$\frac{dy}{dx} = 2x + 4x^3 \quad \frac{dy}{dx} = \cos x + (-1)e^{-x}$$

Second Derivative

$$\frac{df_x}{dx} = f''(x) = f_{xx}$$

$$y = x^2 + x^4$$

$$\frac{dy}{dx} = 2x + 4x^3$$

$$\frac{d^2y}{dx^2} = 2 + 12x^2$$

Discrete Derivative

$$\frac{df}{dx} = \lim_{\Delta x \rightarrow 0} \frac{f(x) - f(x - \Delta x)}{\Delta x} = f'(x)$$

$$\frac{df}{dx} = \frac{f(x) - f(x - 1)}{1} = f'(x)$$

$$\frac{df}{dx} = f(x) - f(x - 1) = f'(x)$$

(Finite Difference)

Discrete Derivative

$$\frac{df}{dx} = f(x) - f(x-1) = f'(x) \quad \text{Left difference}$$

$$\frac{df}{dx} = f(x) - f(x+1) = f'(x) \quad \text{Right difference}$$

$$\frac{df}{dx} = f(x+1) - f(x-1) = f'(x) \quad \text{Center difference}$$

Example

	F(x)=10	10	10	10	20	20	20
Left difference	F'(x)=0	0	0	0	10	0	0
	F''(x)=0	0	0	0	10	-10	0

-1 1 left difference
1 -1 right difference
-1 0 1 center difference

Derivatives in Two Dimensions

$f(x, y)$

(partial Derivatives) $\frac{\partial f}{\partial x} = f_x = \lim_{\Delta x \rightarrow 0} \frac{f(x, y) - f(x - \Delta x, y)}{\Delta x}$

$\frac{\partial f}{\partial y} = f_y = \lim_{\Delta y \rightarrow 0} \frac{f(x, y) - f(x, y - \Delta y)}{\Delta y}$

(f_x, f_y) Gradient Vector

magnitude $= \sqrt{(f_x^2 + f_y^2)}$

direction $= q = \tan^{-1} \frac{f_y}{f_x}$

$\Delta^2 f = f_{xx} + f_{yy}$ = Laplacian

Derivatives of an Image

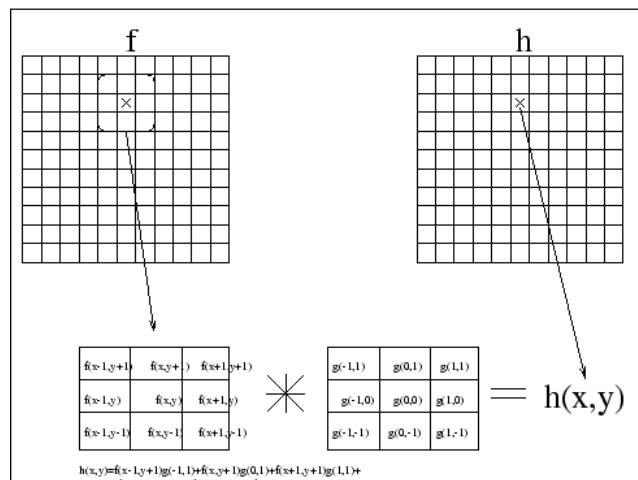
	-1 0 1	-1 -1 -1	
Derivative & average	-1 0 1	0 0 0	Prewit
	-1 0 1	1 1 1	
	f_x	f_y	

$$I(x, y) = \begin{bmatrix} 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \end{bmatrix} \quad I_x = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 30 & 30 & 0 & 0 \\ 0 & 30 & 30 & 0 & 0 \\ 0 & 30 & 30 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Derivatives of an Image

$$I(x, y) = \begin{bmatrix} 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \\ 10 & 10 & 20 & 20 & 20 \end{bmatrix} \quad I_y = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Convolution



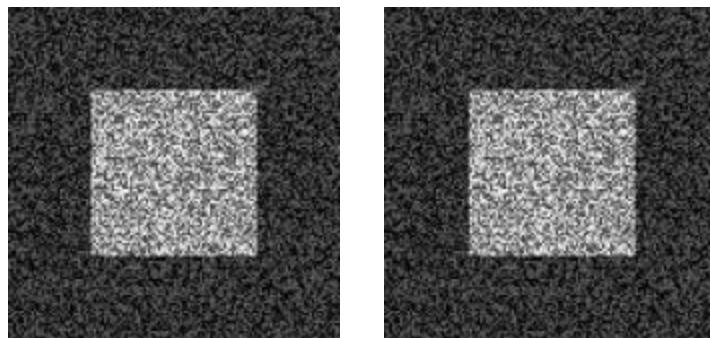
Convolution (contd)

$$h(x, y) = \sum_{i=-1}^1 \sum_{j=-1}^1 f(x+i, y+j) g(i, j)$$
$$h(x, y) = f(x, y) * g(x, y)$$

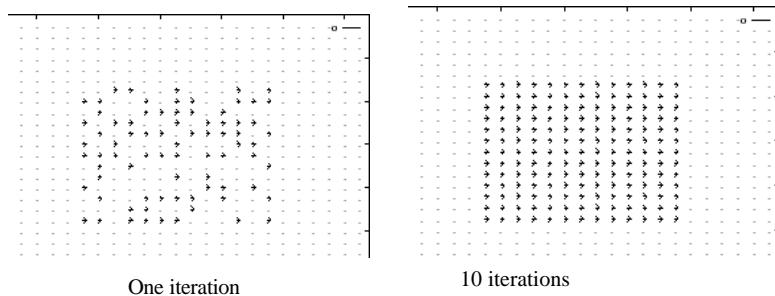
Derivative Masks

$$\begin{array}{c} \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}_{\text{first image}} \quad \begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix}_{\text{first image}} \quad \begin{bmatrix} -1 & -1 \\ -1 & -1 \end{bmatrix}_{\text{first image}} \\ \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}_{\text{second image}} \quad \begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix}_{\text{second image}} \quad \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}_{\text{second image}} \\ f_x \qquad \qquad \qquad f_y \qquad \qquad \qquad f_t \end{array}$$

Synthetic Images



Results



Homework Due 9/5/00

- Derive Euler Angles matrix from three rotations around x, y and Z. (Lecture-1, slide 27).
- Derive bi-quadratic motion model from the projective motion model using Taylor series. (Lecture-2, slide 12).
- Verify 3-D rigid motion using instantaneous motion model can be written as a cross product of rotational velocities Ω and object location (X). Lecture-2, slide 22.
- Verify that pseudo perspective motion model can be derived assuming planar scene and perspective projection. Lecture-2, Slide 26.