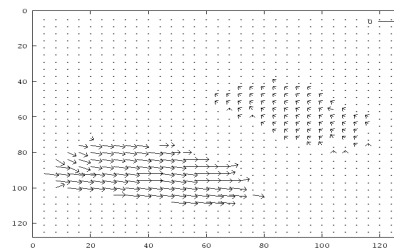
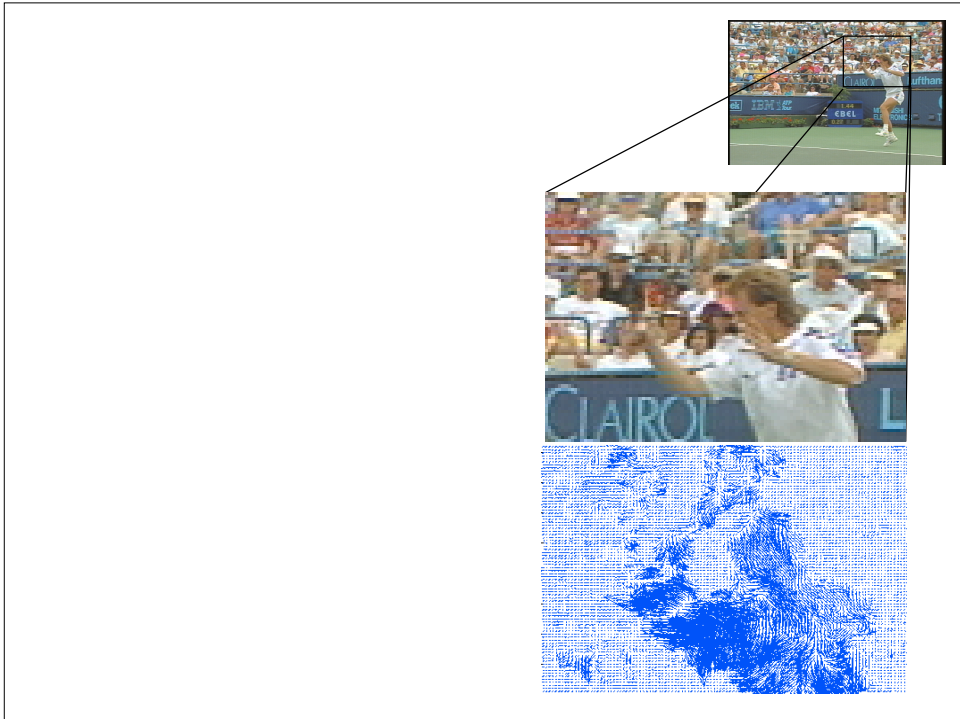
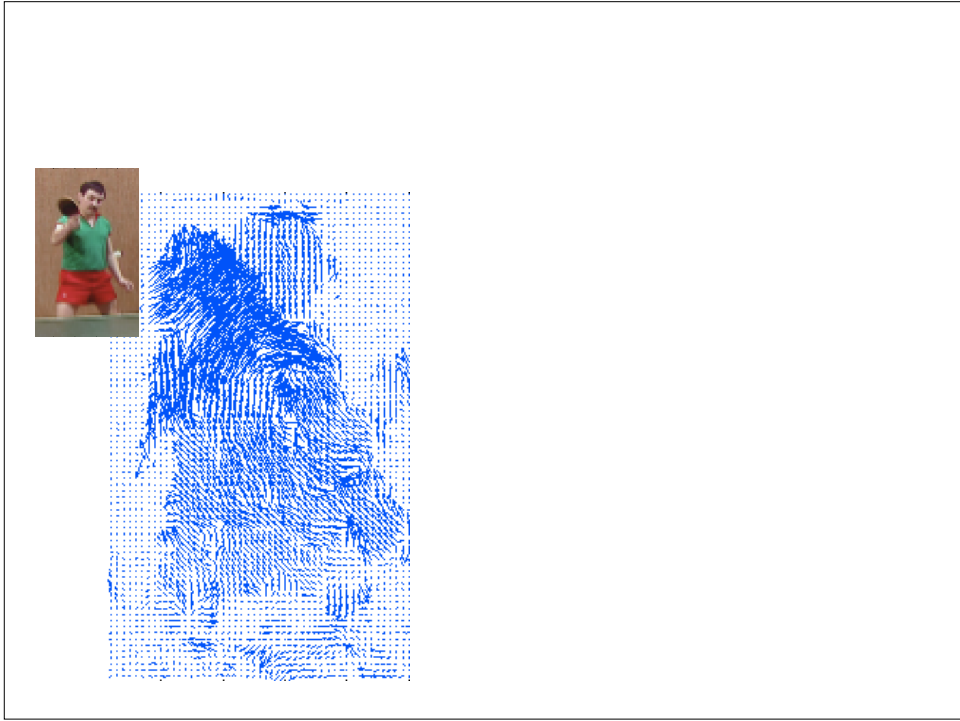


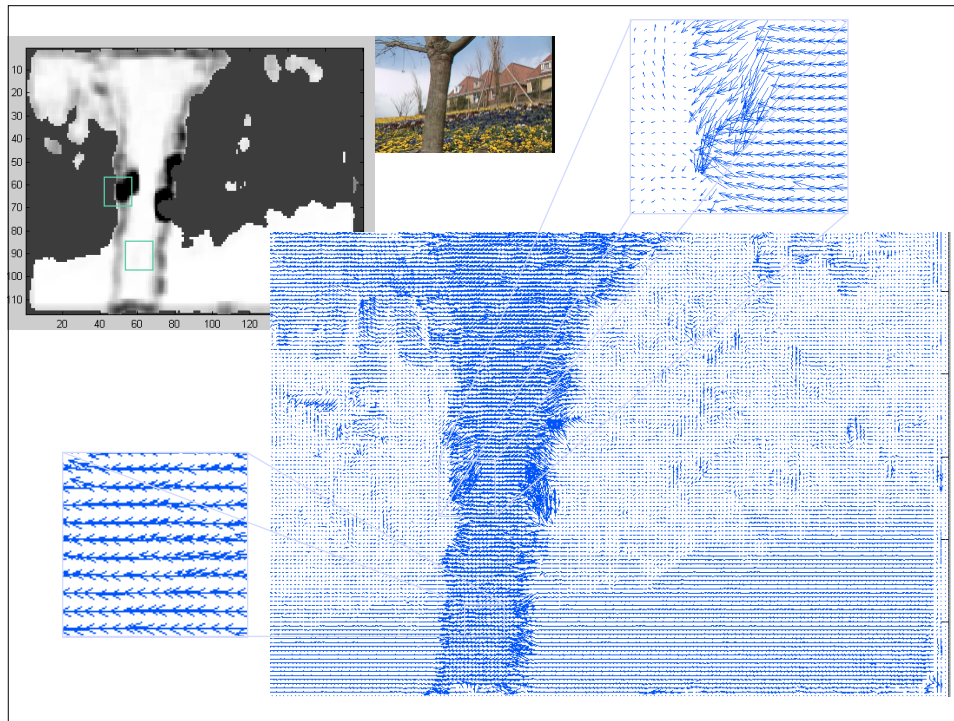
# Lecture-3

## Computing Optical Flow

### Hamburg Taxi seq







## Horn&Schunck Optical Flow

$f(x, y, t)$  Image Sequence

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

brightness constancy eq

## Horn&Schunck Optical Flow

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

↓ Taylor Series

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

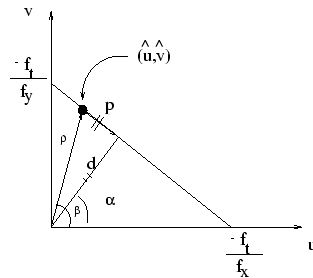
$$f_x dx + f_y dy + f_t dt = 0$$

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=normal flow

p=parallel flow

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

Equation of st.line

## Horn&Schunck (contd)

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



min

variational calculus

$$(f_x u + f_y v + f_t) f_x + \lambda(\nabla^2 u) = 0$$

$$u = u_{av} - \lambda f_x \frac{P}{D}$$

$$(f_x u + f_y v + f_t) f_y + \lambda(\nabla^2 v) = 0$$

$$v = v_{av} - \lambda f_y \frac{P}{D}$$



discrete version

$$(f_x u + f_y v + f_t) f_x + \lambda(u \nabla^2 u_{av}) = 0$$

$$P = f_x u_{av} + f_y v_{av} + f_t$$

$$D = \lambda + f_x^2 + f_y^2$$

$$(f_x u + f_y v + f_t) f_y + \lambda(v \nabla^2 v_{av}) = 0$$

$$\nabla^2 u = u_{xx} + u_{yy}$$

## Algorithm-1

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

$$u^k = u_{av}^{k-1} - \lambda f_x \frac{P}{D}$$

$$P = f_x u_{av} + f_y v_{av} + f_t$$

$$v^k = v_{av}^{k-1} - \lambda f_y \frac{P}{D}$$

$$D = \lambda + f_x^2 + f_y^2$$

## 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

$$\frac{d^2y}{dx^2} = f''(x) = f_{xx}$$

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

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

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

## 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	$F'(x)=0$	0	0	0	10	0	0
difference	$F''(x)=0$	0	0	0	10	-10	0

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

## Derivatives in Two Dimensions

(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

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

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

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

## Derivatives of an Image

	$\begin{bmatrix} 1 & 0 & 1 \\ 1 & 0 & 1 \\ 1 & 0 & 1 \end{bmatrix}$	$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 1 & 1 & 1 \end{bmatrix}$	Prewit
Derivative & average	$f_x$	$f_y$	

$I(x, y) = \begin{bmatrix} 0 & 10 & 20 & 20 & 20 \\ 0 & 10 & 20 & 20 & 20 \\ 0 & 10 & 20 & 20 & 20 \\ 0 & 10 & 20 & 20 & 20 \\ 0 & 10 & 20 & 20 & 20 \end{bmatrix}$	$I_x = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \\ 0 & 30 & 30 & 0 & 0 \\ 0 & 30 & 30 & 0 & 0 \\ 0 & 30 & 30 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$
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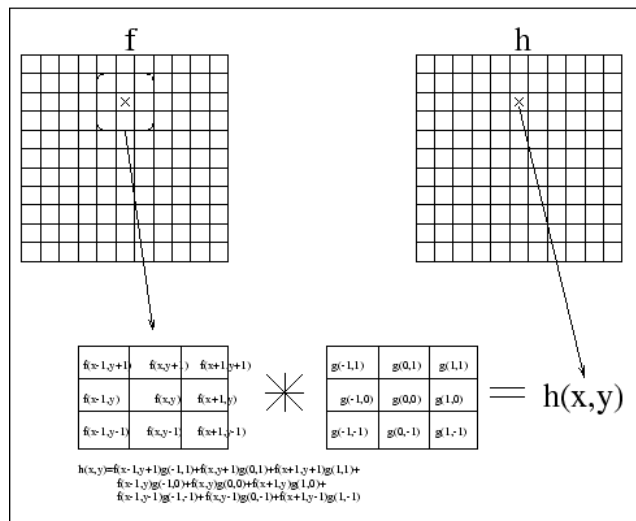
## Derivatives of an Image

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

$$\begin{array}{ccccc}
 & & 0 & \square \frac{1}{4} & 1 \\
 & & \square \frac{1}{4} & 1 & \square \frac{1}{4} \\
 & & 0 & \square \frac{1}{4} & 0 \\
 & & & & f_{xx} + f_{yy}
 \end{array}$$

# Convolution



## Convolution (contd)

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

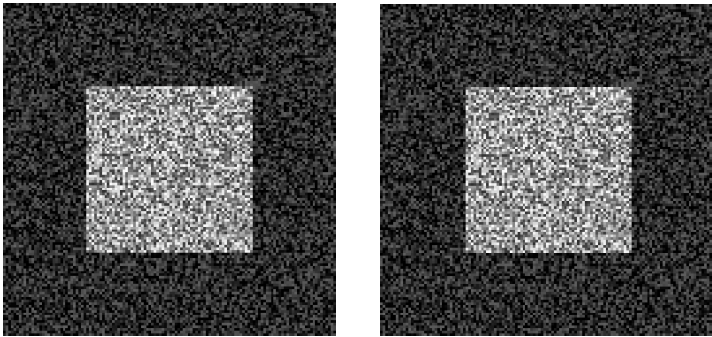
$$h(x, y) = f(x, y) * g(x, y)$$

$$\begin{aligned} h(x, y) = & f(x-1, y-1)g(-1, -1) + f(x, y-1)g(0, -1) + f(x+1, y-1)g(1, -1) \\ & + f(x-1, y)g(-1, 0) + f(x, y)g(0, 0) + f(x+1, y)g(1, 0) \\ & + f(x-1, y+1)g(-1, 1) + f(x, y+1)g(0, 1) + f(x+1, y+1)g(1, 1) \end{aligned}$$

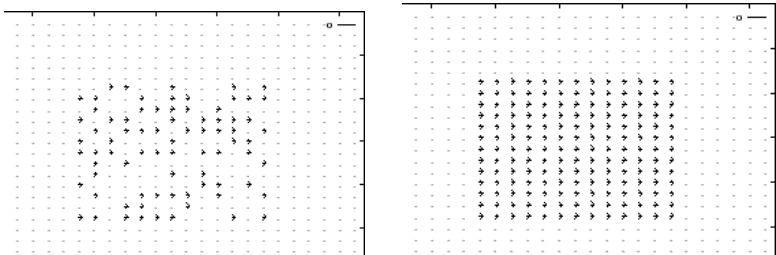
## Derivative Masks

$\begin{array}{cc} \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \\ \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \end{array}$ <p>first image</p>	$\begin{array}{cc} \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \\ \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \end{array}$ <p>first image</p>	$\begin{array}{cc} \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \\ \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \end{array}$ <p>first image</p>
$\begin{array}{cc} \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \\ \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \end{array}$ <p>second image</p>	$\begin{array}{cc} \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \\ \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \end{array}$ <p>second image</p>	$\begin{array}{cc} \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \\ \begin{array}{ c } \hline 1 \\ \hline \end{array} & \begin{array}{ c } \hline 1 \\ \hline \end{array} \end{array}$ <p>second image</p>
$f_x$	$f_y$	$f_t$

# Synthetic Images



# Results



One iteration

10 iterations

$$\square = 4$$

## Homework Due 9/9/02

- Derive Euler Angles matrix from three rotations around x, y and Z. (Lecture-2, page 3, do not hand in).
- Derive bi-quadratic motion model from the projective motion model using Taylor series. (Lecture-2, page 11).
- Verify 3-D rigid motion using instantaneous motion model can be written as a cross product of rotational velocities and object location (X). Lecture-2, page 16.
- Verify that pseudo perspective motion model can be derived assuming planar scene and perspective projection. Lecture-2, page 18.