

# Lecture-17

## Computing Optical Flow: Lucas & Kanade Global Flow

### Lucas & Kanade (Least Squares)

- Optical flow eq

$$f_x u + f_y v = f_t$$

- Consider 3 by 3 window

$$f_{x1} u + f_{y1} v = f_{t1}$$

⋮

$$f_{x9} u + f_{y9} v = f_{t9}$$

$$\mathbf{A} \mathbf{u} = \mathbf{f}_t$$

## Lucas & Kanade

$$\mathbf{A}\mathbf{u} = \mathbf{f}_t$$

$$\mathbf{A}^T \mathbf{A}\mathbf{u} = \mathbf{A}^T \mathbf{f}_t$$

$$\mathbf{u} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{f}_t$$



$$\min \sum (f_{xi}u + f_{yi}v + f_t)^2$$

## Lucas & Kanade

$$\min \sum (f_{xi}u + f_{yi}v + f_t)^2$$



$$\sum (f_{xi}u + f_{yi}v + f_t)f_{xi} = 0$$

$$\sum (f_{xi}u + f_{yi}v + f_t)f_{yi} = 0$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} (f_{xi}u + f_{yi}v + f_{ti})f_{xi} = 0$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} (f_{xi}u + f_{yi}v + f_{ti})f_{yi} = 0$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} f_{xi}^2 u + \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} f_{xi}f_{yi}v = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} f_{xi}f_{ti}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} f_{xi}f_{yi}u + \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} f_{yi}^2 v = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} f_{yi}f_{ti}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} f_{xi}^2 \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \end{bmatrix} u + \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \\ f_{xi}f_{yi} \end{bmatrix} v = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \\ f_{xi}f_{ti} \end{bmatrix}$$

## Lucas & Kanade

$$\begin{bmatrix} u \\ v \end{bmatrix} = \frac{1}{\begin{bmatrix} f_{xi}^2 & f_{xi}f_{yi} \\ f_{xi}f_{yi} & f_{yi}^2 \end{bmatrix}} \begin{bmatrix} f_{xi}f_{ti} \\ f_{yi}f_{ti} \end{bmatrix}$$

$$u = \frac{f_{yi}^2 f_{xi}f_{ti} + f_{xi}f_{yi} f_{yi}f_{ti}}{f_{xi}^2 f_{yi}^2 + (f_{xi}f_{yi})^2}$$

$$v = \frac{f_{xi}f_{ti} f_{xi}f_{yi} + f_{xi}^2 f_{yi}f_{ti}}{f_{xi}^2 f_{yi}^2 + (f_{xi}f_{yi})^2}$$

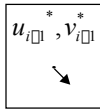
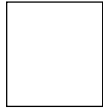
## Comments

- Horn-Schunck and Lucas-Kanade optical method works only for small motion.
- If object moves faster, the brightness changes rapidly, 2x2 or 3x3 masks fail to estimate spatiotemporal derivatives.
- Pyramids can be used to compute large optical flow vectors.

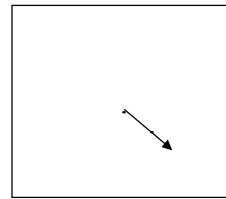
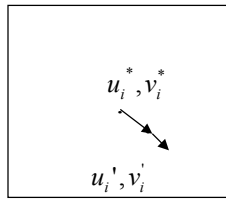
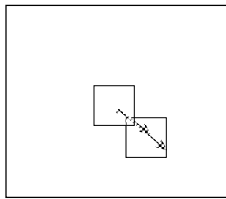
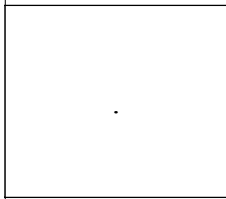
## Lucas Kanade with Pyramids

- Compute ‘simple’ LK at highest level
- At level  $i$ 
  - Take flow  $u_{i-1}, v_{i-1}$  from level  $i-1$
  - bilinear interpolate it to create  $u_i^*, v_i^*$  matrices of twice resolution for level  $i$
  - multiply  $u_i^*, v_i^*$  by 2
  - compute  $f_t$  from a block displaced by  $u_i^*(x,y), v_i^*(x,y)$
  - Apply LK to get  $u_i'(x, y), v_i'(x, y)$  (the correction in flow)
  - Add corrections  $u_i', v_i'$ , *i.e.*  $u_i = u_i^* + u_i'$ ,  
 $v_i = v_i^* + v_i'$ .

# Pyramids



$$u_i = u_i^* + u_i', v_i = v_i^* + v_i'$$



$f_1$  pyramid

$f_2$  pyramid

# Interpolation

	0	1	2	3
0	•	•	•	•
$u = 1$	•	•	•	•
2	•	•	•	•
3	•	•	•	•

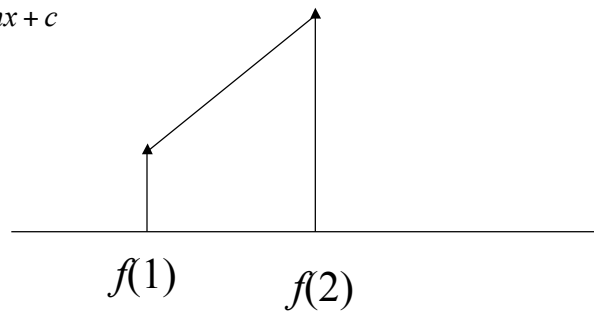
	0	1	2	3
0	•	•	•	•
$v = 1$	•	•	•	•
2	•	•	•	•
3	•	•	•	•

	0	1	2	3	4	5	6	7
0	•	◦	•	◦	•	◦	•	◦
1	◦	◦	◦	◦	◦	◦	◦	◦
2	•	◦	•	◦	•	◦	•	◦
$u^* = 3$	◦	◦	◦	◦	◦	◦	◦	◦
4	•	◦	•	◦	•	◦	•	◦
5	◦	◦	◦	◦	◦	◦	◦	◦
6	•	◦	•	◦	•	◦	•	◦
7	◦	◦	◦	◦	◦	◦	◦	◦

	0	1	2	3	4	5	6	7
0	•	◦	•	◦	•	◦	•	◦
1	◦	◦	◦	◦	◦	◦	◦	◦
2	•	◦	•	◦	•	◦	•	◦
$v^* = 3$	◦	◦	◦	◦	◦	◦	◦	◦
4	•	◦	•	◦	•	◦	•	◦
5	◦	◦	◦	◦	◦	◦	◦	◦
6	•	◦	•	◦	•	◦	•	◦
7	◦	◦	◦	◦	◦	◦	◦	◦

## 1-D Interpolation

$$y = mx + c$$
$$f(x) = mx + c$$



## 2-D Interpolation

$$f(x, y) = a_1 + a_2x + a_3y + a_4xy$$

Bilinear

X	X
O	X

## Bi-linear Interpolation

**Four nearest points of (x,y) are:**

$$(\underline{x}, \underline{y}), (\bar{x}, \underline{y}), (\underline{x}, \bar{y}), (\bar{x}, \bar{y})$$

$$(3,5), (4,5), (3,6), (4,6)$$

$$\underline{x} = \text{int}(x) \quad 3 \quad (3.2, 5.6)$$

$$\underline{y} = \text{int}(y) \quad 5 \quad X_{(3,6)} \quad X_{(4,6)}$$

$$\bar{x} = \underline{x} + 1 \quad 4 \quad X_{(3,5)}^{\circ} \quad X_{(4,5)}$$

$$\bar{y} = \underline{y} + 1 \quad 6$$

$$f(\underline{x}, \underline{y}) = \bar{\bar{\alpha}}_x \bar{\bar{\alpha}}_y f(\underline{x}, \underline{y}) + \underline{\bar{\alpha}}_x \bar{\bar{\alpha}}_y f(\bar{x}, \underline{y}) +$$

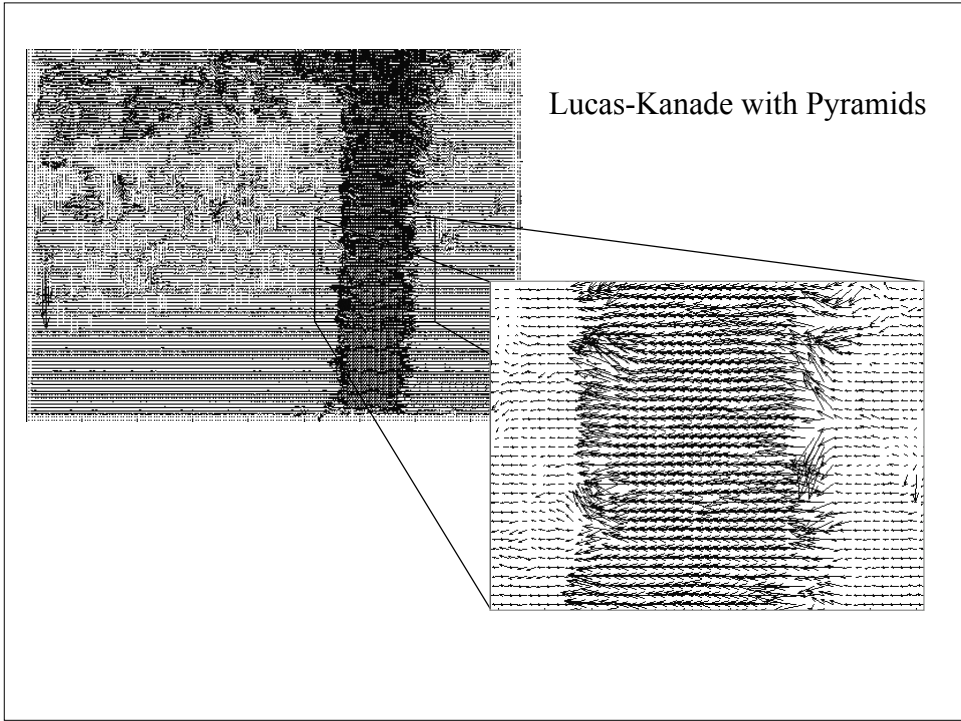
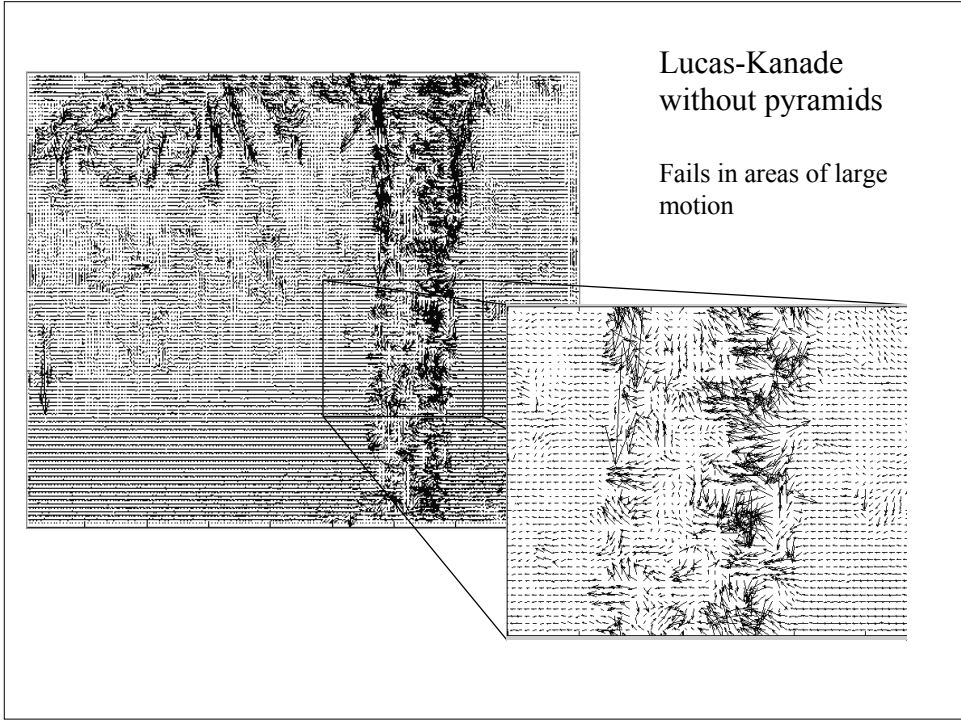
$$\bar{\bar{\alpha}}_x \underline{\bar{\alpha}}_y f(\underline{x}, \bar{y}) + \underline{\bar{\alpha}}_x \underline{\bar{\alpha}}_y f(\bar{x}, \bar{y})$$

$$\bar{\bar{\alpha}}_x = \bar{x} - x \quad \bar{\bar{\alpha}}_x = \bar{x} - x = 4 - 3.2 = .8$$

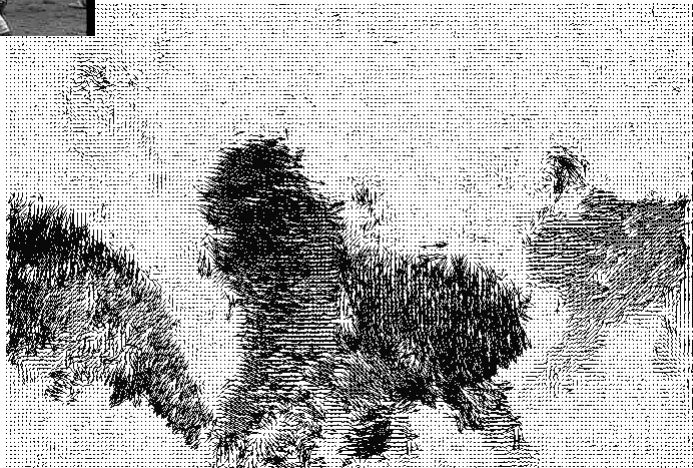
$$\bar{\bar{\alpha}}_y = \bar{y} - y \quad \bar{\bar{\alpha}}_y = \bar{y} - y = 6 - 5.6 = .4$$

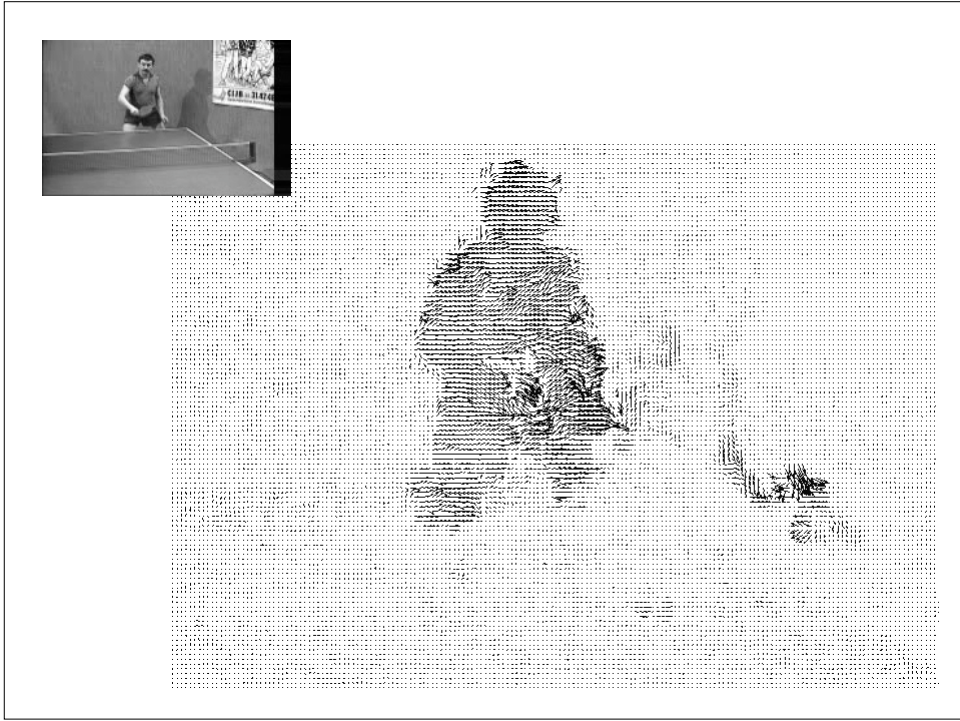
$$\underline{\bar{\alpha}}_x = x - \underline{x} \quad \underline{\bar{\alpha}}_x = x - \underline{x} = 3.2 - 3 = .2$$

$$\underline{\bar{\alpha}}_y = y - \underline{y} \quad \underline{\bar{\alpha}}_y = y - \underline{y} = 5.6 - 5 = .6$$









## Global Flow

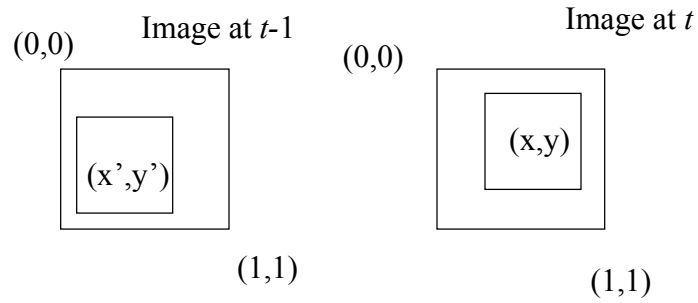
# Anandan

Affine

## Global Motion

- Estimate motion using all pixels in the image.
- Parametric flow gives an equation, which describes optical flow for each pixel.
  - Affine
  - Projective
- Global motion can be used to
  - generate mosaics
  - Object-based segmentation

## Affine



$$u(x, y) = a_1x + a_2y + b_1$$

$$v(x, y) = a_3x + a_4y + b_2$$

$$X_t = X_{t-1} U$$

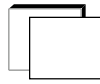
## Affine

$$u(x, y) = a_1x + a_2y + b_1$$

$$v(x, y) = a_3x + a_4y + b_2$$

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} a_1 & a_2 \\ a_3 & a_4 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

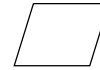
# Spatial Transformations



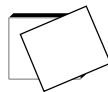
translation



rotation



shear



Rigid (rotation and translation)



affine

## Anandan

$$u(x, y) = a_1x + a_2y + b_1$$

$$v(x, y) = a_3x + a_4y + b_2$$

•Affine

$$\begin{bmatrix} u(x, y) \\ v(x, y) \end{bmatrix} = \begin{bmatrix} x & y & 1 & 0 & 0 \\ 0 & 0 & 0 & x & y & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ b_1 \\ a_3 \\ a_4 \\ b_2 \end{bmatrix}$$

**$\mathbf{u}(\mathbf{x}) = \mathbf{X}(\mathbf{x})\mathbf{a}$**

## Anandan

$$\mathbf{u}(\mathbf{x}) = \mathbf{X}(\mathbf{x})\mathbf{a}$$

Optical flow constraint eq  $f_x u + f_y v = -f_t$

$$E(\mathbf{a}) = \sum_{\mathbf{x} \in f(x,y)} (f_t + f_x^T \mathbf{a})^2$$

$$f_x = \begin{bmatrix} f_x \\ f_y \end{bmatrix}$$

$$E(\mathbf{a}) = \sum_{\mathbf{x} \in f(x,y)} (f_t + f_x^T \mathbf{X}(\mathbf{x})\mathbf{a})^2$$

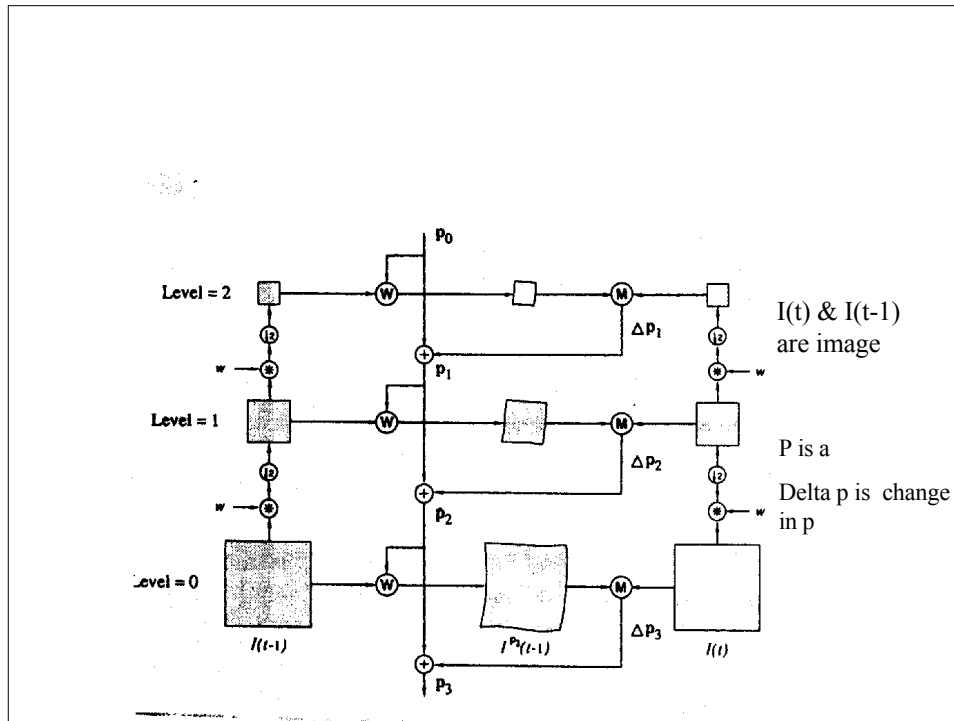
min 

$$\left[ \sum_{\mathbf{x}} \mathbf{X}^T(\mathbf{x}) (f_x)(f_x)^T \mathbf{X}(\mathbf{x}) \right] \mathbf{a} = \sum_{\mathbf{x}} \mathbf{X}^T(\mathbf{x}) f_x f_t$$
$$A\mathbf{x} = b$$

Linear system

## Basic Components

- Pyramid construction
- Motion estimation
- Image warping
- Coarse-to-fine refinement



## Image Warping

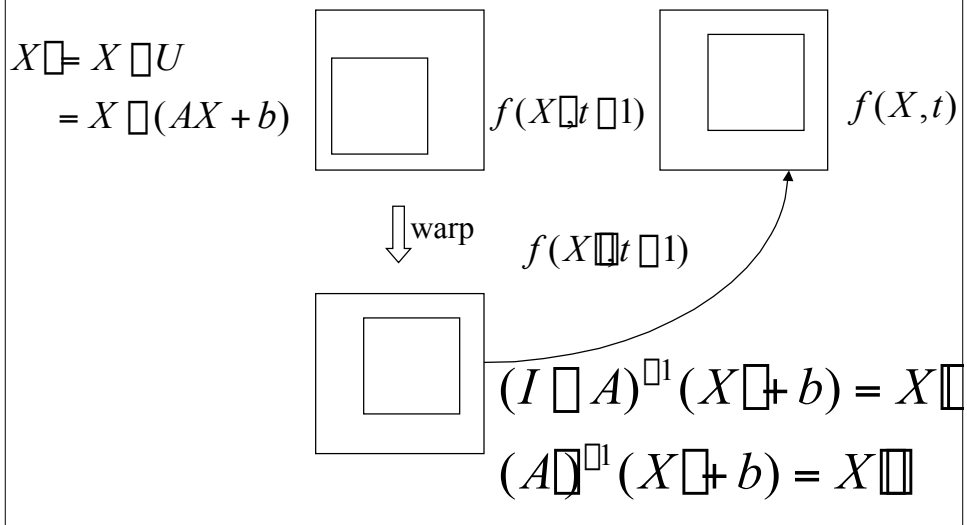
- Warping an image  $f$  into image  $h$  using some transformation  $g$ , involves mapping intensity at each pixel  $(x,y)$  in image  $f$  to a pixel  $(g(x),g(y))$  image  $h$  such that

$$(x \square y \square) = (g(x), g(y))$$

- In case of affine transformation,  $x = (x, y)$  is transformed to  $x \square = (x \square y \square)$  as:

$$x \square = Ax + b$$

## Image Warping



## Image Warping

$X_t = X_{t-1}U = X_{t-1}(AX + b)$       **Image at time t: X**  
 $X_t = (I - A)^{-1}X_{t-1} + b$       **Image at time t-1: X'**  
 $X_t = A^{-1}X_{t-1} + b$   
 $X_{t-1} + b = A^{-1}X_t$   
 $(A)^{-1}(X_{t-1} + b) = X_{t-1}$

$\downarrow$

$(A)^{-1}(X_{t-1} + b) = X_{t-1}$        $X_{t-1} \quad X_t$



## Image Warping

- How about values in  $(x, y)$  are not integer.
- But image is sampled only at integer rows and columns
  - Instead of converting  $(x, y)$  to  $(i, j)$  and copying at  $(i, j)$  we can convert integer values  $(i, j)$  to  $(x, y)$  and copy at  $(x, y)$

## Image Warping

- But how about the values in  $(x, y)$  are not integer.
- Perform bilinear interpolation to compute at non-integer values.

## Image Warping

$$(A^{-1})^T (X' + b) = X$$

$$(X' + b) = (A^{-1})^T X$$

$$X' = (A^{-1})^T X - b \quad X' \approx X$$

## Warping

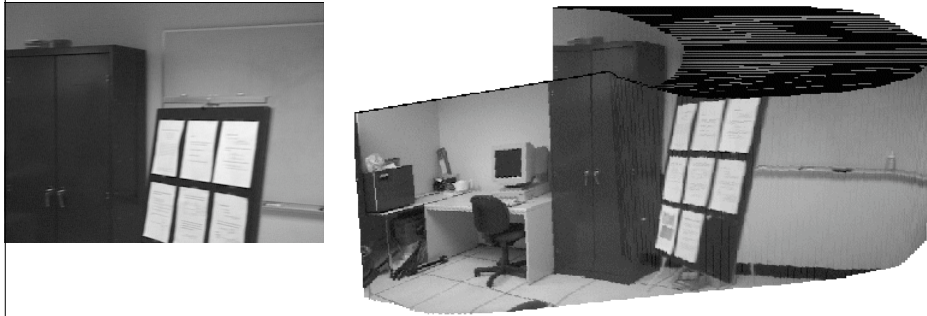


## Show Demos

## Video Mosaic



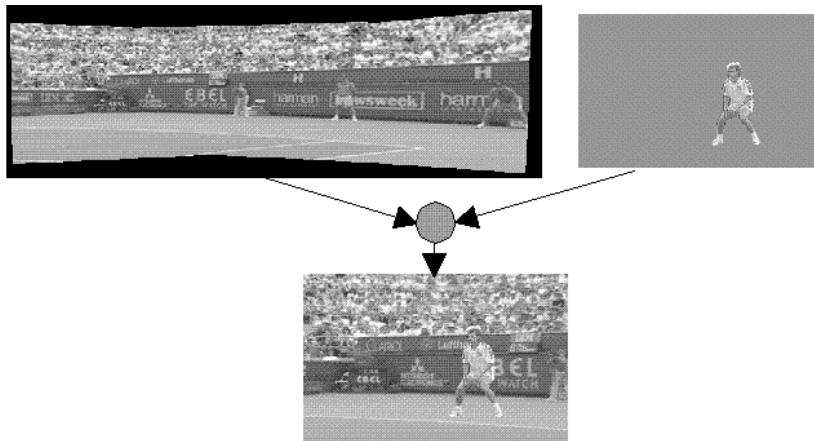
## Video Mosaic



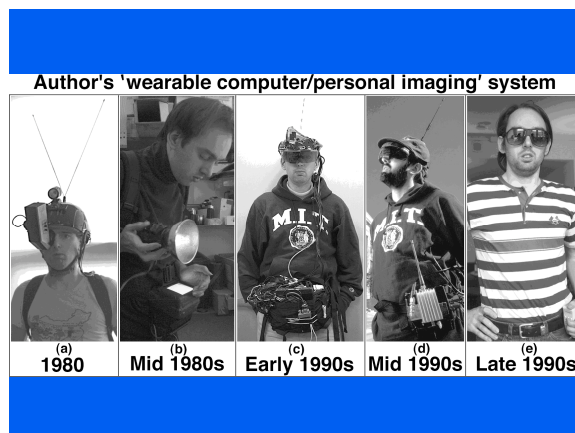
## Video Mosaic



# Sprite



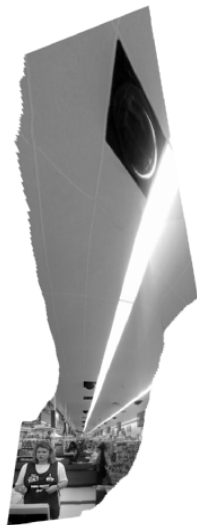
# Steve Mann



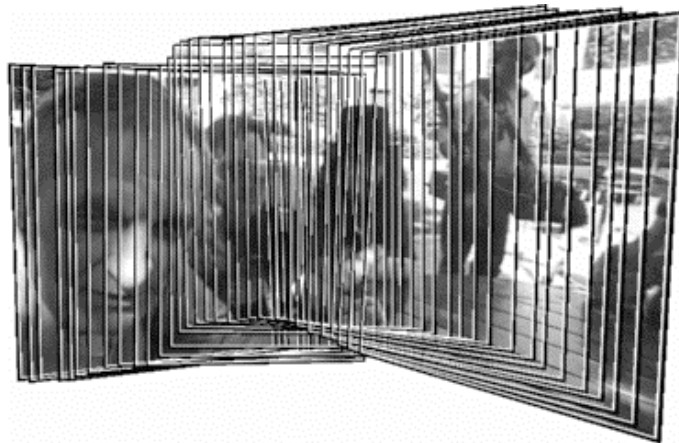
# Building



# Wal-Mart



## Scientific American Frontiers



## Scientific American Frontiers



## Head-mounted Camera at Restaurant



## MIT Media Lab





## Webpages

- <http://n1nlf1.eecg.toronto.edu/tip.ps.gz>  
Video Orbits of the projective group, S. Mann and R. Picard.
- <http://wearcam.org/pencigraphy>  
(C code for generating mosaics)