

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

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

⋮

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

$$\begin{bmatrix} f_{x1} & f_{y1} \\ \vdots & \vdots \\ f_{x9} & f_{y9} \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} -f_{t1} \\ \vdots \\ -f_{t9} \end{bmatrix}$$

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

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

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



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

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

Lucas & Kanade

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

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

$$u = \frac{-\sum f_{yi}^2 \sum f_{xi}f_{ti} + \sum f_{xi}f_{yi} \sum f_{yi}f_{ti}}{\sum f_{xi}^2 \sum f_{yi}^2 - \sum f_{xi}f_{yi} \sum f_{yi}^2}$$

$$v = \frac{\sum f_{xi}f_{ti} \sum f_{xi}f_{ti} - \sum f_{xi}^2 \sum f_{yi}f_{ti}}{\sum f_{xi}^2 \sum f_{yi}^2 - \sum f_{xi}f_{yi} \sum f_{yi}^2}$$

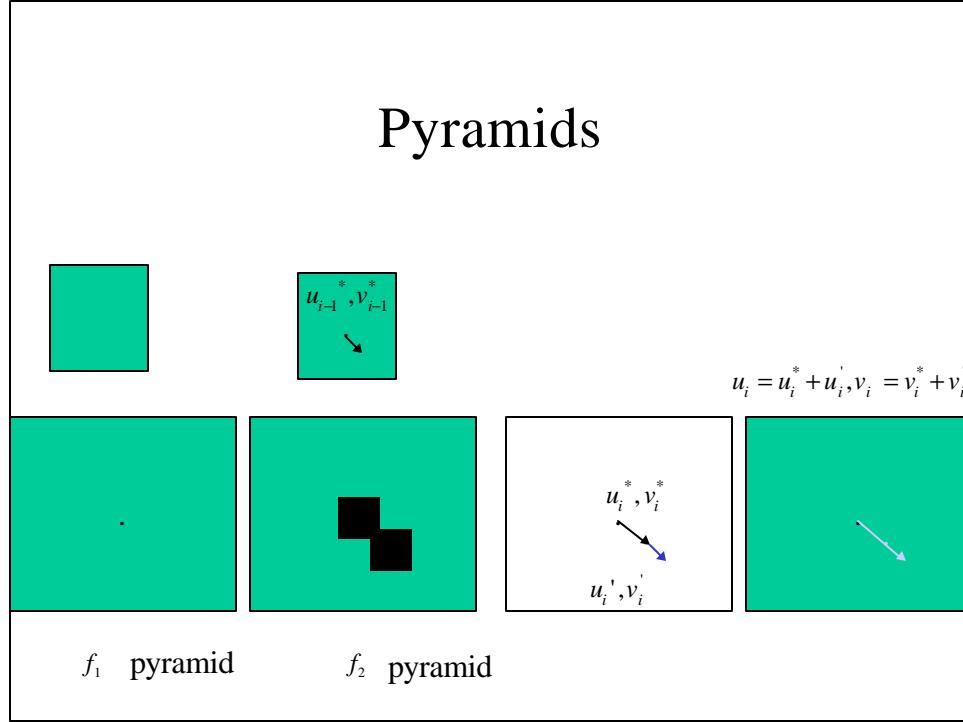
Comments

- 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
 - 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



Interpolation

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

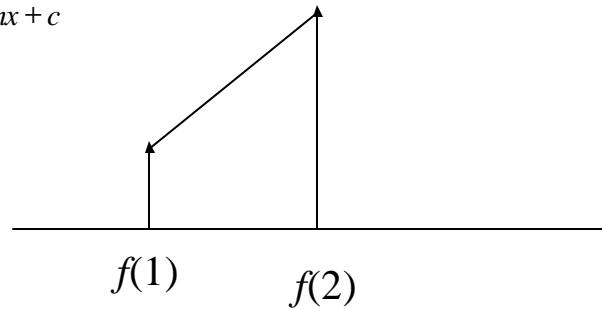
	0	1	2	3	4	5	6	7
0	•	◦	•	◦	•	◦	•	◦
1	◦	◦	◦	◦	◦	◦	◦	◦
$u^* = 3$	◦	◦	◦	◦	◦	◦	◦	◦
4	•	◦	•	◦	•	◦	•	◦
5	◦	◦	◦	◦	◦	◦	◦	◦
6	•	◦	•	◦	•	◦	•	◦
7	◦	◦	◦	◦	◦	◦	◦	◦

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

1-D Interpolation

$$y = mx + c$$

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



2-D Interpolation

$$f(x,y) = a_1 + a_2x + a_3y + a_4xy \quad \text{Bilinear}$$

X X
O X

Bi-linear Interpolation

Four nearest points of (x,y) are:

$$(\underline{x}, \underline{y}), (\bar{x}, \underline{y}), (\underline{x}, \bar{y}), (\bar{x}, \bar{y})
(3,5), (4,5), (3,6), (4,6)$$

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

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

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

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

$$f'(x, y) = \overline{\underline{\mathbf{e}_x}} \overline{\underline{\mathbf{e}_y}} f(\underline{x}, \underline{y}) + \underline{\underline{\mathbf{e}_x}} \overline{\underline{\mathbf{e}_y}} f(\overline{x}, \underline{y}) +$$

$$\overline{\underline{\mathbf{e}_x}} \underline{\underline{\mathbf{e}_y}} f(\underline{x}, \overline{y}) + \underline{\underline{\mathbf{e}_x}} \underline{\underline{\mathbf{e}_y}} f(\overline{x}, \overline{y})$$

$$\overline{\underline{\mathbf{e}_x}} = \overline{x} - x$$

$$\overline{\underline{\mathbf{e}_x}} = \overline{x} - x = 4 - 3.2 = .8$$

$$\overline{\underline{\mathbf{e}_y}} = \overline{y} - y$$

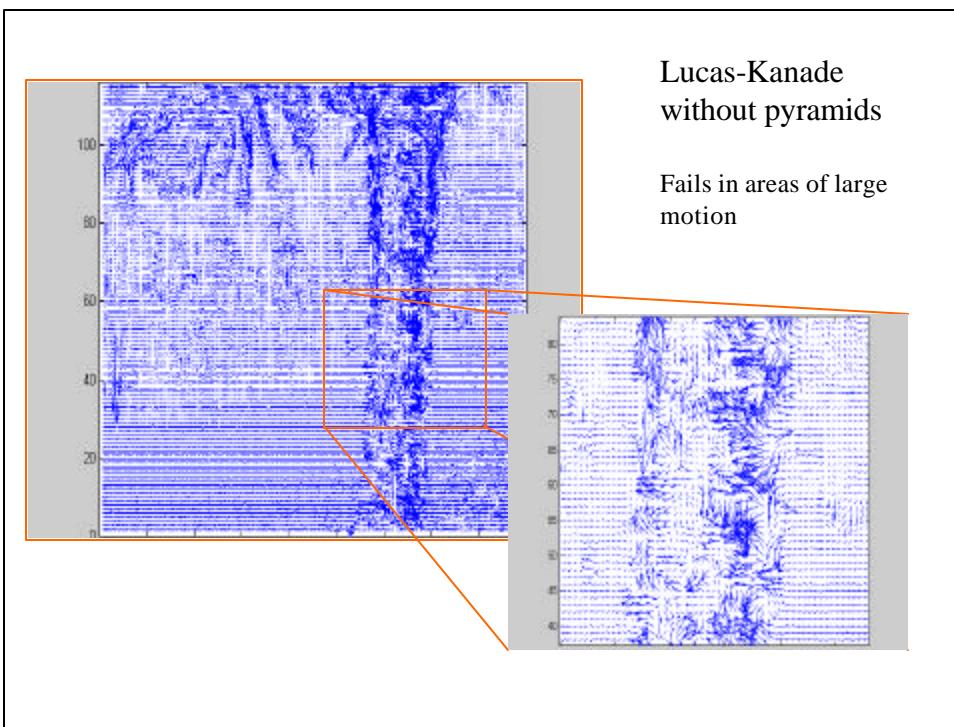
$$\overline{\underline{\mathbf{e}_y}} = \overline{y} - y = 6 - 5.6 = .4$$

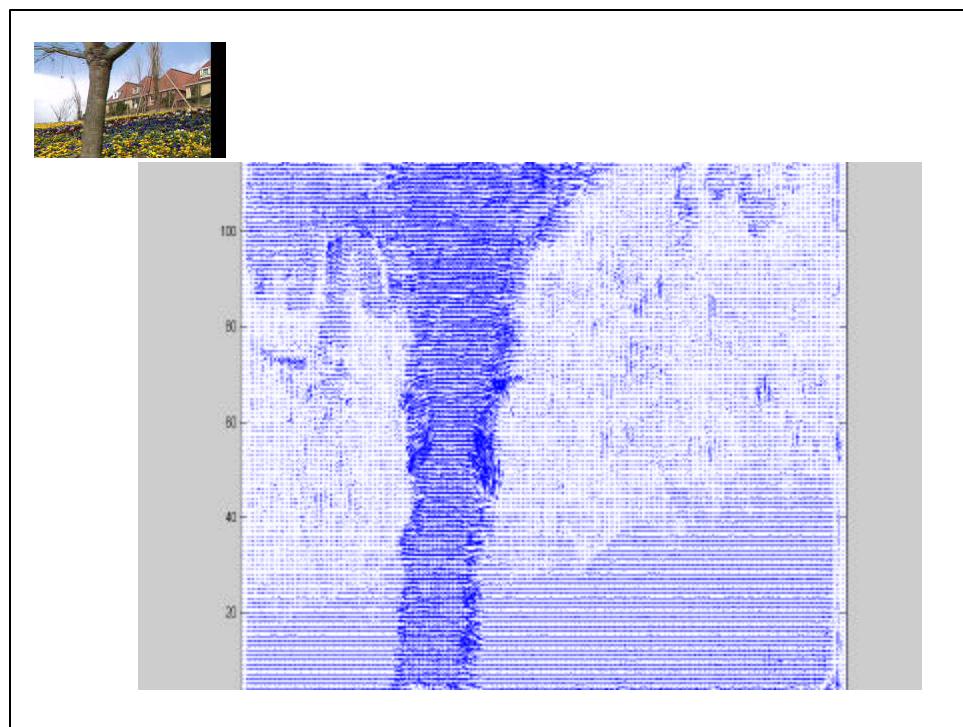
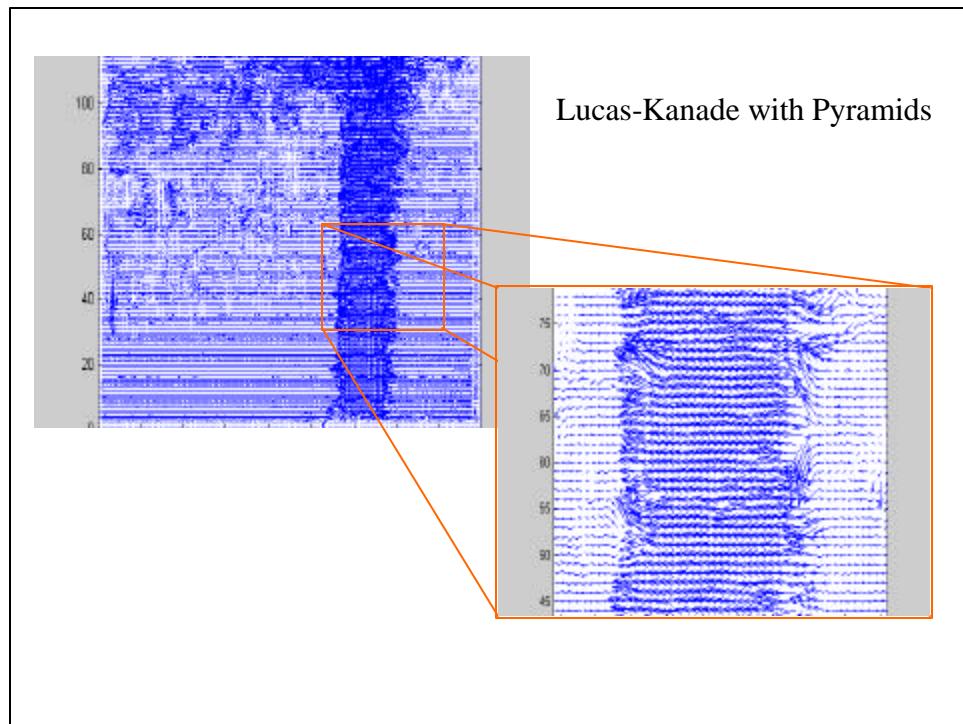
$$\underline{\underline{\mathbf{e}_x}} = x - \underline{x}$$

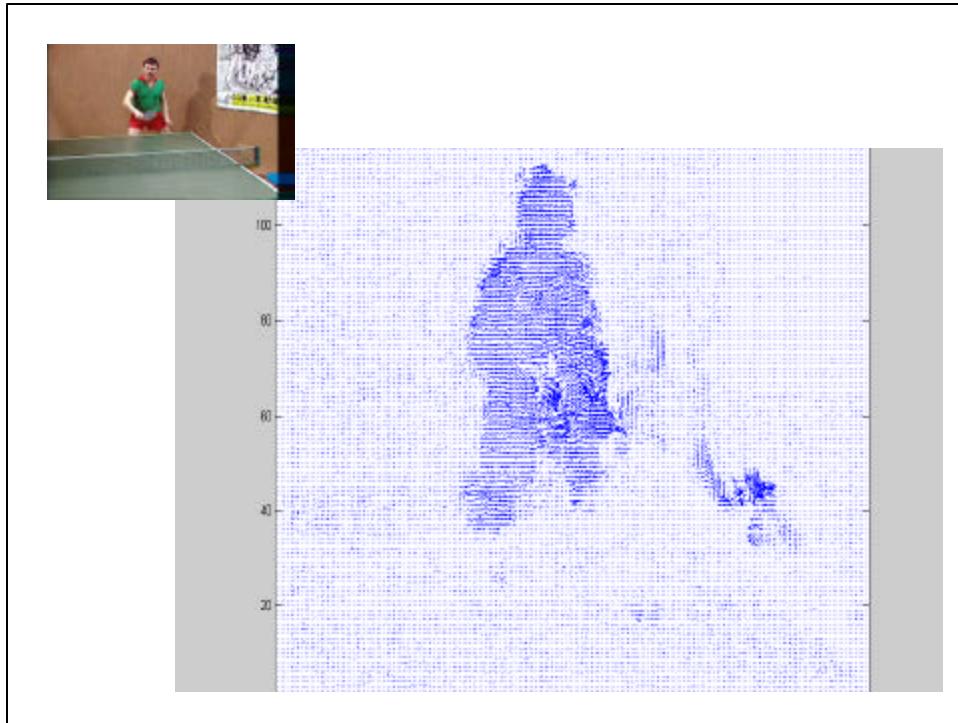
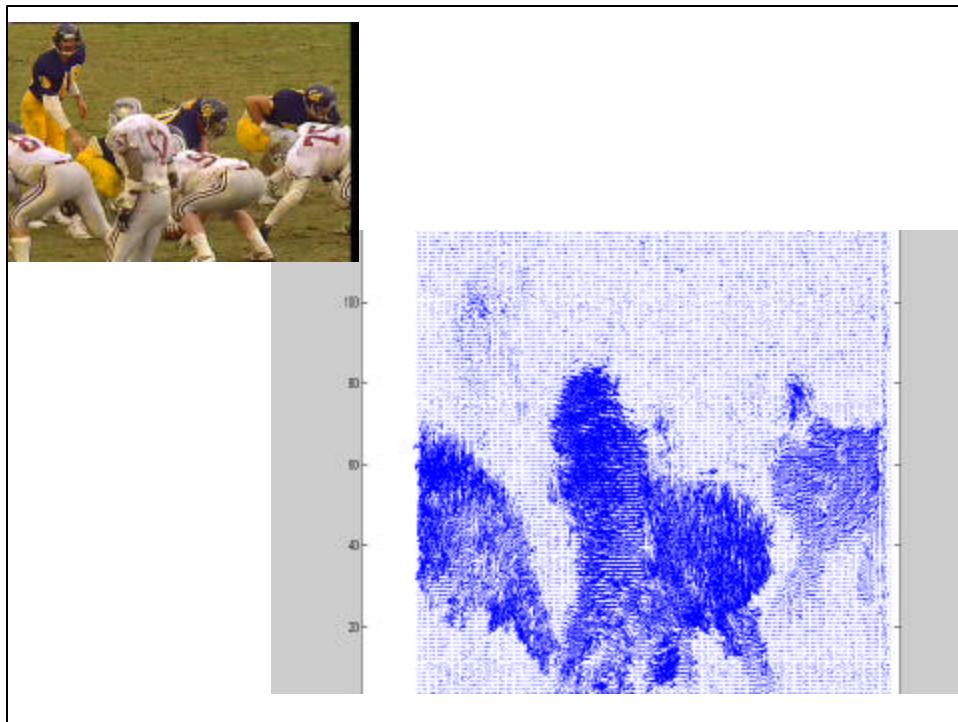
$$\underline{\underline{\mathbf{e}_x}} = x - \underline{x} = 3.2 - 2 = .2$$

$$\underline{\underline{\mathbf{e}_y}} = y - \underline{y}$$

$$\underline{\underline{\mathbf{e}_y}} = y - \underline{y} = 5.6 - 5 = .6$$







Global Flow

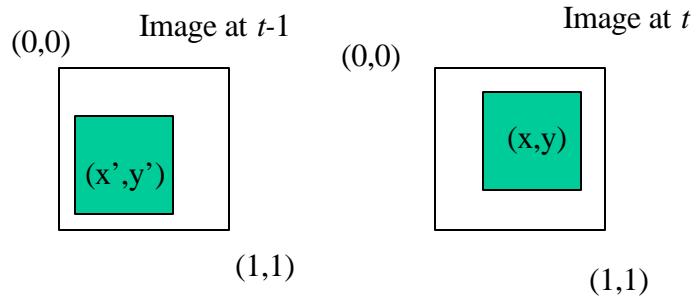
Anandan

Affine

Global Motion

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

Affine



$$u(x, y) = a_1 x + a_2 y + b_1$$

$$v(x, y) = a_3 x + a_4 y + b_2$$

$$X' = X - U$$

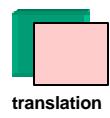
Affine

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

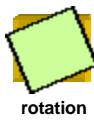
$$v(x, y) = a_3x + a_4y + b_2$$

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

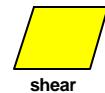
Spatial Transformations



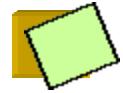
translation



rotation



shear



Rigid (rotation and translation)



affine

Anandan

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

$$v(x, y) = a_3x + a_4y + b_2 \quad \bullet \text{Affine}$$

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

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

Anandan

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

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

$$E(\mathbf{da}) = \sum_{\forall x \in f(x, y)} (f_t + f_x^T \mathbf{du})^2 \quad f_x = \begin{bmatrix} f_x \\ f_y \end{bmatrix}$$

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

$$\min \quad \downarrow$$

$$\left[\sum X^T (f_x)(f_x)^T X \right] \ddot{\mathbf{a}} = - \sum X^T f_x f_t$$

$$Ax = b$$

Linear system

Basic Components

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

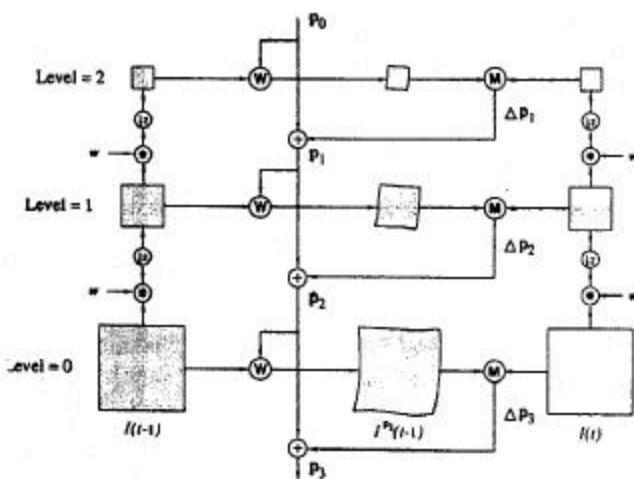


Image Warping

- 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 $(g(x), g(y))$ in image h such that

$$(x', y') = (g(x), g(y))$$

- In case of affine transformation, $\mathbf{x} = (x, y)$ is transformed to $\mathbf{x}' = (x', y')$ as:

$$\mathbf{x}' = \mathbf{Ax} + \mathbf{b}$$

Image Warping

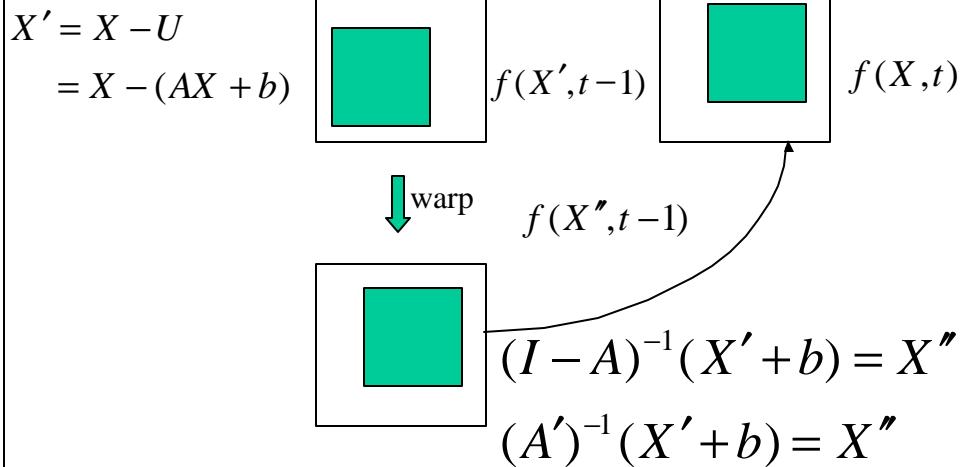


Image Warping

$$\begin{aligned} X' &= X - U = X - (AX + b) && \text{Image at time t: } \mathbf{X} \\ X' &= (I - A)X - b && \text{Image at time t-1: } \mathbf{X}' \\ X' &= A'X - b \\ X' + b &= A'X \\ (A')^{-1}(X' + b) &= X \\ &\downarrow \\ (A')^{-1}(X' + b) &= X'' && X' \rightarrow X'' \end{aligned}$$

Image Warping

- How about values in X' are not integer.
- But image is sampled only at integer rows and columns
 - Instead of converting X' to X'' and copying at integer values to X'' we can convert X' to X'' and copy at

Image Warping

- But how about the values in θ are not integer.
- Perform bilinear interpolation to compute at non-integer values.

Image Warping

$$(A')^{-1}(X' + b) = X''$$

$$(X' + b) = (A')X''$$

$$X' = (A')X'' - b \quad X'' \rightarrow X'$$

Warping



Show Demos

Video Mosaic



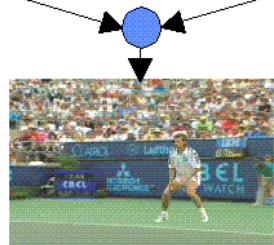
Video Mosaic



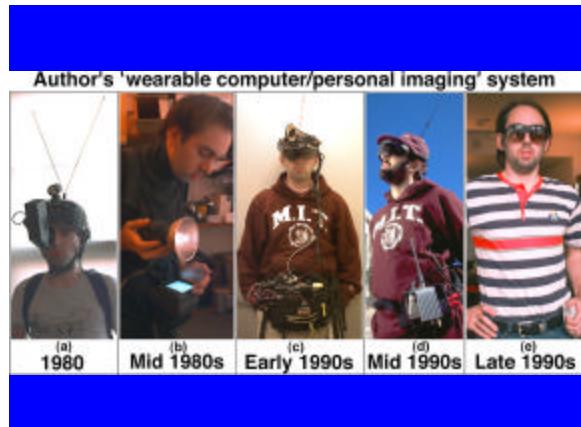
Video Mosaic



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)