

# Motion Models

- Image Transformations to relate two images
- 3D Rigid motion
- Perspective & Orthographic Transformation
- Planar Scene Assumption
- Transformations
  - Translation
  - Rotation
  - Rigid
  - Affine
  - Homography
  - Pseudo Perspective

# Global Flow

Lecture-9

# 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
  - Remove camera (ego) motion (motion compensation)
  - Object-based segmentation
  - generate mosaics

# Contents

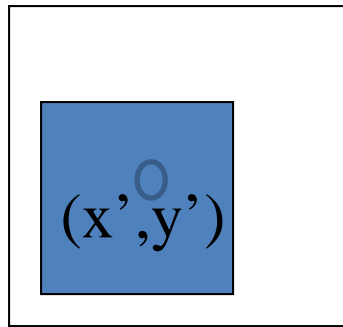
- Bergen *et al* method
  - Affine transformation
- Mann & Piccard
  - Homography (Projective)
  - Pseudo Perspective
  - Bi-linear
- Image Warping
- Applications
  - Mosaics
  - COCOA system

Bergan et al

Affine

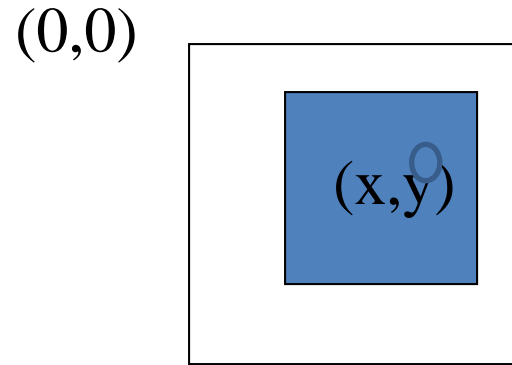
# Affine

(0,0) Image at  $t-1$



(1,1)

Image at  $t$



(1,1)

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

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

$$U = X - X'$$

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix} - \begin{bmatrix} x' \\ y' \end{bmatrix}$$

# Affine

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

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

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

# Bergan et al

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

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

•Affine

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

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



# Bergan et al

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

Optical flow constraint eq

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

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

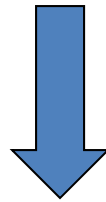
$$E(\mathbf{a}) = \sum_{\forall x \in f(x,y)} (f_t + f_x^T \mathbf{X}(\mathbf{x})\mathbf{a})^2$$

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

$$f_{\mathbf{X}} = \begin{bmatrix} f_x \\ f_y \end{bmatrix}$$

# Bergan et al

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



min

$$\left[ \sum X^T (f_X (f_X)^T X) \right] \delta a = - \sum X^T f_X f_t \quad \text{Homework}$$

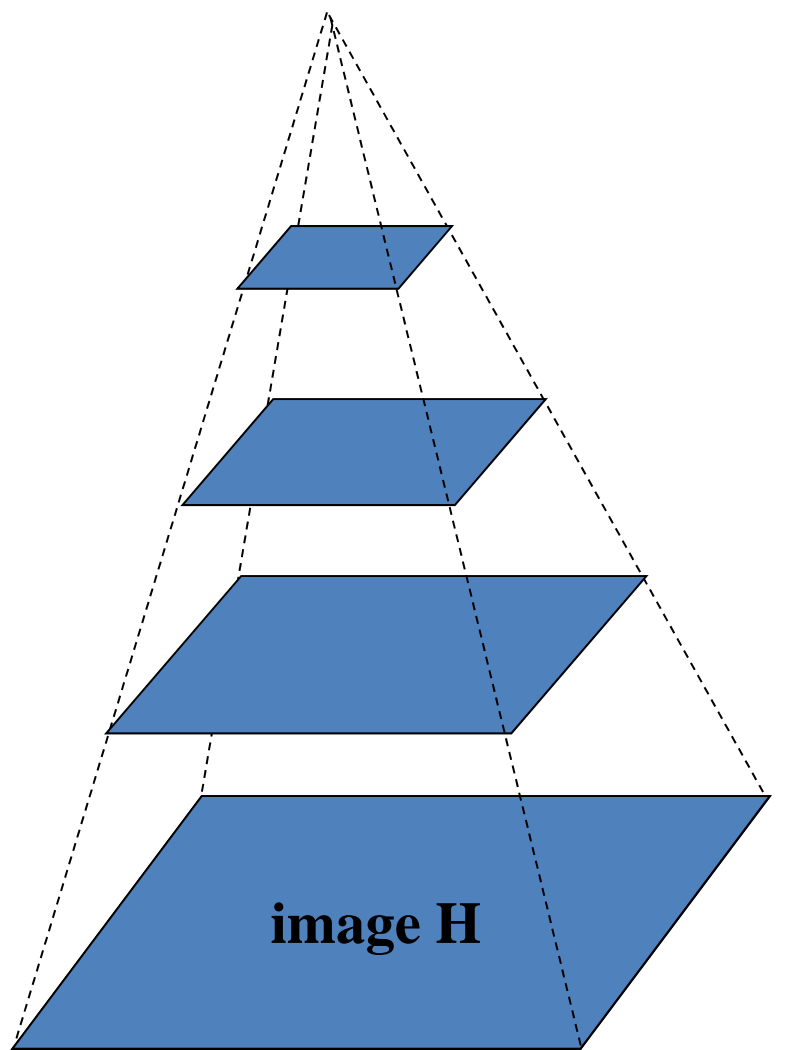
$$Aa = B$$

Linear system

# Basic Components

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

# Coarse-to-fine global flow estimation



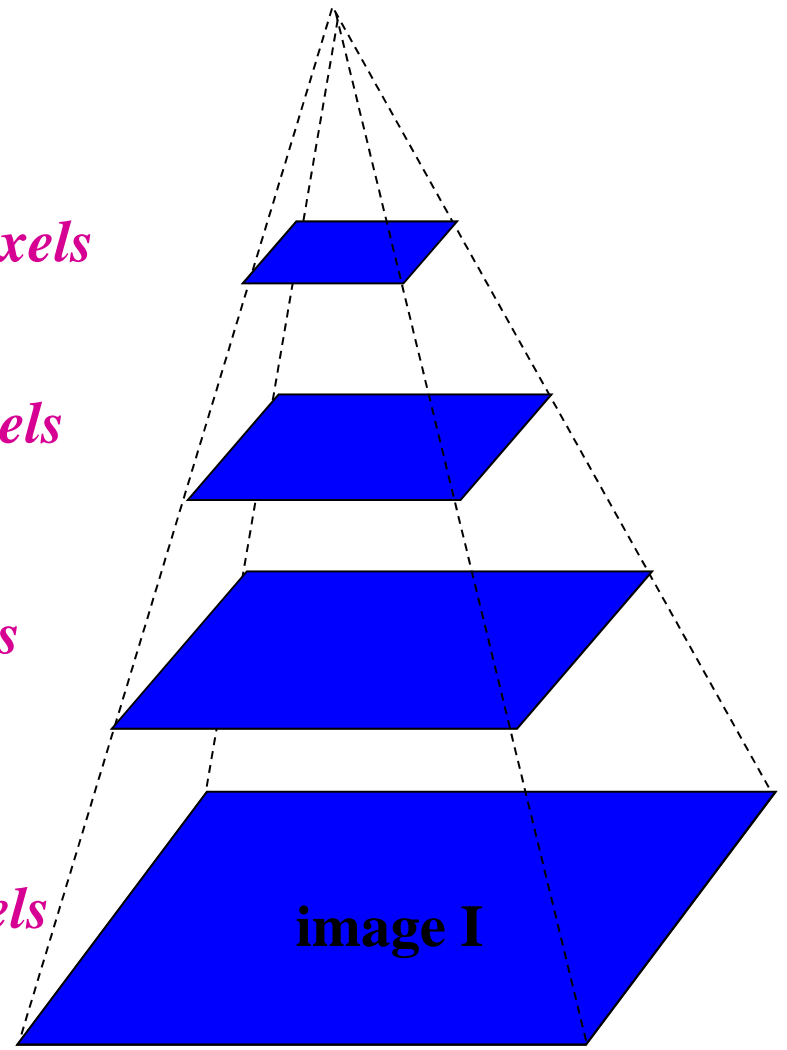
**Gaussian pyramid of image H**

*$u=1.25$  pixels*

*$u=2.5$  pixels*

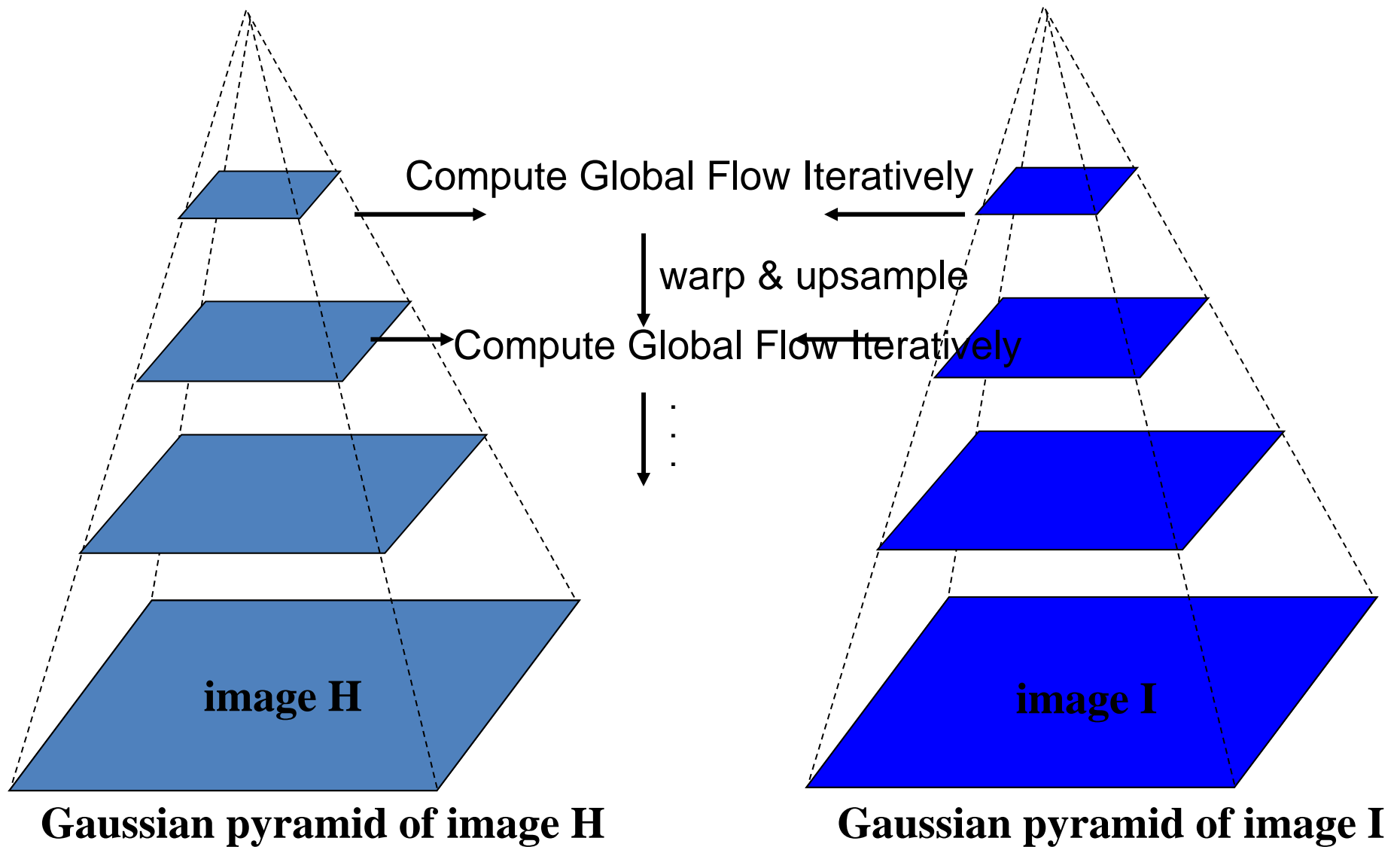
*$u=5$  pixels*

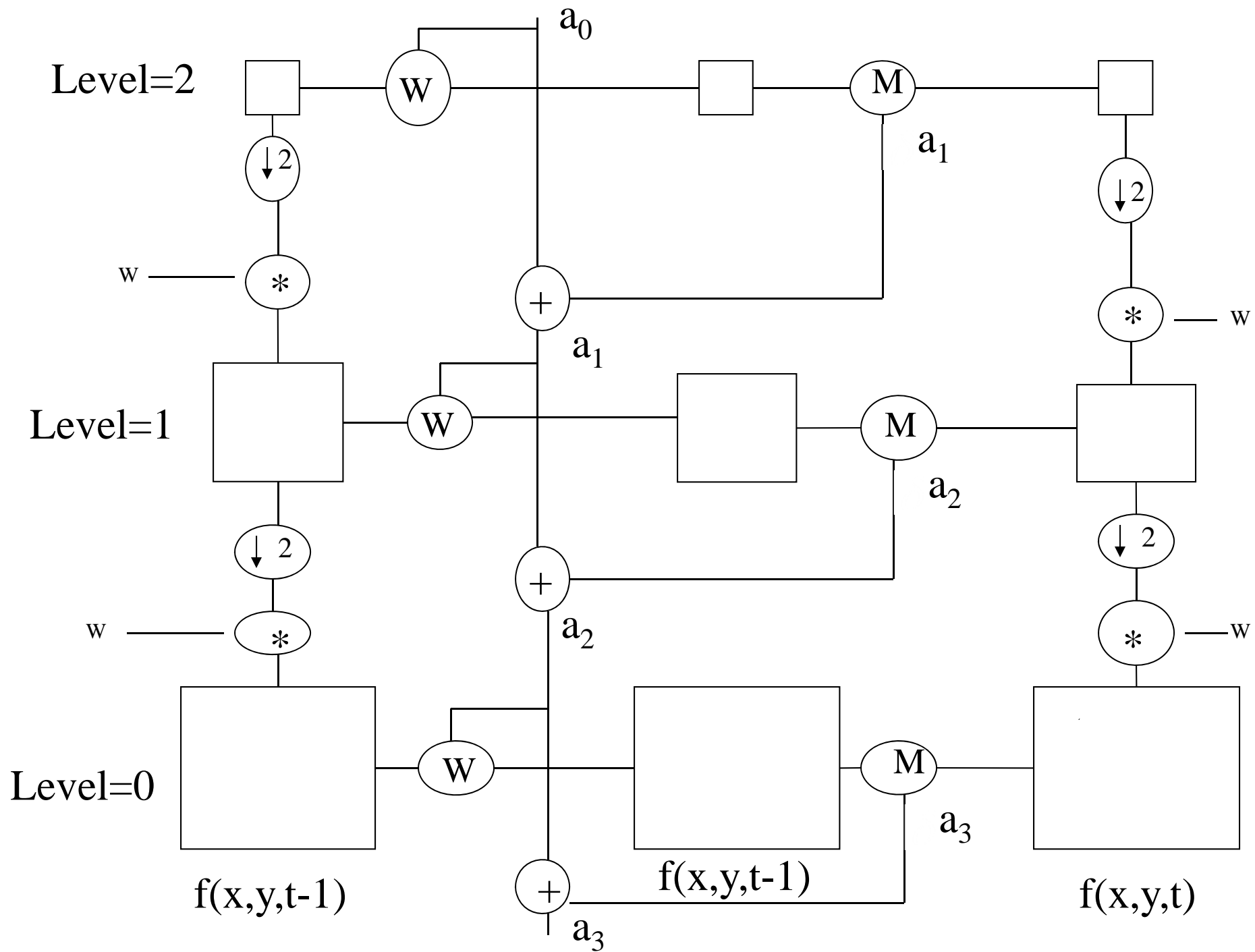
*$u=10$  pixels*



**Gaussian pyramid of image I**

# Coarse-to-fine global flow estimation





# Estimation of Global Flow

## Single Iteration



Image 't'

Compute **A** and **B**

Solve  $Aa = B$



Image 't+1'

Warp by **a**



# Estimation of Global Flow

Iterative

Initial Estimate  $\mathbf{a} = [a_1 \ a_2 \ b_1 \ a_3 \ a_4 \ b_2]^T$



Image 't'

Warp by  $\mathbf{a}$



Image 't+1'

Compute  $\mathbf{A}$  and  $\mathbf{B}$

Solve  $\mathbf{A}\delta\mathbf{a} = \mathbf{B}$

Warp by  $\delta\mathbf{a}$





# Estimation of Global Flow

Iterative

Initial Estimate  $\mathbf{a} = [a_1 \ a_2 \ b_1 \ a_3 \ a_4 \ b_2]^T$



Image 't'

Warp by  $\mathbf{a}$



Iterate



Image 't+1'

Compute  $\mathbf{A}$  and  $\mathbf{B}$

Solve  $\mathbf{A}\delta\mathbf{a} = \mathbf{B}$

Update  $\mathbf{a}$



# 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

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

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

$$\mathbf{x}' = \mathbf{A}\mathbf{x} + \mathbf{b} \quad \text{Displacement model}$$

$$\mathbf{U} = \mathbf{x} - \mathbf{x}' = \mathbf{A}\mathbf{x} + \mathbf{b} \quad \text{Instantaneous model}$$

# Image Warping (Bergan et al)

$$X' = X - U = X - (AX + b)$$

$$X' = (I - A)X - b$$

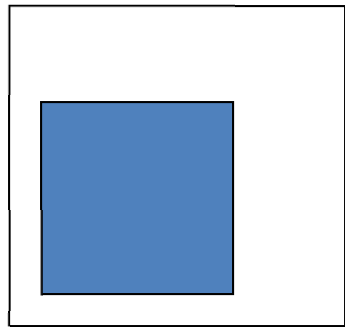
$$X' = A'X - b$$

$$X' + b = A'X$$

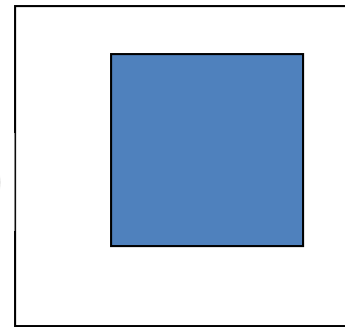
$$(A')^{-1}(X' + b) = X$$

↓ warp

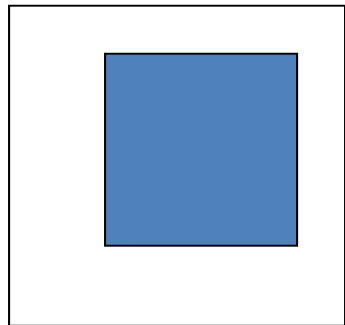
$$(A')^{-1}(X' + b) = X''$$



$f(X', t-1)$

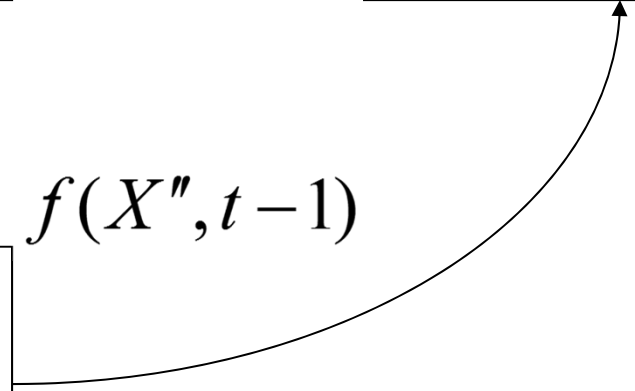


$f(X, t)$



$f(X'', t-1)$

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



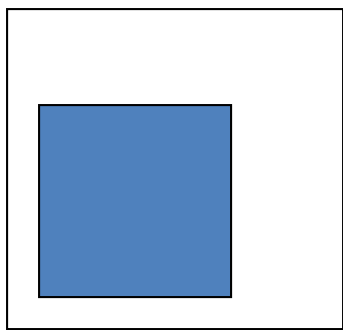
# Image Warping

- How about values in  $X'' = (x'', y'')$  are not integer.
- But image is sampled only at integer rows and columns
  - Instead of converting  $X'$  to  $X''$  and copying  $f(X', t-1)$  at  $f(X'', t-1)$  we can convert integer values  $X''$  to  $X'$  and copy  $f(X', t-1)$  at  $f(X'', t-1)$

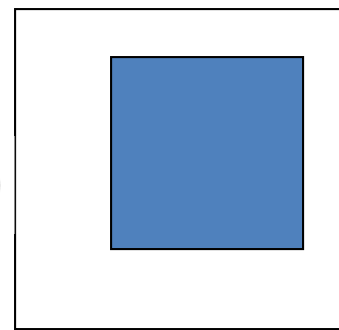
# Image Warping

$$X' = X - U$$

$$X' = X - (AX + b)$$



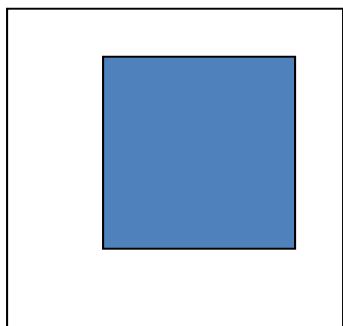
$f(X', t-1)$



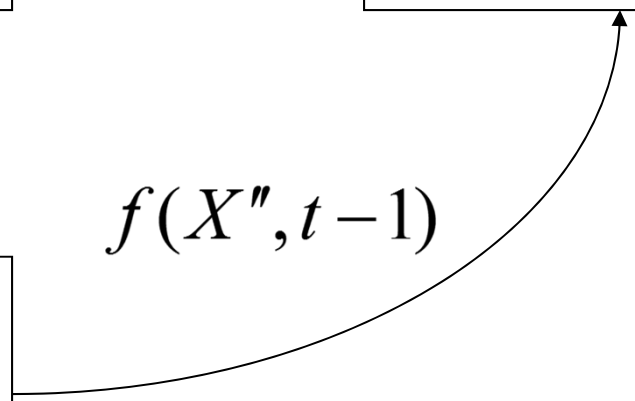
$f(X, t)$

$$X' = X'' - (AX'' + b)$$

↓ warp



$f(X'', t-1)$



# Image Warping

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

# Warping

$f(X', t-1)$



$f(X', t-1)$



Warped image at  $t-1$



Difference image before



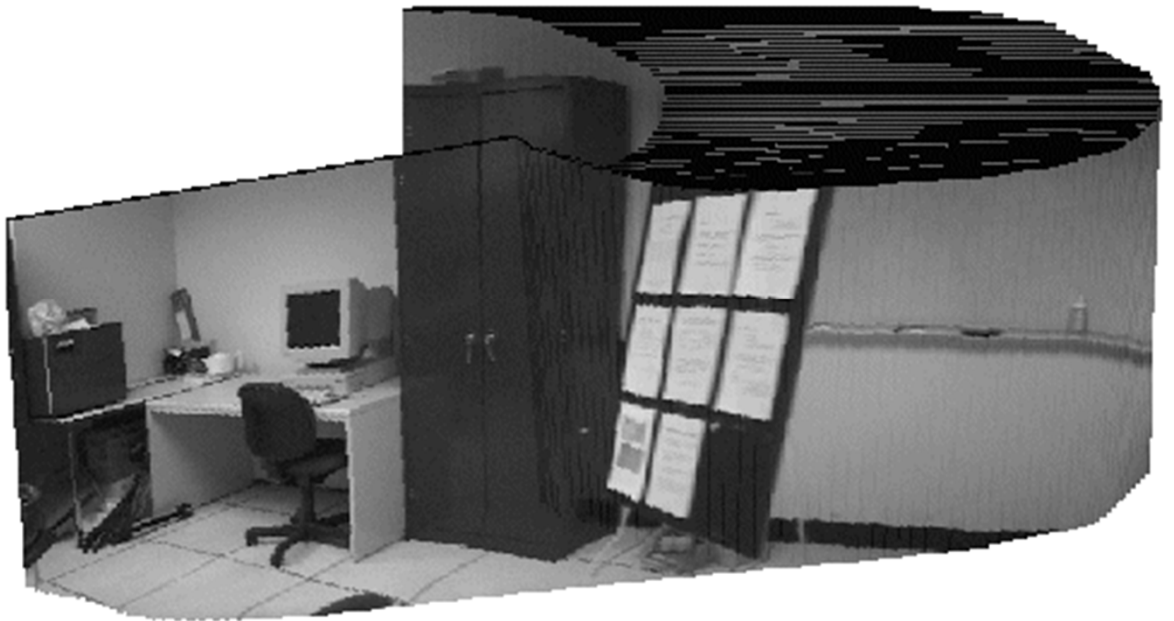
Difference image after

# Video Mosaic





# Video Mosaic



# Video Mosaic



# Sprite



# Mann & Picard

Projective

# Projective Flow (weighted)

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

Optical Flow const.  
equation

$$\mathbf{u}^T \mathbf{f}_x + f_t = 0$$

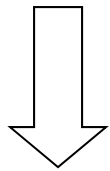
$$\mathbf{x}' = \frac{A \mathbf{x} + \mathbf{b}}{\mathbf{C}^T \mathbf{x} + 1}$$

Projective transform

$$\mathbf{u} = \mathbf{x}' - \mathbf{x} = \frac{A \mathbf{x} + \mathbf{b}}{\mathbf{C}^T \mathbf{x} + 1} - \mathbf{x}$$

# Projective Flow (weighted)

$$\begin{aligned}\mathcal{E}_{flow} &= \sum (\mathbf{u}^T \mathbf{f}_x + f_t)^2 \\ &= \sum \left( \left( \frac{A\mathbf{x} + \mathbf{b}}{\mathbf{C}\mathbf{x}^T + 1} - \mathbf{x} \right)^T \mathbf{f}_x + f_t \right)^2 \\ &= \sum \left( (A\mathbf{x} + \mathbf{b} - (\mathbf{C}\mathbf{x}^T + 1)\mathbf{x})^T \mathbf{f}_x + (\mathbf{C}\mathbf{x}^T + 1)f_t \right)^2\end{aligned}$$



**minimize**

Homework

# Projective Flow (weighted)

$$\left(\sum \phi \phi^T\right) \mathbf{a} = \sum \left(\mathbf{x}^T \mathbf{f}_x - f_t\right) \phi$$

$$\mathbf{a} = \left[ a_1, a_2, b_1, a_3, a_4, b_2, c_1, c_2 \right]^T$$

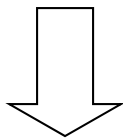
$$\phi^t = [f_x x, f_x y, f_x, f_y x, f_y y, f_y, x f_t - x^2 f_x - x y f_y, y f_t - x y f_x - y^2 f_y]$$

Projective Flow (unweighted)



# Pseudo-Perspective

$$\mathbf{x}' = \frac{A \mathbf{x} + \mathbf{b}}{\mathbf{C}^T \mathbf{x} + 1}$$



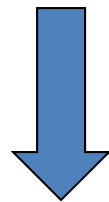
**Taylor Series**

$$x + u = a_1 + a_2 x + a_3 y + a_4 x^2 + a_5 xy$$

$$y + v = a_6 + a_7 x + a_8 y + a_4 xy + a_5 y^2$$

# Bilinear

$$\mathbf{x}' = \frac{A \mathbf{x} + \mathbf{b}}{\mathbf{C}^T \mathbf{x} + 1}$$



Taylor Series & remove  
Square terms

$$u + x = a_1 + a_2 x + a_3 y + a_4 xy$$

$$v + y = a_5 + a_6 x + a_7 y + a_8 xy$$

# Projective Flow (unweighted)

$$\mathcal{E}_{flow} = \sum (\mathbf{u}^T \mathbf{f}_x + f_t)^2$$

**Minimize**

# Bilinear and Pseudo-Perspective

$$(\sum \Phi \Phi^T) \mathbf{q} = -\sum f_t \Phi$$

$$\Phi^T = [f_x(xy, x, y, 1), f_y(xy, x, y, 1)] \quad \mathbf{bilinear}$$

$$\Phi^T = [f_x(x, y, 1), f_y(x, y, 1), c_1, c_2]$$

$$c_1 = x^2 f_x + xy f_y$$

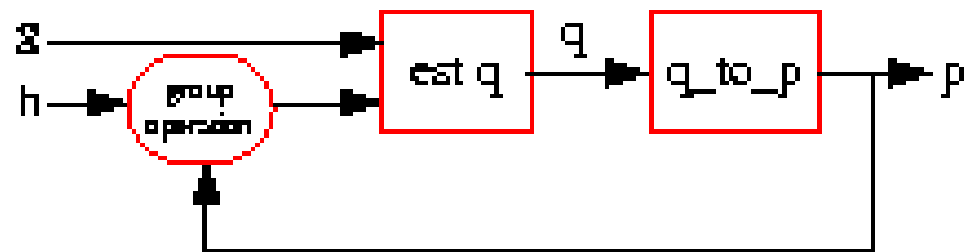
$$c_2 = xy f_x + y^2 f_y$$

**Pseudo perspective**

Homework

# Algorithm-1

- Estimate “q” (using approximate model, e.g. bilinear model).
- Relate “q” to “p”
  - select four points S1, S2, S3, S4
  - apply approximate model using “q” to compute
  - estimate exact “p”:



# Determining Projective transformation using point correspondences

- If point correspondences  $(x,y) \leftrightarrow (x',y')$  are known
- $a$ 's can be determined by least squares fit

$$x' = \frac{a_1x + a_2y + a_3}{a_7x + a_8y + 1}$$

$$y' = \frac{a_4x + a_5y + a_6}{a_7x + a_8y + 1}$$

$$a_7x'x + a_8x'y + x' = a_1x + a_2y + a_3$$

$$a_7y'x + a_8y'y + y' = a_4x + a_5y + a_6$$

$$x' = a_1x + a_2y + a_3 - a_7x'x - a_8x'y$$

$$y' = a_4x + a_5y + a_6 - a_7y'x - a_8y'y$$

$$a_1x + a_2y + a_3 - a_7x'x - a_8x'y = x'$$

$$a_4x + a_5y + a_6 - a_7y'x - a_8y'y = y'$$

Two rows for each point  $i$

$$\begin{bmatrix} x_i & y_i & 1 & 0 & 0 & \vdots & 0 & -x_i x'_i & -y_i x'_i \\ 0 & 0 & 0 & x_i & y_i & 1 & -x_i y'_i & -y_i y'_i \\ \vdots & & & & & & & & \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_6 \\ a_7 \\ a_8 \end{bmatrix} = \begin{bmatrix} \vdots \\ x'_i \\ y'_i \\ \vdots \end{bmatrix}$$

# Determining Projective transformation using point correspondences

$$\begin{bmatrix} x_i & y_i & 1 & 0 & 0 & \vdots & 0 & -x_i x'_i & -y_i x'_i \\ 0 & 0 & 0 & x_i & y_i & \vdots & 1 & -x_i y'_i & -y_i y'_i \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_6 \\ a_7 \\ a_8 \end{bmatrix} = \begin{bmatrix} \vdots \\ x'_i \\ y'_i \\ \vdots \end{bmatrix}$$

$$Aa = \mathbf{x}'$$

$$a = (A^T A)^{-1} A^T \mathbf{x}'$$



# Final Algorithm

- A Gaussian pyramid of three or four levels is constructed for each frame in the sequence.
- The parameters “p” are estimated at the top level of the pyramid, between the two lowest resolution images, “g” and “h”, using algorithm-1.

# Final Algorithm

- The estimated “p” is applied to the next higher resolution image in the pyramid, to make images at that level nearly congruent.
- The process continues down the pyramid until the highest resolution image in the pyramid is reached.

# Video Mosaics

- Mosaic aligns different pieces of a scene into a larger piece, and seamlessly blend them.
  - High resolution image from low resolution images
  - Increased field of view

# Steps in Generating A Mosaic

- Take pictures
- Pick reference image
- Determine transformation between frames
- Warp all images to the same reference view

# Applications of Mosaics

- Virtual Environments
- Computer Games
- Movie Special Effects
- Video Compression

# Steve Mann

Author's 'wearable computer/personal imaging' system



# Sequence of Images



# Projective Mosaic





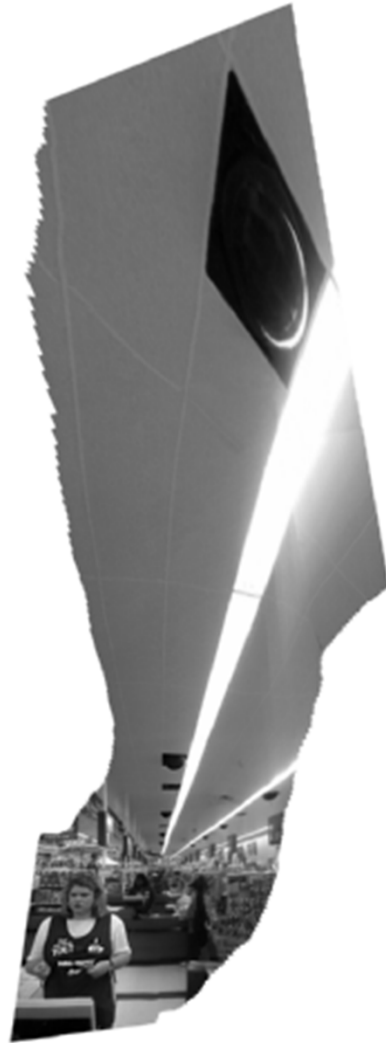
# Affine Mosaic



# Building



# Wal-Mart



# Scientific American Frontiers



# Scientific American Frontiers



# Head-mounted Camera at Restaurant



# MIT Media Lab



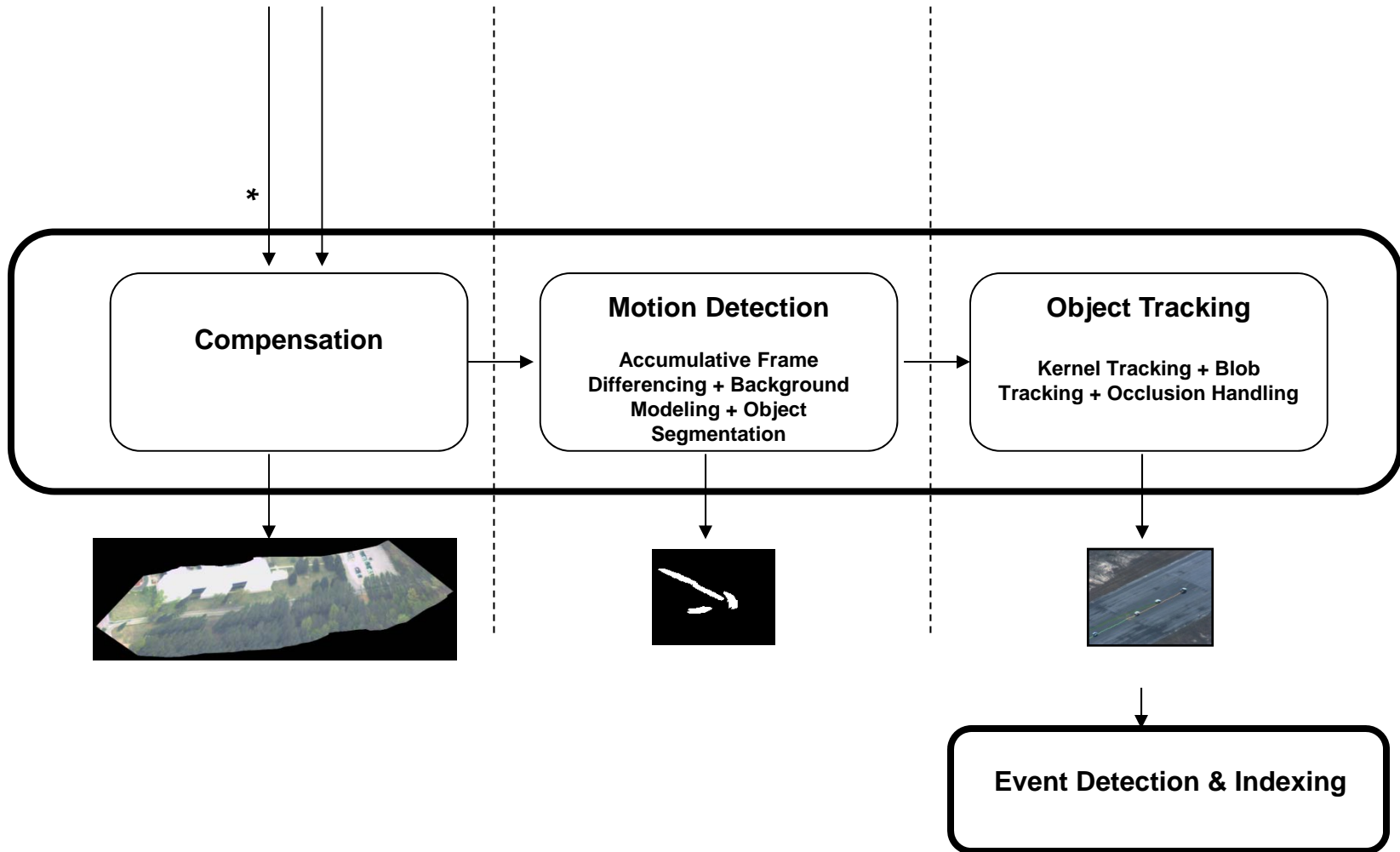
# COCOA: A System for Processing of Aerial Videos



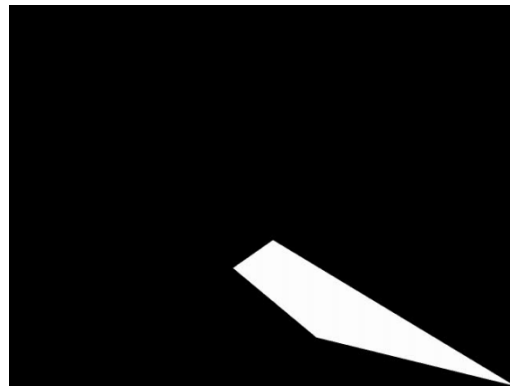
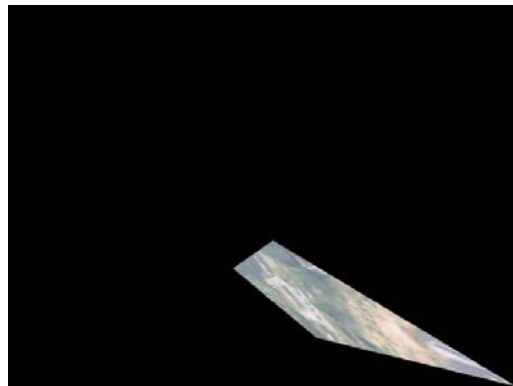
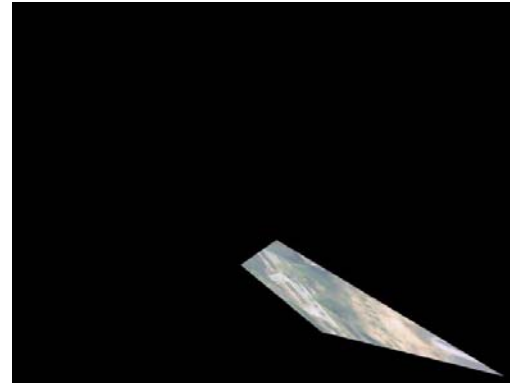
Saad Ali and Mubarak Shah, [COCOA - Tracking in Aerial Imagery](#), SPIE Airborne Intelligence, Surveillance, Reconnaissance (ISR) Systems and Applications, Orlando, 2006.



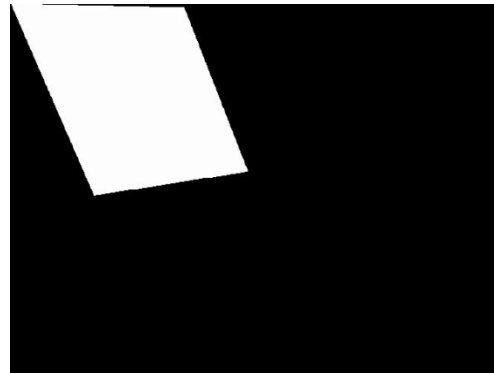
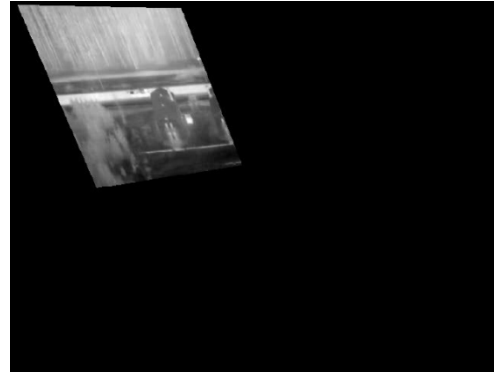
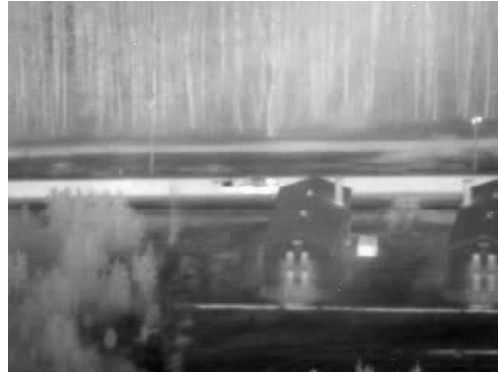
# COCOA – System Flow



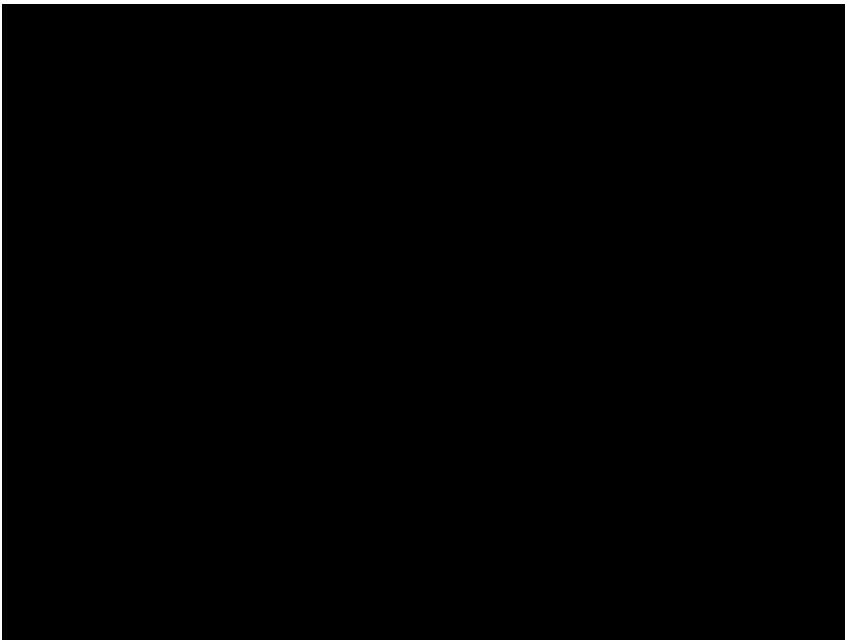
# Registration Result - I



# Registration Result - II



# Detection Results



# Tracking Results



# References

- J. Bergen, P. Anandan, K. Hanna, and R. Hingorani, “Hierarchical Model-Based Motion Estimation”, ECCV-92, pp 237-22.
- Video orbits of the projective group a simple approach to featureless estimation of parameters S Mann, RW Picard - Image Processing, IEEE Transactions on, 1997
- Saad Ali and Mubarak Shah, COCOA - Tracking in Aerial Imagery, SPIE Airborne Intelligence, Surveillance, Reconnaissance (ISR) Systems and Applications, Orlando, 2006.
- R. Szeliski. “Video mosaics for virtual environments”, IEEE Computer Graphics and Applications, pages,22-30, March 1996.