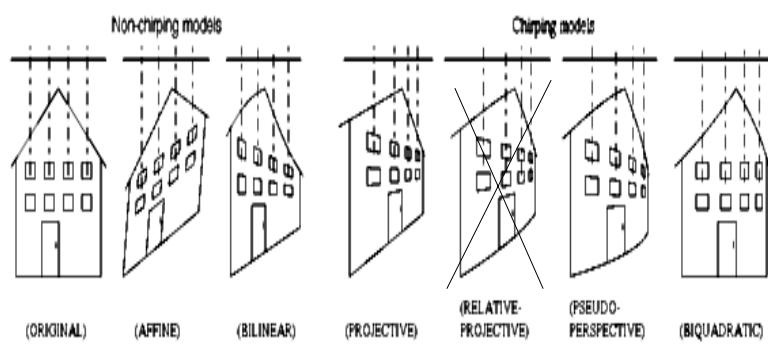


# Lecture-3



## Displacement Models (contd)



## Affine Mosaic



## Projective Mosaic



# Instantaneous Velocity Model



## 3-D Rigid Motion

$$\begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} 1 & -\alpha & \beta \\ \alpha & 1 & -\gamma \\ -\beta & \gamma & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix}$$
$$\begin{bmatrix} X' - X \\ Y' - Y \\ Z' - Z \end{bmatrix} = \begin{bmatrix} 0 & -\alpha & \beta \\ \alpha & 0 & -\gamma \\ -\beta & \gamma & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix}$$
$$\begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{Z} \end{bmatrix} = \begin{bmatrix} 0 & -\Omega_3 & \Omega_2 \\ \Omega_3 & 0 & -\Omega_1 \\ -\Omega_2 & \Omega_1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} V_1 \\ V_2 \\ V_3 \end{bmatrix}$$

$$\begin{bmatrix} X' \\ Y' \\ Z' \end{bmatrix} = \left( \begin{bmatrix} 0 & -\alpha & \beta \\ \alpha & 0 & -\gamma \\ -\beta & \gamma & 0 \end{bmatrix} + \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \right) \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix}$$
$$\begin{aligned} \dot{X} &= \Omega_2 Z - \Omega_3 Y + V_1 \\ \dot{Y} &= \Omega_3 X - \Omega_1 Z + V_2 \\ \dot{Z} &= \Omega_1 Y - \Omega_2 X + V_3 \end{aligned}$$



## 3-D Rigid Motion

$$\dot{X} = \Omega_2 Z - \Omega_3 Y + V_1$$

$$\dot{Y} = \Omega_3 X - \Omega_1 Z + V_2$$

$$\dot{Z} = \Omega_1 Y - \Omega_2 X + V_3$$

$$\mathbf{X} = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix},$$

$$\dot{\mathbf{X}} = \boldsymbol{\Omega} \times \mathbf{X} + \mathbf{V}$$

$$\boldsymbol{\Omega} = \begin{bmatrix} \Omega_1 \\ \Omega_2 \\ \Omega_3 \end{bmatrix}$$

Cross Product



## Orthographic Projection

$$\dot{X} = \Omega_2 Z - \Omega_3 Y + V_1$$

$$\dot{Y} = \Omega_3 X - \Omega_1 Z + V_2$$

$$\dot{Z} = \Omega_1 Y - \Omega_2 X + V_3$$

$$y = Y$$

$$x = X$$

$$u = \dot{x} = \Omega_2 Z - \Omega_3 y + V_1$$

$$v = \dot{y} = \Omega_3 x - \Omega_1 Z + V_2$$

(u,v) is optical flow



## Plane+orthographic(Affine)

$$Z = a + bX + cY$$

$$b_1 = V_1 + a\Omega_2$$

$$u = V_1 + \Omega_2 Z - \Omega_3 Y$$

$$a_1 = b\Omega_2$$

$$v = V_2 + \Omega_3 X - \Omega_1 Z$$

$$a_2 = c\Omega_2 - \Omega_3$$

(1) Show this  $u = b_1 + a_1 x + a_2 y$

$$v = b_2 + a_3 x + a_4 y$$



$$b_2 = V_2 - a\Omega_1$$

$$a_3 = \Omega_3 - b\Omega_1$$

$$a_4 = -c\Omega_1$$

$$\mathbf{u} = \mathbf{A} \mathbf{x} + \mathbf{b}$$



## Perspective Projection (arbitrary flow)

$$x = \frac{fX}{Z} \quad u = \dot{x} = \frac{fZ\dot{X} - fX\dot{Z}}{Z^2} = f \frac{\dot{X}}{Z} - x \frac{\dot{Z}}{Z}$$

$$y = \frac{fY}{Z} \quad v = \dot{y} = \frac{fZ\dot{Y} - fY\dot{Z}}{Z^2} = f \frac{\dot{Y}}{Z} - y \frac{\dot{Z}}{Z}$$

$$\dot{X} = \Omega_2 Z - \Omega_3 Y + V_1$$

$$\dot{Y} = \Omega_3 X - \Omega_1 Z + V_2$$

$$\dot{Z} = \Omega_1 Y - \Omega_2 X + V_3$$

$$u = f\left(\frac{V_1}{Z} + \Omega_2\right) - \frac{V_3}{Z}x - \Omega_3y - \frac{\Omega_1}{f}xy + \frac{\Omega_2}{f}x^2$$

$$v = f\left(\frac{V_2}{Z} - \Omega_1\right) + \Omega_3x - \frac{V_3}{Z}y + \frac{\Omega_2}{f}xy - \frac{\Omega_1}{f}y^2$$



## Plane+Perspective (pseudo perspective)

$$u = f\left(\frac{V_1}{Z} + \Omega_2\right) - \frac{V_3}{Z}x - \Omega_3y - \frac{\Omega_1}{f}xy + \frac{\Omega_2}{f}x^2 \quad Z = a + bX + cY$$

$$v = f\left(\frac{V_2}{Z} - \Omega_1\right) + \Omega_3x - \frac{V_3}{Z}y + \frac{\Omega_2}{f}xy - \frac{\Omega_1}{f}y^2 \quad \frac{1}{Z} = \frac{1}{a} - \frac{b}{a}x - \frac{c}{a}y$$



(2) Show this

$$u = a_1 + a_2x + a_3y + a_4x^2 + a_5xy$$

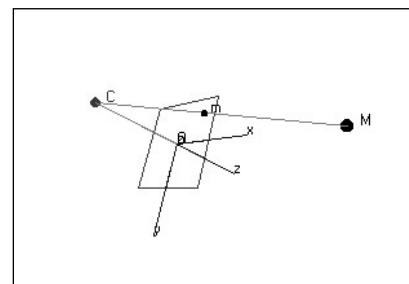
$$v = a_6 + a_7x + a_8y + a_4xy + a_5y^2$$



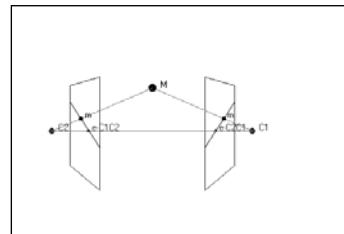
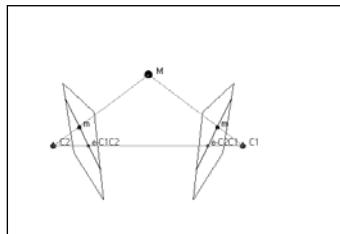
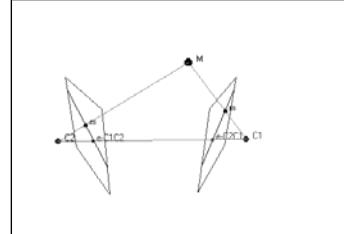
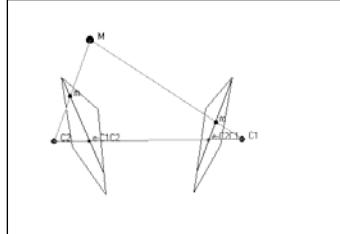
## 3D Scenes

- When the scene is neither planar nor distant, the relation between two consecutive frames is described by the *fundamental matrix*.

- What the heck is epipolar geometry?



## Epi-polar Geometry



## Epi-polar Geometry

### Coplanar Condition

$$(\mathbf{B} - \mathbf{A}) \cdot \mathbf{A} \times \mathbf{B} = 0$$

■  $\mathbf{M}' = \mathbf{R}(\mathbf{M} - \mathbf{T})$

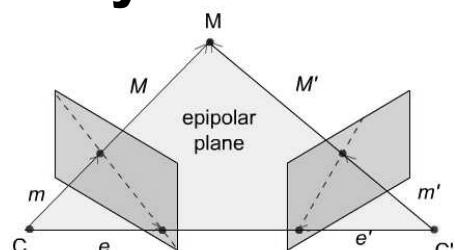
■  $(\mathbf{M} - \mathbf{T})^T \cdot \mathbf{T} \times \mathbf{M} = 0$

■  $\mathbf{R}^{-1} = \mathbf{R}^T$

■  $(\mathbf{R}^T \mathbf{M})^T \cdot \mathbf{T} \times \mathbf{M} = 0$

■  $\mathbf{A} \times \mathbf{B} = \mathbf{S} \cdot \mathbf{B}$  therefore  $\mathbf{M}'^T \mathbf{R} \mathbf{S} \mathbf{M} = 0$

■  $\mathbf{M}'^T \mathbf{E} \mathbf{M} = 0$  where  $\mathbf{E} = \mathbf{R} \mathbf{S}$  is the essential matrix



- a1**      The Essential Matrix is the mapping between points and epipolar lines  
subail, 10/9/2001

## Vector Cross Product to Matrix-vector multiplication

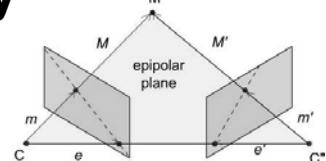
$$A \times B = SB$$

$$S = \begin{bmatrix} 0 & -A_z & A_y \\ A_z & 0 & -A_x \\ -A_y & A_x & 0 \end{bmatrix}$$



## Epipolar Geometry

- The *Fundamental Matrix* is the *Essential Matrix* described in pixel coordinates.



- $m$  and  $m'$  are the points in pixel coordinates corresponding to  $\mathbf{m}$  and  $\mathbf{m}'$

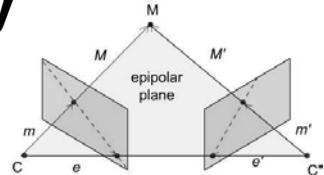
- $\mathbf{M} = \mathbf{P}^{-1} \mathbf{m}$  and  $\mathbf{M}' = \mathbf{P}'^{-1} \mathbf{m}'$

- $\mathbf{M}^T \mathbf{E} \mathbf{M} = \mathbf{0}$     $\mathbf{m}^T \mathbf{E} \mathbf{m} = \mathbf{0}$
- $(\mathbf{P}'^{-1} \mathbf{m}')^T \mathbf{E} (\mathbf{P}^{-1} \mathbf{m}) = \mathbf{m}'^T \mathbf{P}'^{-T} \mathbf{E} \mathbf{P}^{-1} \mathbf{m}$



# Epipolar Geometry

- The *Fundamental Matrix* is the *Essential Matrix* described in pixel coordinates.
- $(P'^{-1}m')^T E (P^{-1}m) = m'^T P'^{-T} E P^{-1} m = 0$
- $m^T F m = 0$  where  $F$  is the fundamental matrix
- $F = P'^{-T} E P^{-1}$  The  $P$ s are the Intrinsic parameters
- The Fundamental Matrix maps a point to the it's epipolar line in pixel coordinates



## Fundamental Matrix

- Fundamental matrix captures the relationship between the corresponding points in two views.

$$m^T F m' = 0,$$

$$\begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix}^T \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix} \begin{bmatrix} x'_i \\ y'_i \\ 1 \end{bmatrix} = 0,$$

$$\begin{bmatrix} x_i \\ y_i \\ 1 \end{bmatrix}^T \begin{bmatrix} f_{11}x' + f_{12}y'_i + f_{13} \\ f_{21}x' + f_{22}y'_i + f_{23} \\ f_{31}x' + f_{32}y'_i + f_{33} \end{bmatrix} = 0,$$

$$x_i(f_{11}x' + f_{12}y'_i + f_{13}) + y_i(f_{21}x' + f_{22}y'_i + f_{23}) + (f_{31}x' + f_{32}y'_i + f_{33}) = 0$$



## Fundamental Matrix

$$x_i(f_{11}x' + f_{12}y'_i + f_{13}) + y_i(f_{21}x' + f_{22}y'_i + f_{23}) + (f_{31}x' + f_{32}y'_i + f_{33}) = 0$$

$$x_i x' f_{11} + x_i y'_i f_{12} + x_i f_{13} + y_i x' f_{21} + x' y'_i f_{22} + y'_i f_{23} + x' f_{31} + y'_i f_{32} + f_{33} = 0$$

$$Mf = \begin{bmatrix} x'_1 x_1 & x'_1 y_1 & x'_1 & y'_1 x_1 & y'_1 y_1 & y'_1 & x_1 & y_1 & 1 \\ x'_2 x_2 & x'_2 y_2 & x'_2 & y'_2 x_2 & y'_2 y_2 & y'_2 & x_2 & y_2 & 1 \\ \vdots & \vdots \\ x'_n x_n & x'_n y_n & x'_n & y'_n x_n & y'_n y_n & y'_n & x_n & y_n & 1 \end{bmatrix} f = 0$$

$$M \text{ is } 9 \text{ by } n \text{ matrix} \quad f = [f_{11} \quad f_{12} \quad f_{13} \quad f_{21} \quad f_{22} \quad f_{23} \quad f_{31} \quad f_{32} \quad f_{33}]$$

To solve the equation, the rank(M) must be 8.



## Computation of Fundamental Matrix



# Normalized 8-point algorithm

- Objective:
  - Compute fundamental matrix  $F$  such that  $\mathbf{x}_i' F \mathbf{x}_i = 0$
- Algorithm
  - Normalize the image by  $\hat{\mathbf{x}}_i = T \mathbf{x}_i$  and  $\hat{\mathbf{x}}'_i = T' \mathbf{x}'_i$ 
    - Find centroid of points in each image, determine the range, and normalize all points between 0 and 1
  - Linear solution
    - determining the eigen vector corresponding to the smallest eigen value of  $A$ ,
$$Af = \begin{bmatrix} \hat{x}_1 \hat{x}_1 & \hat{x}_1 \hat{y}_1 & \hat{x}'_1 & \hat{y}'_1 \hat{x}_1 & \hat{y}'_1 \hat{y}_1 & \hat{y}'_1 & \hat{x}_1 & \hat{y}_1 & 1 \\ \dots & \dots \\ \hat{x}'_8 \hat{x}_8 & \hat{x}'_8 \hat{y}_8 & \hat{x}'_8 & \hat{y}'_8 \hat{x}_8 & \hat{y}'_8 \hat{y}_8 & \hat{y}'_8 & \hat{x}'_8 & \hat{y}'_8 & 1 \end{bmatrix} f = 0$$
  - Construct  $\hat{F} = \begin{bmatrix} f_1 & f_2 & f_3 \\ f_4 & f_5 & f_6 \\ f_7 & f_8 & f_9 \end{bmatrix}$
  - Normalize  $\hat{F} = \hat{F} / \|\hat{F}\|$



# Normalized 8-point algorithm

- Constraint enforcement
  - SVD decomposition
$$\hat{F} = U \begin{bmatrix} \sigma_1 & 0 & 0 \\ 0 & \sigma_2 & 0 \\ 0 & 0 & \sigma_3 \end{bmatrix} V'$$

$$(\sigma_1 \geq \sigma_2 \geq \sigma_3)$$
- Rank enforcement
- De-normalization:

$$F = T^{-1T} \tilde{F} T \quad (3) \text{ Show this}$$

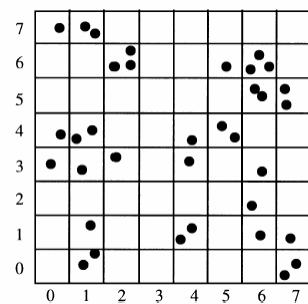


$$T = \begin{bmatrix} a_x & 0 & d_x \\ 0 & a_y & d_y \\ 0 & 0 & 1 \end{bmatrix}$$



## Robust Fundamental Matrix Estimation (by Zhang)

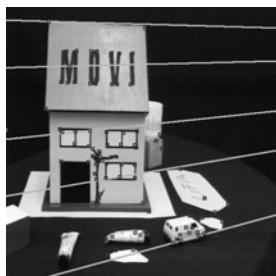
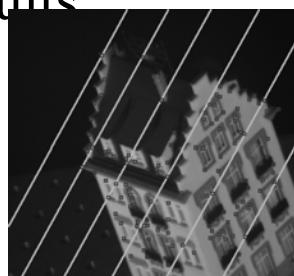
- Uniformly divide the image into  $8 \times 8$  grid.
- Randomly select 8 grid cells and pick one pair of corresponding points from each grid cell, then use Hartley's 8-point algorithm to compute Fundamental Matrix  $F_i$ .
- For each  $F_i$ , compute the median of the squared residuals  $R_i$ .
  - $R_i = \text{median}_k [d(p_{1k}, F_i p_{2k}) + d(p_{2k}, F_i p_{1k})]$
- Select the best  $F_i$  according to  $R_i$ .
- Remove outlier correspondences if  $R_k > Th$ .
- Using the remaining points compute the fundamental Matrix  $F$  by weighted least square method.



# Results



# Results



## Homework Due 9/16/2004

- Show (1), (2) and (3).

