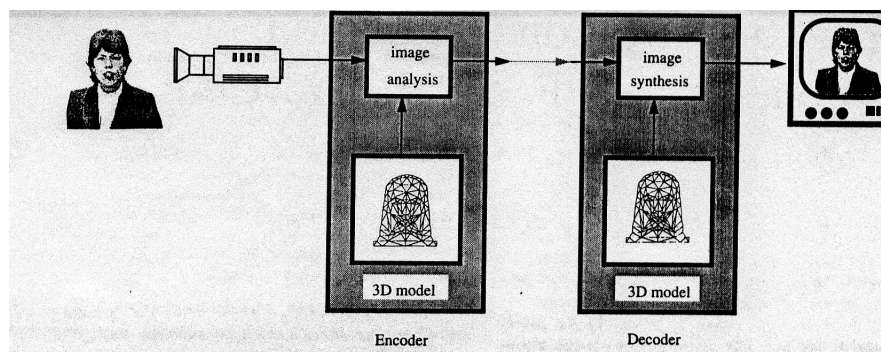


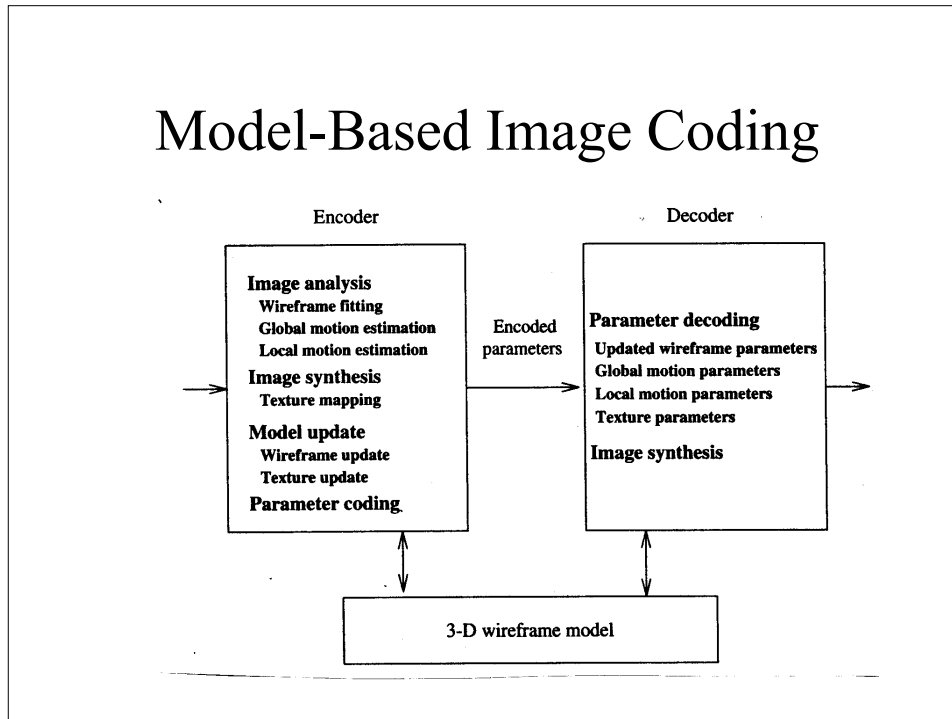
# Lecture-9

Model-base Video Compression  
Li, Teklap

## Model-Based Image Coding



# Model-Based Image Coding



# Model-Based Image Coding

- The transmitter and receiver both possess the same 3D face model and texture images.
- During the session, at the transmitter the facial motion parameters: global and local, are extracted.
- At the receiver the image is synthesized using estimated motion parameters.
- The difference between synthesized and actual image can be transmitted as residuals.

# Candide Model

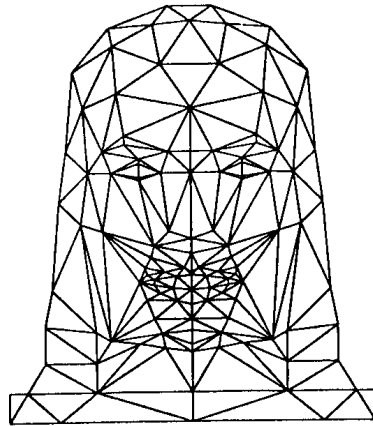


Fig. 2. Wire-frame model of the face.

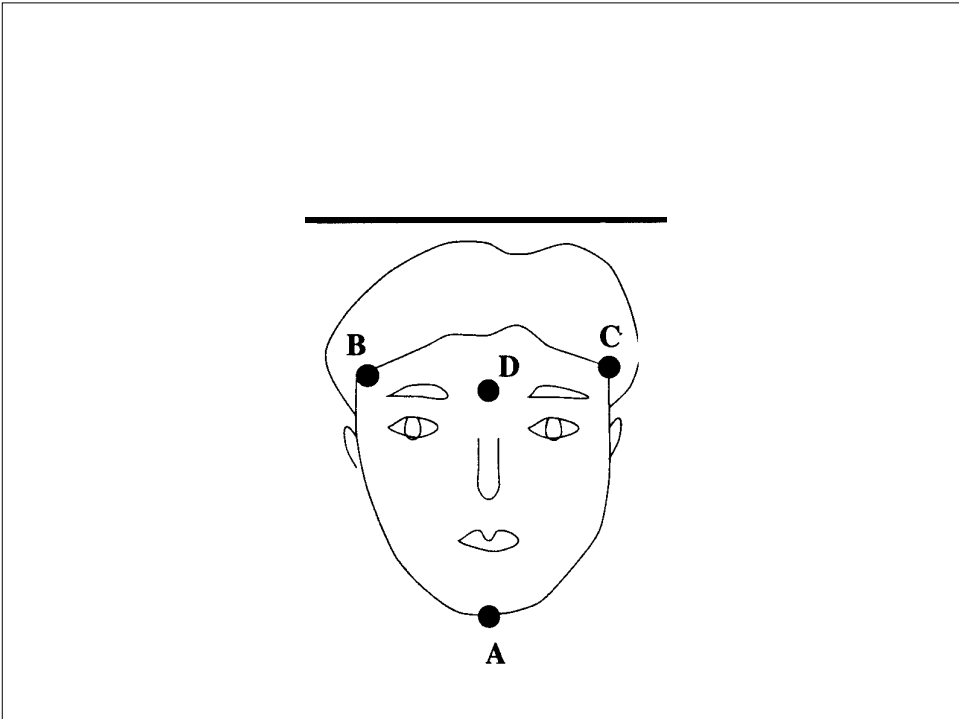
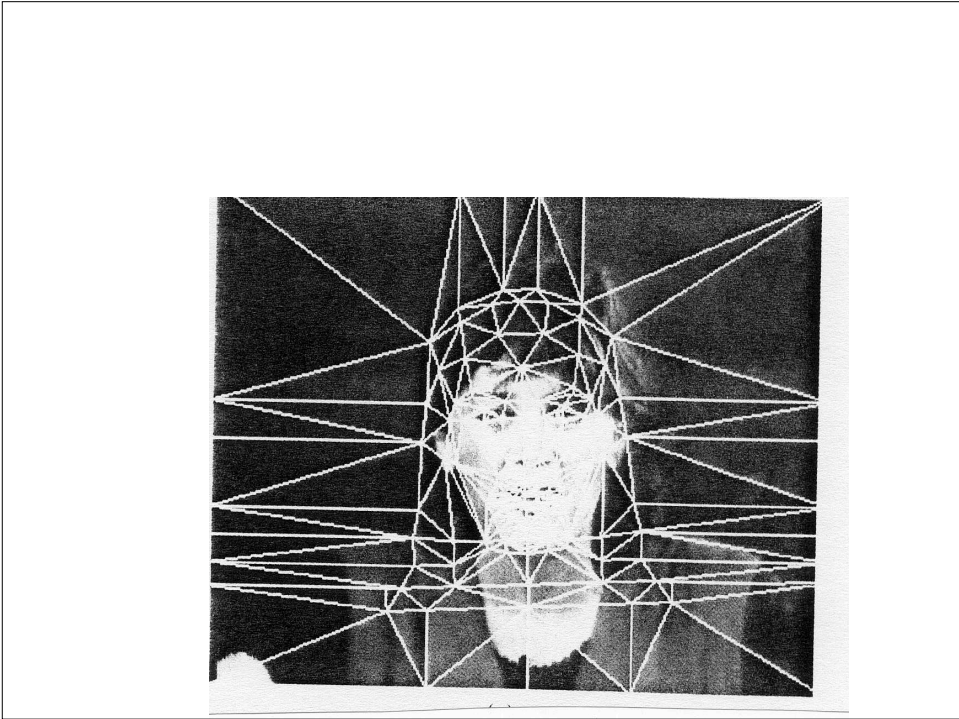
[../CANDIDE.HTM](#)

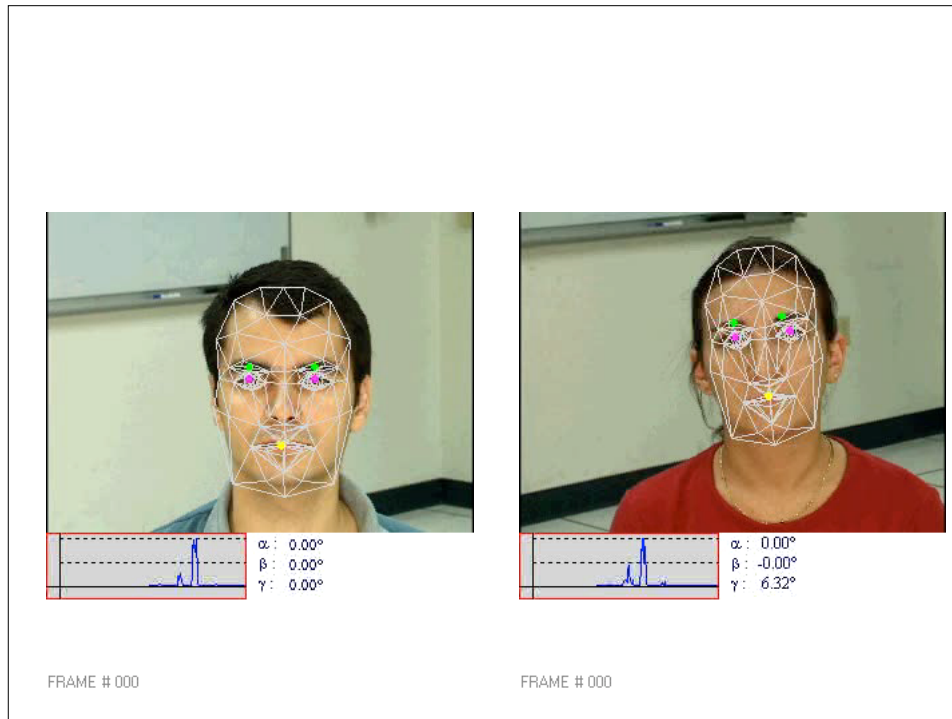
## Face Model

- Candide model has 108 nodes, 184 polygons.
- Candide is a generic head and shoulder model. It needs to be conformed to a particular person's face.
- Cyberware scan gives head model consisting of 460,000 polygons.

## Wireframe Model Fitting

- Fit orthographic projection of wireframe to the frontal view of speaker using Affine transformation.
  - Locate three to four features in the image and the projection of a model.
  - Find parameters of Affine transformation using least squares fit.
  - Apply Affine to all vertices, and scale  $\sqrt{(a_1^2 + a_2^2)}/2$  depth.

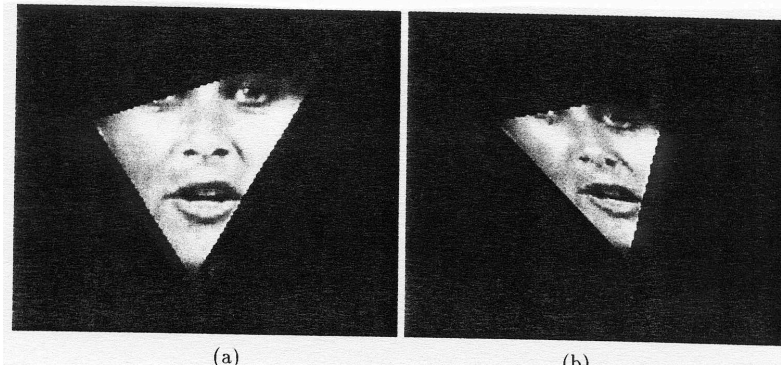




## Synthesis

- Collapse initial wire frame onto the image to obtain a collection of triangles.
- Map observed texture in the first frame into respective triangles.
- Rotate and translate the initial wire frame according to global and local motion, and collapse onto the next frame.
- Map texture within each triangle from first frame to the next frame by interpolation.

## Texture Mapping



## Video Phones

Motion Estimation

### Perspective Projection (optical flow)

$$u = f\left(\frac{V_1}{Z} + \alpha_2\right) - \frac{V_3}{Z}x - \alpha_3y - \frac{\alpha_1}{f}xy + \frac{\alpha_2}{f}x^2$$

$$v = f\left(\frac{V_2}{Z} - \alpha_1\right) + \alpha_3x - \frac{V_3}{Z}y + \frac{\alpha_2}{f}xy - \frac{\alpha_1}{f}y^2$$

### Optical Flow Constraint Eq

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



$$\begin{aligned}
& f_x \left( f \left( \frac{V_1}{Z} + \square_2 \right) \square \frac{V_3}{Z} x \square \square_3 y \square \frac{\square_1}{f} xy + \frac{\square_2}{f} x^2 \right) + f_y \\
& \left( f \left( \frac{V_2}{Z} \square \square_1 \right) + \square_3 x \square \frac{V_3}{Z} y + \frac{\square_2}{f} xy \square \frac{\square_1}{f} y^2 \right) + f_t = 0 \\
& \left( f_x \frac{f}{Z} \right) V_1 + \left( f_y \frac{f}{Z} \right) V_2 + \left( \frac{f}{Z} (f_x x \square f_y y) \right) V_3 + \\
& \left( \square f_x \frac{xy}{f} + f_y \frac{y^2}{f} \square f_y f \right) \square_1 + \left( f_x f + f_x \frac{x^2}{f} + f_y \frac{xy}{f} \right) \square_2 + \\
& (f_x y + f_y x) \square_3 = \square f_t
\end{aligned}$$

$$\begin{aligned}
& \left( f_x \frac{f}{Z} \right) V_1 + \left( f_y \frac{f}{Z} \right) V_2 + \left( \frac{f}{Z} (f_x x \square f_y y) \right) V_3 + \\
& \left( \square f_x \frac{xy}{f} + f_y \frac{y^2}{f} \square f_y f \right) \square_1 + \left( f_x f + f_x \frac{x^2}{f} + f_y \frac{xy}{f} \right) \square_2 + \\
& (f_x y + f_y x) \square_3 = \square f_t
\end{aligned}$$

**$\mathbf{Ax} = \mathbf{b}$**  Solve by Least Squares

$$\mathbf{x} = (V_1, V_2, V_3, \square_1, \square_2, \square_3)$$



## Problem

- We have used 3D rigid motion, but face is not purely rigid!
- Facial expressions produce non-rigid motion.
- Use global rigid motion and non-rigid deformations.

## 3-D Rigid Motion

$$\mathbf{X}' = \mathbf{R}\mathbf{X} + \mathbf{T}$$

$$\begin{bmatrix} X' \\ Y' \\ Z' \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix}$$

$$\begin{bmatrix} X' \\ Y' \\ Z' \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix}$$

## 3-D Rigid Motion

$$\begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{Z} \end{bmatrix} = \begin{bmatrix} 0 & \omega & 0 \\ -\omega & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} V_x \\ V_y \\ V_z \end{bmatrix}$$

$$\begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{Z} \end{bmatrix} = \begin{bmatrix} 0 & \omega_3 & \omega_2 \\ \omega_3 & 0 & \omega_1 \\ \omega_2 & \omega_1 & 0 \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} + \begin{bmatrix} V_x \\ V_y \\ V_z \end{bmatrix}$$

$$\dot{\mathbf{X}} = \boldsymbol{\omega} \mathbf{X} + \mathbf{V}$$

## 3-D Rigid+Non-rigid Motion

$$\mathbf{X} = \mathbf{R}\mathbf{X} + \mathbf{T} + \mathbf{E}(\mathbf{X})\boldsymbol{\alpha}$$

Facial expressions

Action Units:

- opening of a mouth
- closing of eyes
- raising of eyebrows

$$\boldsymbol{\alpha} = (\alpha_1, \alpha_2, \dots, \alpha_m)^T$$

### 3-D Rigid+Non-rigid Motion

$$\begin{aligned}
 \begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{Z} \end{bmatrix} &= \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} + \sum_{i=1}^m \begin{bmatrix} E_{1i}(\mathbf{X}) \\ E_{2i}(\mathbf{X}) \\ E_{3i}(\mathbf{X}) \end{bmatrix} \\
 \begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{Z} \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \sum_{i=1}^m \begin{bmatrix} E_{1i}(\mathbf{X}) \\ E_{2i}(\mathbf{X}) \\ E_{3i}(\mathbf{X}) \end{bmatrix}
 \end{aligned}$$

### 3-D Rigid+Non-rigid Motion

$$\begin{aligned}
 \begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{Z} \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \sum_{i=1}^m \begin{bmatrix} E_{1i}(\mathbf{X}) \\ E_{2i}(\mathbf{X}) \\ E_{3i}(\mathbf{X}) \end{bmatrix} \\
 \begin{bmatrix} \dot{X} \\ \dot{Y} \\ \dot{Z} \end{bmatrix} &= \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} + \sum_{i=1}^m \begin{bmatrix} E_{1i}(\mathbf{X}) \\ E_{2i}(\mathbf{X}) \\ E_{3i}(\mathbf{X}) \end{bmatrix}
 \end{aligned}$$

$$\dot{\mathbf{X}} = \mathbf{D} \mathbf{X} + \mathbf{D}$$

## 3-D Rigid+Non-rigid Motion

$$\dot{X} = \omega_3 Y + \omega_2 Z + V_1 + \sum_{i=1}^m E_{1i} \omega_i$$

$$\dot{Y} = \omega_3 X - \omega_1 Z + V_2 + \sum_{i=1}^m E_{2i} \omega_i$$

$$\dot{Z} = \omega_2 X + \omega_1 Y + V_3 + \sum_{i=1}^m E_{3i} \omega_i$$

## Perspective Projection (arbitrary flow)

$$x = \frac{fX}{Z}$$

$$y = \frac{fY}{Z}$$

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

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

### Perspective Projection (arbitrary flow)

$$\begin{aligned} \dot{X} &= \alpha_3 Y + \alpha_2 Z + V_1 + \sum_{i=1}^m E_{1i} \Delta_i & u = \dot{x} &= \frac{fZ\dot{X} - fX\dot{Z}}{Z^2} = f \frac{\dot{X}}{Z} - x \frac{\dot{Z}}{Z} \\ \dot{Y} &= \alpha_3 X + \alpha_1 Z + V_2 + \sum_{i=1}^m E_{2i} \Delta_i & v = \dot{y} &= \frac{fZ\dot{Y} - fY\dot{Z}}{Z^2} = f \frac{\dot{Y}}{Z} - y \frac{\dot{Z}}{Z} \\ \dot{Z} &= \alpha_2 X + \alpha_1 Y + V_3 + \sum_{i=1}^m E_{3i} \Delta_i \end{aligned}$$

$$u = f \left( \frac{V_1 + \sum_{i=1}^m E_{1i} \Delta_i}{Z} + \alpha_2 \right) - \frac{V_3 + \sum_{i=1}^m E_{3i} \Delta_i}{Z} x - \alpha_3 y - \frac{\alpha_1}{f} x + \frac{\alpha_2}{f} x^2$$

$$v = f \left( \frac{V_2 + \sum_{i=1}^m E_{2i} \Delta_i}{Z} - \alpha_1 \right) + \alpha_3 x - \frac{V_3 + \sum_{i=1}^m E_{3i} \Delta_i}{Z} y + \frac{\alpha_2}{f} xy - \frac{\alpha_1}{f} y$$

### Optical Flow Constraint Eq

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

$$\mathbf{Ax} = \mathbf{b}$$

$$\mathbf{x} = (V_1, V_2, V_3, \alpha_1, \alpha_2, \alpha_3, \beta_1, \beta_2, \dots, \beta_m)$$





## Lecture-10

# Estimation Using Flexible Wireframe Model

## Main Points

- Model photometric effects
- Simultaneously compute 3D motion and adapt the wireframe model.

## Generalized Optical Flow Constraint

$$f(x, y, t) = \rho N(t) \cdot L$$

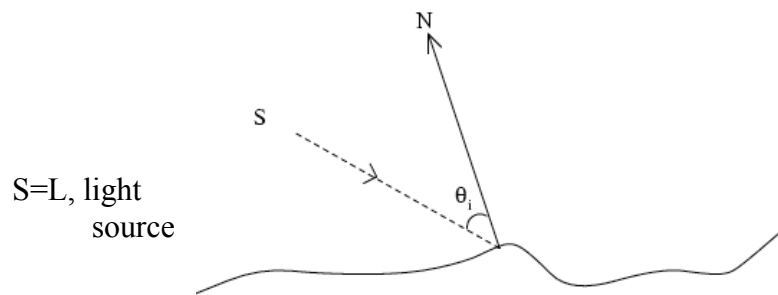
Lambertian Model

$$\frac{df(x, y, t)}{dt} = \rho L \cdot \frac{dN}{dt}$$

Albedo  
Surface Normal  
(-p, -q, 1)

$$f_x u + f_y v + f_t = \rho L \cdot \frac{dN}{dt}$$

## Lambertian Model



$$f(x, y) = n \cdot L = (n_x, n_y, n_z) \cdot (l_x, l_y, l_z)$$

$$f(x, y) = n \cdot L = \left( \frac{1}{\sqrt{p^2 + q^2 + 1}} (\ominus p, \ominus q, 1) \right) \cdot (l_x, l_y, l_z)$$

## Sphere

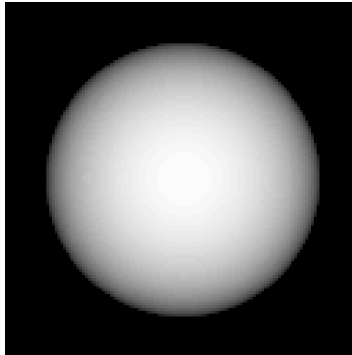
$$z = \sqrt{(R^2 - x^2 - y^2)}$$

$$p = \frac{\partial z}{\partial x} = \ominus \frac{x}{z}$$

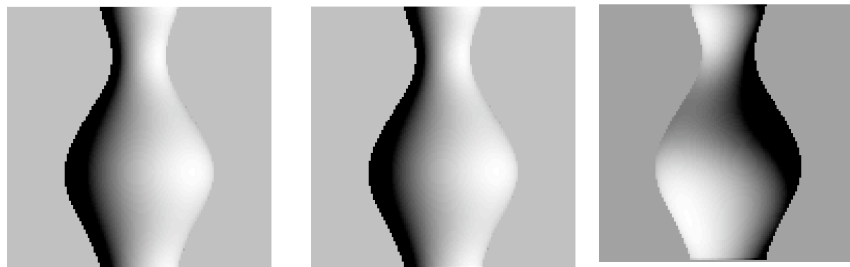
$$q = \frac{\partial z}{\partial y} = \ominus \frac{y}{z}$$

$$(n_x, n_y, n_z) = \frac{1}{R} (x, y, z)$$

# Sphere



# Vase



$(1, 0, 1)$

$(-1, 1, 1)$

$(-1, -1, 1)$

## Orthographic Projection

$$u = \dot{x} = \alpha_2 Z - \alpha_3 y + V_1 \quad (u,v) \text{ is optical flow}$$

$$v = \dot{y} = \alpha_3 x - \alpha_1 Z + V_2$$

$$\dot{\mathbf{X}} = \alpha \mathbf{X} + \mathbf{V}$$

$$\dot{X} = \alpha_2 Z - \alpha_3 Y + V_1$$

$$\dot{Y} = \alpha_3 X - \alpha_1 Z + V_2$$

$$\dot{Z} = \alpha_1 Y - \alpha_2 X + V_3$$

## Optical flow equation

$$f_x(\alpha_2 Z - \alpha_3 y + V_1) + f_y(\alpha_3 x - \alpha_1 Z + V_2) + f_t = \alpha L \frac{dN}{dt}$$

$$f_x(\alpha_2 Z - \alpha_3 y + V_1) + f_y(\alpha_3 x - \alpha_1 Z + V_2) + f_t =$$

$$\alpha L \frac{(\alpha p - \alpha q \alpha)^T}{\sqrt{p^2 + q^2 + 1}} - \frac{(\alpha p, \alpha q, 1)^T}{\sqrt{p^2 + q^2 + 1}}$$

Homework 3.1  
Show this.

Equation 24.8, page 473.

## Error Function

$$E = \sum_i \int_{(x,y) \in \text{thpatch}} e_i^2$$

$$p_i x_1^{(ij)} + q_i x_2^{(ij)} + c_i = p_j x_1^{(ij)} + q_j x_2^{(ij)} + c_j$$

constraint

$$e_i(x, y) = f_x(\alpha_3 y - \alpha_2(p_i x + q_i y + c_i) + V_1) + f_y(\alpha_3 x + \alpha_1(p_i x + q_i y + c_i) + V_2) + f_t$$

$$\alpha \alpha(L_1, L_2, L_3) \cdot \left( \frac{(\alpha_2 + p_i, \alpha_1 + q_i)}{(1 + \alpha_2 p_i, 1 + \alpha_1 q_i)} \right) \cdot \left( \frac{(\alpha_2 + p_i)^2 + (\alpha_1 + q_i)^2 + 1}{(1 + \alpha_2 p_i)^2 + (1 + \alpha_1 q_i)^2 + 1} \right)^{1/2}$$

$$\frac{(\alpha p_i, \alpha q_i, 1)}{(p_i^2 + q_i^2 + 1)^{1/2}}$$

Homework 3.3  
Show this.

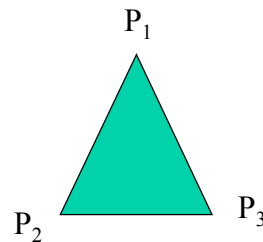
## Equation of a Planar Patch

$$P_1^{(i)} = (X_1^{(i)}, Y_1^{(i)}, Z_1^{(i)})$$

$$P_2^{(i)} = (X_2^{(i)}, Y_2^{(i)}, Z_2^{(i)})$$

$$P_3^{(i)} = (X_3^{(i)}, Y_3^{(i)}, Z_3^{(i)})$$

$$P^{(i)} = (X^{(i)}, Y^{(i)}, Z^{(i)})$$



$$\overline{P^{(i)} P_1^{(i)}} \cdot \overline{(P_2^{(i)} P_1^{(i)} \times P_3^{(i)} P_1^{(i)})} = 0$$

## Equation of a Planar Patch

$$Z^{(i)} = p_i X^{(i)} + q_i Y^{(i)} + c_i$$

Homework 3.2  
Show this.

$$p_i = \frac{(Y_2^{(i)} - Y_1^{(i)})(Z_3^{(i)} - Z_1^{(i)}) - (Z_2^{(i)} - Z_1^{(i)})(Y_3^{(i)} - Y_1^{(i)})}{(X_2^{(i)} - X_1^{(i)})(Y_3^{(i)} - Y_1^{(i)}) - (Y_2^{(i)} - Y_1^{(i)})(X_3^{(i)} - X_1^{(i)})}$$

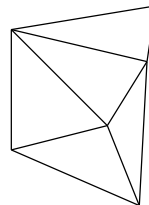
$$q_i = \frac{(Z_2^{(i)} - Z_1^{(i)})(X_3^{(i)} - X_1^{(i)}) - (X_2^{(i)} - X_1^{(i)})(Z_3^{(i)} - Z_1^{(i)})}{(X_2^{(i)} - X_1^{(i)})(Y_3^{(i)} - Y_1^{(i)}) - (Y_2^{(i)} - Y_1^{(i)})(X_3^{(i)} - X_1^{(i)})}$$

$$c_i = Z_1^{(i)} + X_1^{(i)} \frac{(Y_2^{(i)} - Y_1^{(i)})(Z_3^{(i)} - Z_1^{(i)}) - (Z_2^{(i)} - Z_1^{(i)})(Y_3^{(i)} - Y_1^{(i)})}{(X_2^{(i)} - X_1^{(i)})(Y_3^{(i)} - Y_1^{(i)}) - (Y_2^{(i)} - Y_1^{(i)})(X_3^{(i)} - X_1^{(i)})} +$$

$$Y_1^{(i)} \frac{(Z_2^{(i)} - Z_1^{(i)})(X_3^{(i)} - X_1^{(i)}) - (X_2^{(i)} - X_1^{(i)})(Z_3^{(i)} - Z_1^{(i)})}{(X_2^{(i)} - X_1^{(i)})(Y_3^{(i)} - Y_1^{(i)}) - (Y_2^{(i)} - Y_1^{(i)})(X_3^{(i)} - X_1^{(i)})}$$

## Structure of Wireframe Model

- Each triangular patch is either surrounded by two (if it is on the boundary of the wireframe) or three other triangles.



Neighboring patches must intersect at a straight line.

$$Z^{(i)} = p_i X^{(i)} + q_i Y^{(i)} + c_i$$

$$p_i x^{(ij)} + q_i y^{(ij)} + c_i = p_j x^{(ij)} + q_j y^{(ij)} + c_j$$

## Main Points of Algorithm

- Stochastic relaxation.
- Each iteration visit all patches in a sequential order.
  - If, at present iteration none of neighboring patches of  $i$  have been visited yet, then  $p_i$   $q_i$   $c_i$  are all independently perturbed.
  - If, only one of the neighbor,  $j$ , has been visited, then two parameters, say  $p_i$   $q_i$ , are independent and perturbed. The dependent variable  $c_i$  is calculated from the equation:
 
$$c_i = p_j x^{(ij)} + q_j y^{(ij)} + c_j - p_i x^{(ij)} - q_i y^{(ij)}$$



## Main Points of Algorithm

- If two of the neighboring patches, say j and k, have already been visited, i.e., the variables  $p_k$ ,  $q_k$ ,  $c_k$  and  $p_j$ ,  $q_j$ ,  $c_j$  have been updated, then only one variable  $p_i$  is independent, and is perturbed.  $q_i$ ,  $c_i$  can be evaluated as

$$c_i = p_j x^{(ij)} + q_j y^{(ij)} + c_j - p_i x^{(ij)} - q_i y^{(ij)}$$

$$q_i = \frac{p_k x^{(ik)} + q_k y^{(ik)} + c_k - p_i x^{(ik)} - c_i}{y^{(ik)}}$$

## Main Points of Algorithm

- The perturbation of structure parameters (surface normal) for each patch, results in the change of coordinates (X,Y,Z) of the nodes of wireframe.

## Updating of (X,Y,Z):

Patches i, j, k intersect at node n.

$$p_i X^{(n)} + q_i Y^{(n)} + c_i = p_j X^{(n)} + q_j Y^{(n)} + c_j$$

$$p_i X^{(n)} + q_i Y^{(n)} + c_i = p_k X^{(n)} + q_k Y^{(n)} + c_k$$

$$\begin{bmatrix} X^{(n)} \\ Y^{(n)} \end{bmatrix} = \begin{bmatrix} p_i & p_j & q_i & q_j \\ p_i & p_k & q_i & q_k \end{bmatrix}^{-1} \begin{bmatrix} c_j - c_i \\ c_k - c_i \end{bmatrix}$$

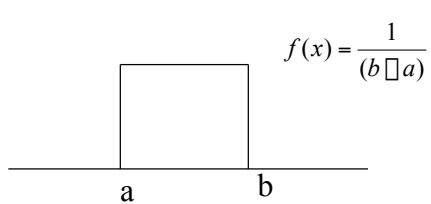
$$Z^{(i)} = p_i X^{(i)} + q_i Y^{(i)} + c_i$$

## Algorithm

- Estimate light source direction
- Initialize coordinates of all nodes using approximately scaled wireframe model (program-2). Determine initial values of surface normals for each triangle.
- Determine initial motion parameters based on selected feature correspondences and their depth values from wireframe model. (Assume motion parameters.)
- (A) Compute the value of error function E.

- If error E is less than some threshold, then stop
- Else
  - Perturb motion parameters (3 rotations and 2 translations) by random amount (zero mean Gaussian, s.d. equal to error E) (you can use uniform distribution)
  - Perturb structure parameters (p,q,c):
    - Perturb p, q, and c of first patch by adding random amount (zero mean Gaussian, s.d. equal to error E)
    - Increment count for all neighbors of patch-1 by 1

## Uniform Distribution



$$\bar{X} = \text{mean} = \frac{(a+b)}{2}$$

$$\sigma^2 = \text{variance} = \frac{(a-b)^2}{12}$$

Use rand() in C to generate random number between a range.

- For patch 2 to n
  - If the count==1
    - » Perturb p and q
    - » Compute c using equation for  $c_i$
    - » Increment the count

$$c_i = p_j x^{(ij)} + q_j y^{(ij)} + c_j \left[ p_i x^{(ij)} \right] \left[ q_i y^{(ij)} \right]$$

- If count==2
  - » Perturb  $p_i$
  - » Compute  $c_i$  and  $q_i$  using equations
 
$$c_i = p_j x^{(ij)} + q_j y^{(ij)} + c_j \left[ p_i x^{(ij)} \right] \left[ q_i y^{(ij)} \right]$$

$$q_i = \frac{p_k x^{(ik)} + q_k y^{(ik)} + c_k \left[ p_i x^{(ik)} \right] \left[ c_i \right]}{y^{(ik)}}$$
  - » Increment the count
- If p, q and c for at least three patches intersecting at node are updated, then update the coordinates of the node using equation.

$$\begin{bmatrix} X^{(n)} \\ Y^{(n)} \end{bmatrix} = \begin{bmatrix} p_i & p_j & p_j & p_k \\ q_i & q_j & q_j & q_k \end{bmatrix} \begin{bmatrix} c_j \\ c_i \\ c_j + c_k \\ c_i \end{bmatrix}$$

$$Z^{(i)} = p_i X^{(i)} + q_i Y^{(i)} + c_i$$

- Go to step (A)