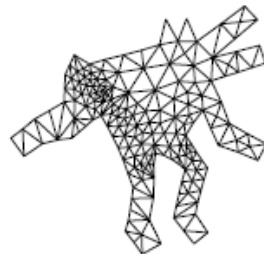
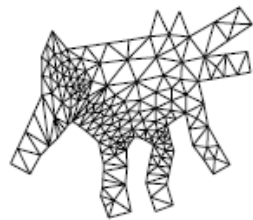
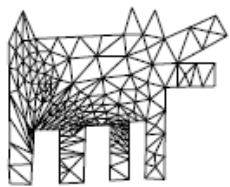


# Rigid Morphing

14 May 2015

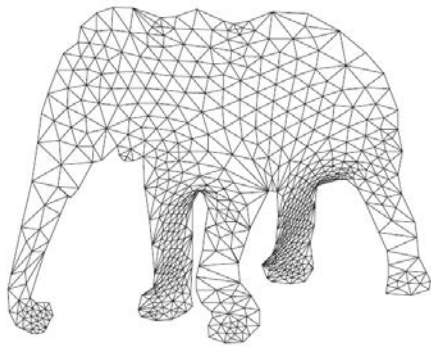
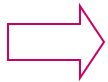
# As-rigid-as-possible shape interpolation

› *Alexa et al.*

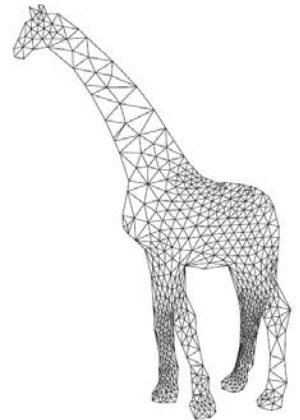


# As Rigid As Possible

linear interpolation

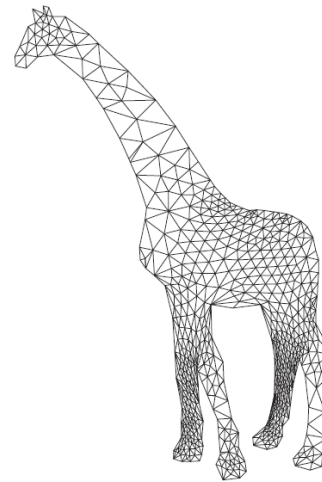
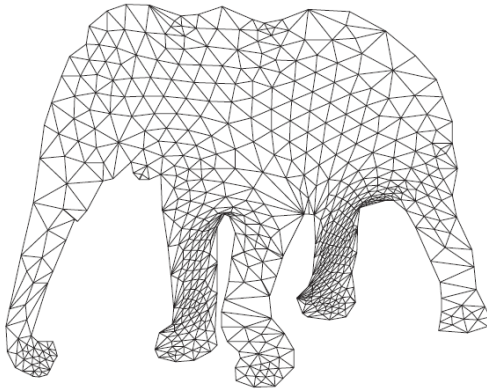


least-distorting morphing



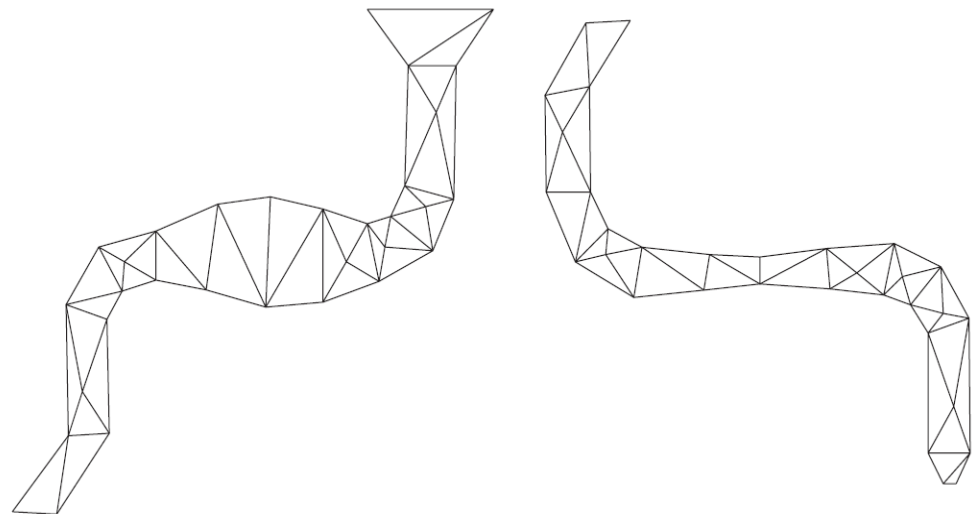
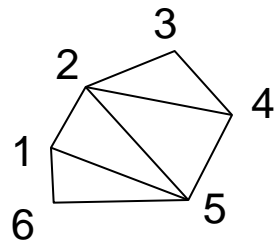
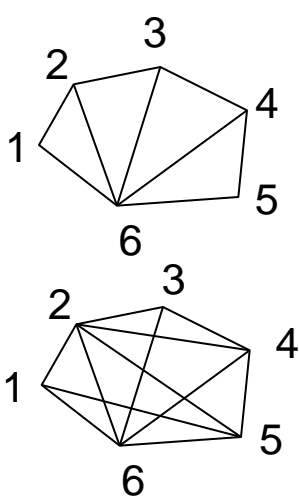
# Isomorphic Dissections of Shapes

- › Compatible triangulation
  - › Constructing a common triangulation for two given polygons



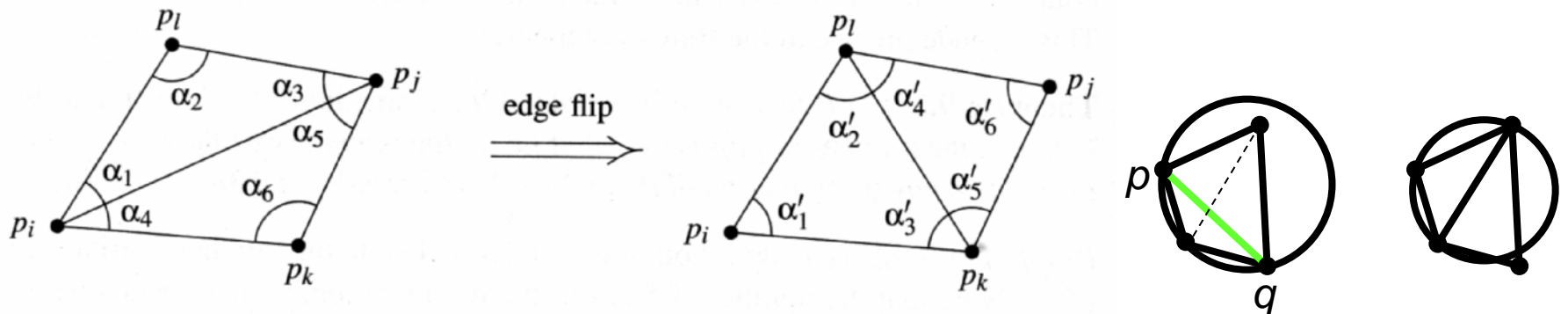
# Basic Algorithm

- › 1. Triangulate each polygon independently, using Delaunay triangulations
- › 2. Map both polygons to a regular  $n$ -gon so that **corresponding boundary vertices coincide**
- › 3. The compatible triangulation is established by **overlaying the two edge sets** in the convex  $n$ -gon. The resulting new interior vertices are then mapped back into the original polygons.



# Mesh Refinement

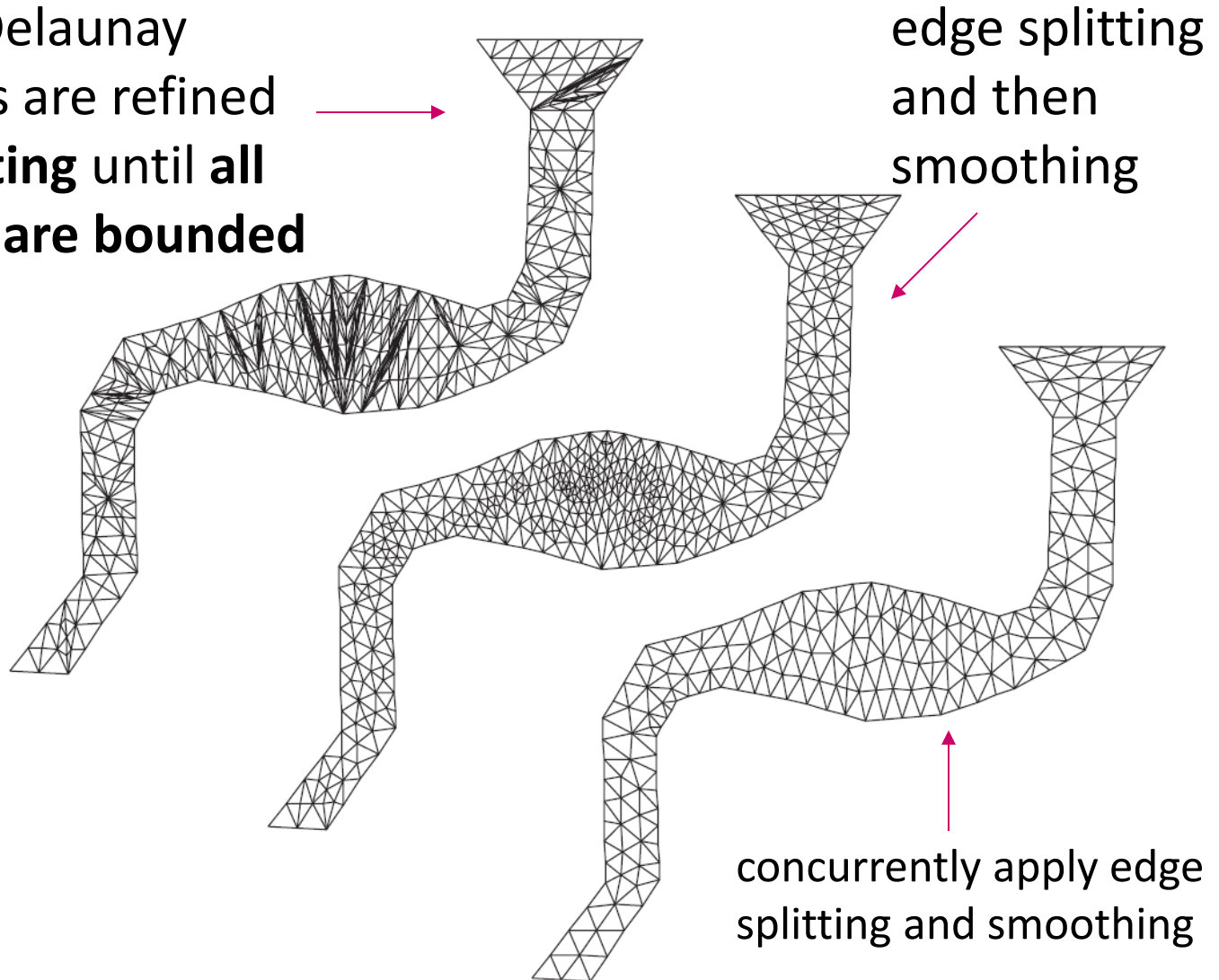
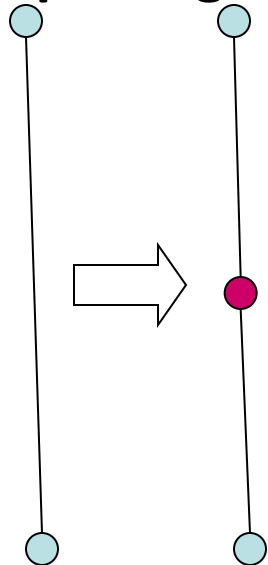
- › Smoothing, applying Delaunay criterion
  - › **Moving interior vertices** to find vertex positions that **maximize the minimum angle** for a given triangulation
  - › **Flipping interior edges** simultaneously in both triangulations. This procedure follows the edge flip criteria used in Delaunay triangulation.



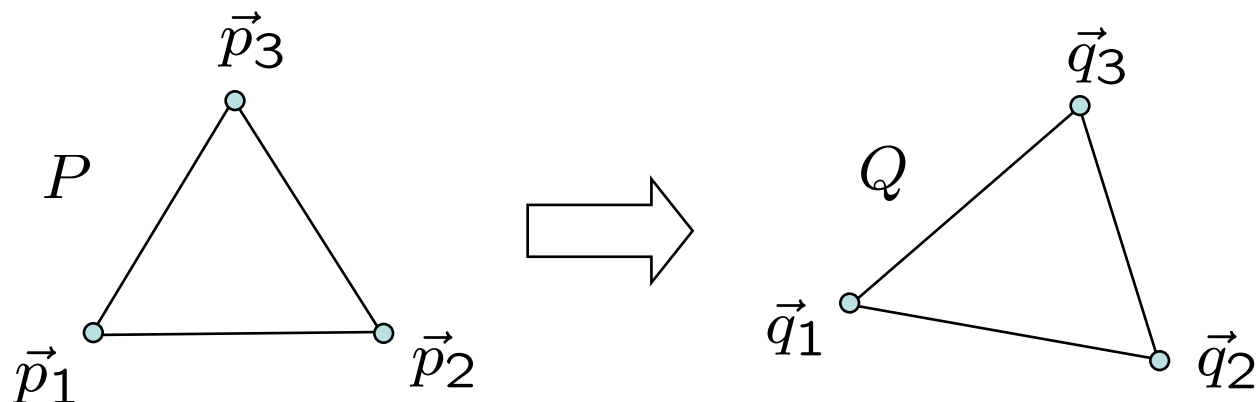
# Edge Splitting

the merged Delaunay triangulations are refined by **edge splitting** until **all edge lengths are bounded**

**edge splitting**



# Least-Distorting Triangle-to-Triangle Morphing



affine transformation

$$A\vec{p}_i + \vec{l} = \begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix} \vec{p}_i + \begin{pmatrix} l_x \\ l_y \end{pmatrix} = \vec{q}_i, \quad i \in \{1, 2, 3\}$$

ignore translation

$$V(t) = (\vec{v}_1(t), \vec{v}_2(t), \vec{v}_3(t))$$

$$V(t) = A(t)P$$



## How to Define $A(t)$ Reasonably?

- › The transformation should be symmetric with respect to  $t$
- › The rotational angle and scale should change linearly
- › The triangle should keep its orientation (not be reflected)
- › The resulting vertices' paths should be **simple**

# Singular Value Decomposition (SVD)

$$A\vec{p}_i + \vec{l} = \underbrace{\begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix}} \vec{p}_i + \begin{pmatrix} l_x \\ l_y \end{pmatrix} = \vec{q}_i, \quad i \in \{1, 2, 3\}$$

rotation matrix is orthogonal  $R^{-1} = R^T$

SVD

$$A = R_\alpha D R_\beta = R_\alpha \begin{pmatrix} s_x & 0 \\ 0 & s_y \end{pmatrix} R_\beta, \quad s_x, s_y > 0$$

$$\text{SVD: } A = U D V^T$$

$$\begin{aligned} [\mathbf{R}_a, \mathbf{D}, \mathbf{R}_b] &= \text{svd}(\mathbf{A}); \\ \mathbf{R}_b &= \mathbf{R}_b'; \end{aligned}$$

$$A^T A = (V D U^T)(U D V^T) = V D V^T$$

$$A A^T = (U D V^T)(V D U^T) = U D U^T$$

# Rotation + Scale, Shear

$$A = R_\alpha D R_\beta = R_\alpha (R_\beta R_\beta^T) D R_\beta = \\ (R_\alpha R_\beta) (R_\beta^T D R_\beta) = R_\gamma S = R_\gamma \begin{pmatrix} s_x & s_h \\ s_h & s_y \end{pmatrix}$$

$$[R_a, D, R_b] = \text{svd}(A);$$

$$R_b = R_b';$$

$$R_c = R_a * R_b;$$

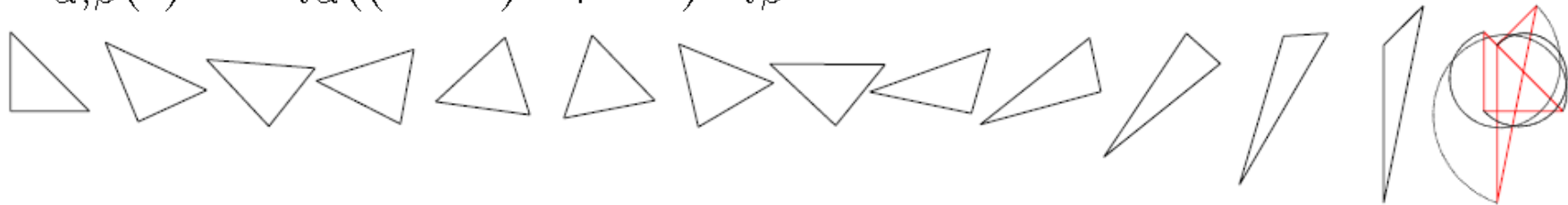
$$S = R_b' * D * R_b;$$

$$\text{theta} = \text{atan2}(-R_c(1,2), R_c(1,1));$$

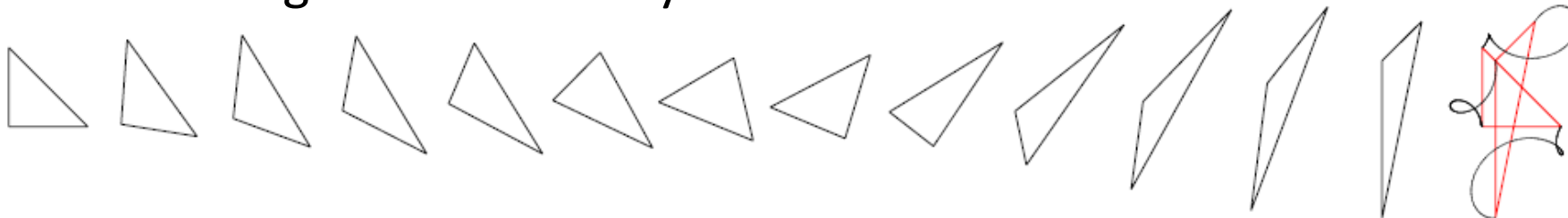
## linear interpolation



$$A_{\alpha,\beta}(t) = R_{t\alpha}((1-t)I + tD)R_{t\beta}$$



## rotation angle subtracted by $2\pi$



## preferred morphing



$$A_{\gamma}(t) = R_{t\gamma}((1-t)I + tS) \quad R_{t\gamma} = \begin{bmatrix} \cos(\theta t) & -\sin(\theta t) \\ \sin(\theta t) & \cos(\theta t) \end{bmatrix}$$

# Recover the Affine Transformation

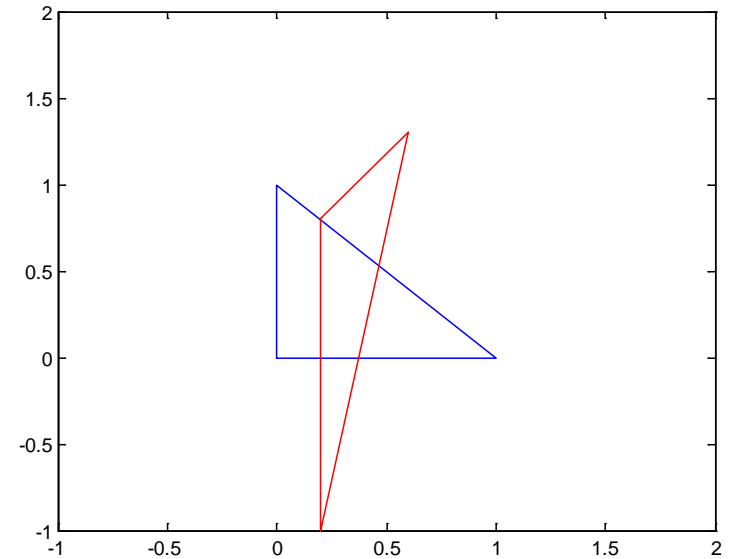
six unknowns

$$\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} a_1 & a_2 & l_x \\ a_3 & a_4 & l_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$$

three vertices, six equations

$$\begin{bmatrix} p_{1x} & p_{1y} & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & p_{1x} & p_{1y} & 1 \\ p_{2x} & p_{2y} & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & p_{2x} & p_{2y} & 1 \\ p_{3x} & p_{3y} & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & p_{3x} & p_{3y} & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ l_x \\ a_3 \\ a_4 \\ l_y \end{bmatrix} = \begin{bmatrix} q_{1x} \\ q_{1y} \\ q_{2x} \\ q_{2y} \\ q_{3x} \\ q_{3y} \end{bmatrix}$$





```
P = [0, 0; 1, 0; 0, 1];  
Q = [0.2, 0.8; 0.2, -1; 0.6, 1.3];  
figure; hold on  
plot([P(:,1); P(1,1)], ...  
      [P(:,2); P(1,2)], 'b');  
plot([Q(:,1); Q(1,1)], ...  
      [Q(:,2); Q(1,2)], 'r');  
axis([-1 2 -1 2])
```

```
M = zeros(6, 6);
for k = 1 : 3
    M(2*k-1, 1:2) = P(k, :);
    M(2*k-1, 3) = 1;
    M(2*k, 4:5) = P(k, :);
    M(2*k, 6) = 1;
end
b = reshape(Q', 6, 1);
a = M\b;
A = reshape(a, 3, 2)';
L = A(1:2, 3)'; % translation
A = A(1:2,1:2); % rotation+scaling+shearing
```

```

[Ra, D, Rb] = svd(A);
Rb = Rb';
Rc = Ra*Rb;
S = Rb'*D*Rb;
theta = atan2(-Rc(1,2), Rc(1,1));
figure;
for t = 0:0.1:1
    thetat = theta*t;
    Rct = [cos(thetat) -sin(thetat); ...
           sin(thetat) cos(thetat)];
    At = Rct*((1-t)*eye(2) + t*S);
    V = At*P';
    V = V';
    V = V + t*repmat(L, 3, 1);
    plot([V(:,1); V(1,1)], [V(:,2); V(1,2)], 'b')
    axis([-1 2 -1 2])
    pause(0.2)
end

```



# Closed-Form Vertex Paths for a Triangulation

$$\begin{array}{ll} \text{triangulation} & \text{triangles} \\ \mathcal{T} = \{T_{\{i,j,k\}}\} & P_{\{i,j,k\}} = (\vec{p}_i, \vec{p}_j, \vec{p}_k) \\ & Q_{\{i,j,k\}} = (\vec{q}_i, \vec{q}_j, \vec{q}_k) \end{array}$$

$$\begin{array}{l} A_{\{i,j,k\}}(t) \quad \Rightarrow \quad V(t) = (v_1(t), \dots, v_n(t)), t \in [0, 1] \\ V(0) = (\vec{p}_1, \dots, \vec{p}_n) \quad V(1) = (\vec{q}_1, \dots, \vec{q}_n) \end{array}$$

desired affine transformation

$$B_{\{i,j,k\}}(t)\vec{p}_f + \vec{l} = \vec{v}_f(t), \quad f \in \{i, j, k\}$$

$$\text{minimize } E_{V(t)} = \sum_{\{i,j,k\} \in \mathcal{T}} \left\| \underset{\substack{\uparrow \\ \text{actual}}}{A_{\{i,j,k\}}(t)} - B_{\{i,j,k\}}(t) \right\|^2$$

$\|\cdot\|$  is the Frobenius norm

# Matrix Form

$$\text{minimize } E_{V(t)} = \sum_{\{i,j,k\} \in \mathcal{T}} \left\| \underset{\substack{\uparrow \\ \text{actual}}}{A_{\{i,j,k\}}(t)} - B_{\{i,j,k\}}(t) \right\|^2$$

$\|\cdot\|$  is the Frobenius norm

$$u^T = (1, v_{2_x}(t), v_{2_y}(t), \dots, v_{n_x}(t), v_{n_y}(t))$$

$$E_{V(t)} = u^T \begin{pmatrix} c & G^T \\ G & H \end{pmatrix} u \quad \begin{array}{l} G \in \mathbb{R}^{2n \times 1} \\ H \in \mathbb{R}^{2n \times 2n} \end{array}$$

$$H \begin{pmatrix} v_{2_x}(t) \\ v_{2_y}(t) \\ \vdots \end{pmatrix} = -G \quad \text{can be solved by "\"}$$

$$V(t) = -H^{-1}G(t)$$

# Good Properties

- › For a given  $t$ , the solution is unique
- › The solution requires only **one matrix inversion** for a specific source and target shape. Every intermediate shape is found by multiplying the inverted matrix by a vector  $G(t)$
- › The vertex path is infinitely smooth, starts exactly in the source shape, and ends exactly in the target shape

# Symmetric Solutions

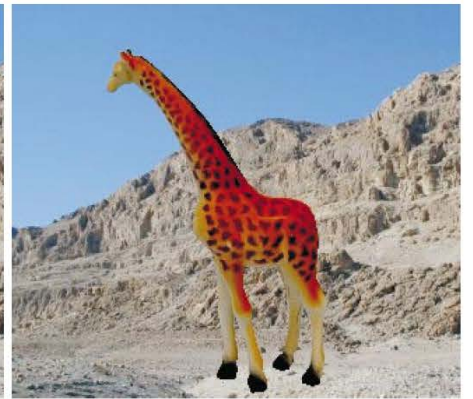
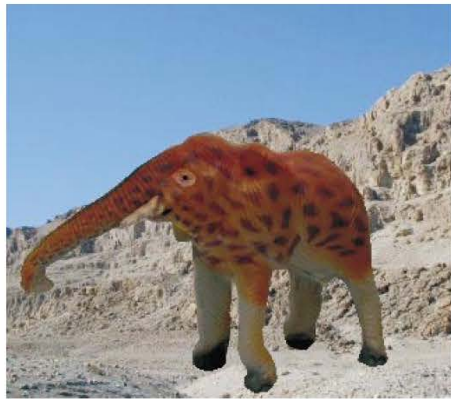
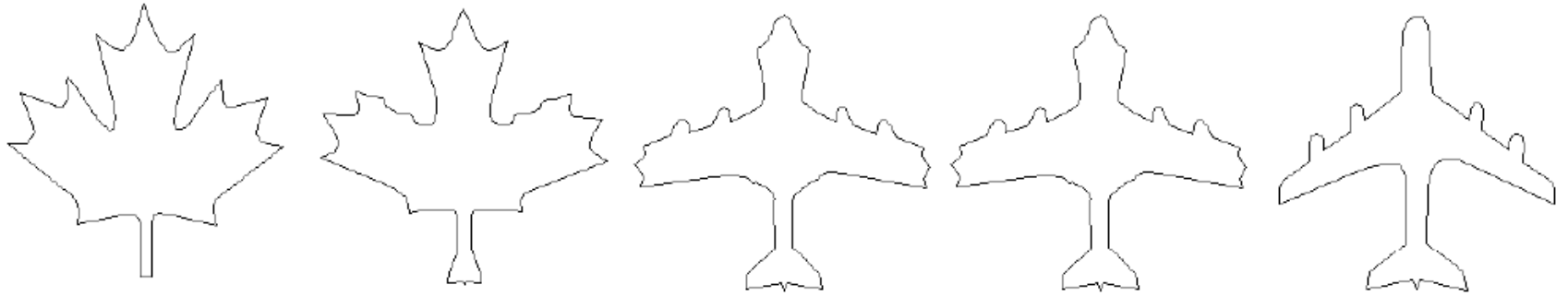
- › Consider both directions

$$E_{V(t)} = (1 - t)E_{V(t)}^{\rightarrow} + tE_{V(t)}^{\leftarrow}$$

$$E_{V(t)}^{\rightarrow} = \sum_{f \in \text{Tri}} \|A_f^{\rightarrow}(t) - B_f^{\rightarrow}(t)\|^2$$

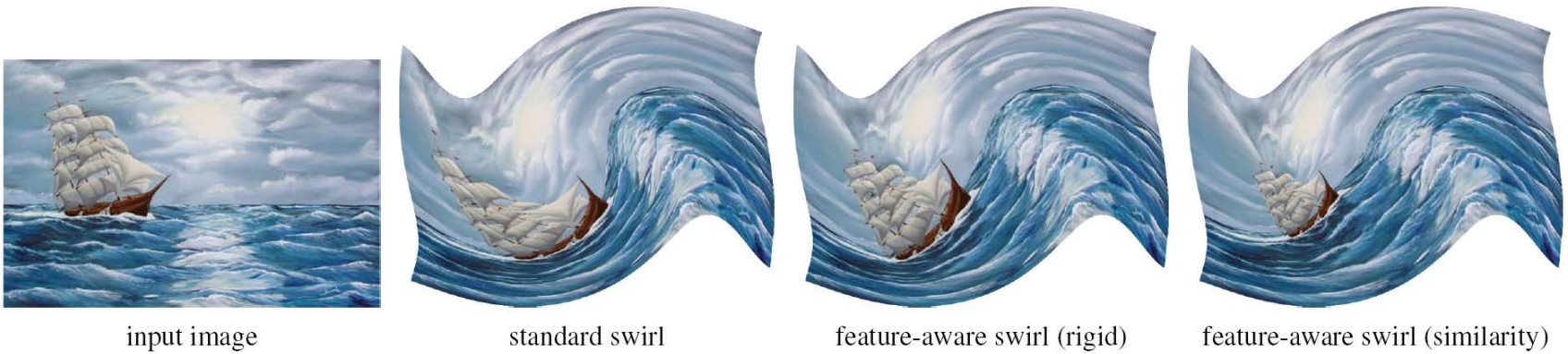
$$E_{V(t)}^{\leftarrow} = \sum_{f \in \text{Tri}} \|A_f^{\leftarrow}(1 - t) - B_f^{\leftarrow}(1 - t)\|^2$$

# Results

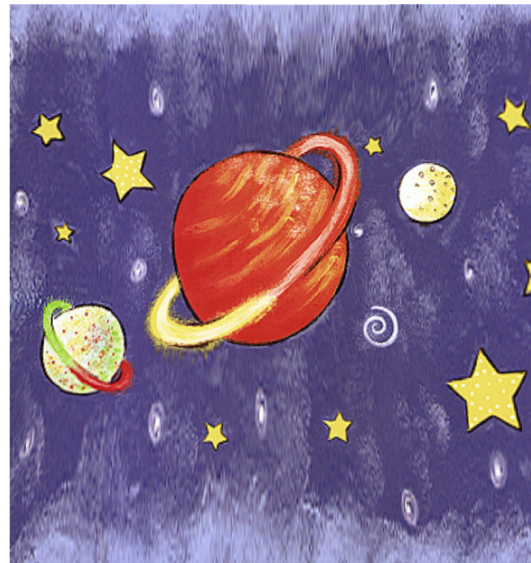


# Feature-Aware Warping

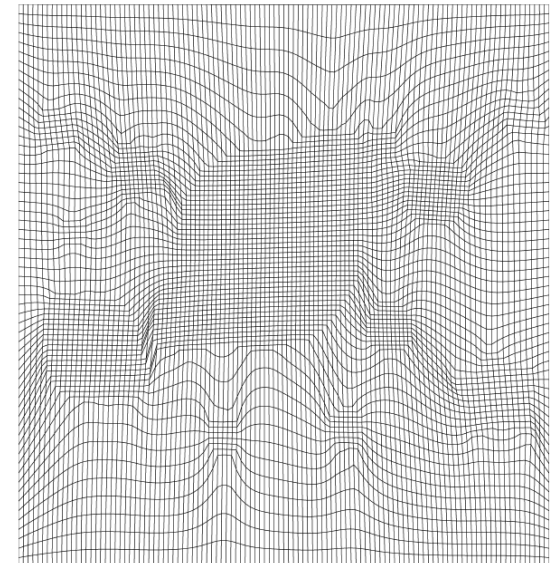
[Gal, Sorkine, and Cohen-Or]



input image and its feature mask



vertical stretch  $\times 2$

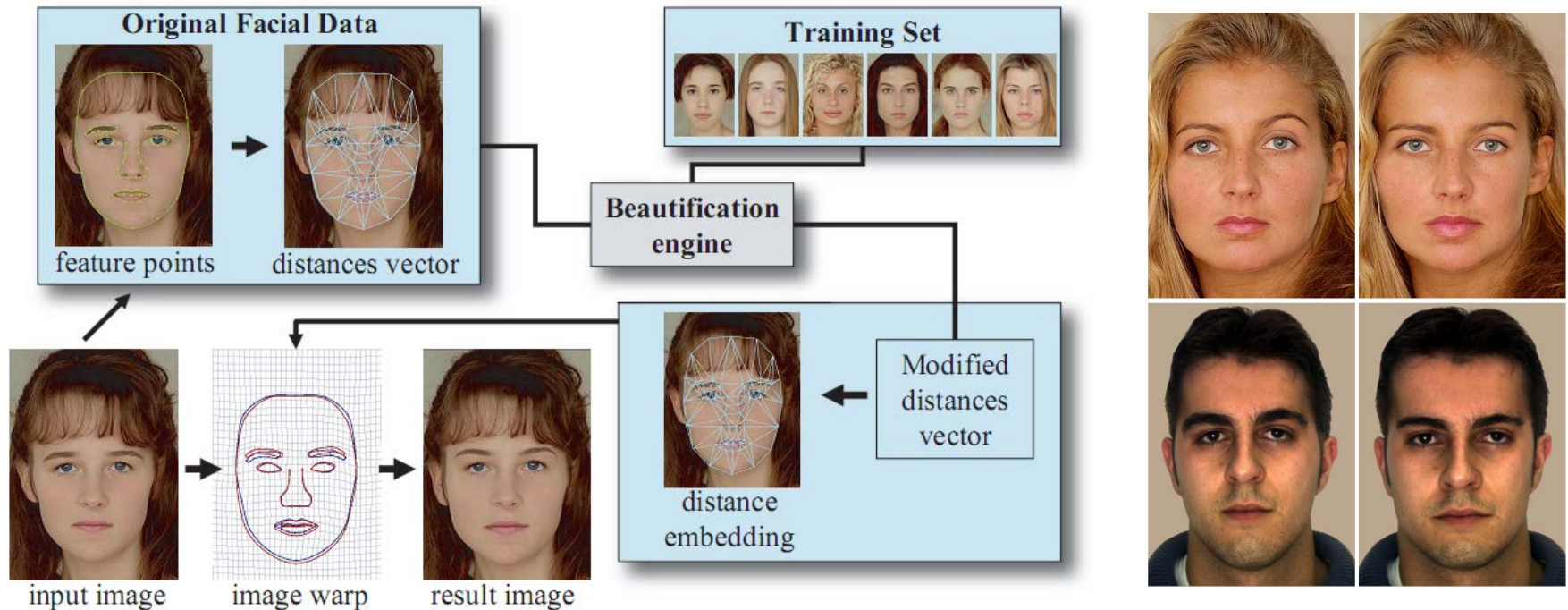


underlying grid

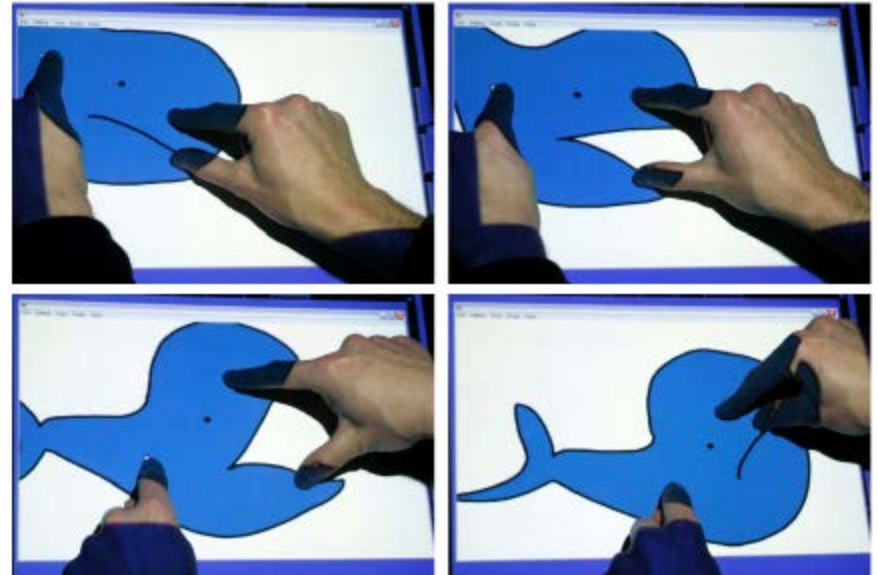


# Data-Driven Enhancement of Facial Attractiveness

- › Leyvand et al., SIGGRAPH 2008
- › <http://www.youtube.com/watch?v=IVbrUuwK-8g>



# As-Rigid-as-Possible Shape Manipulation



## › Demo Java Applets

- › <http://www-ui.is.s.u-tokyo.ac.jp/~takeo/research/rigid/index.html>



# Regenerative Morphing

- › Shechtman, Rav-Acha, Irani, and Seitz
  - › CVPR 2010
- › Bidirectional similarity
- › PatchMatch



# Being John Malkovich

- › Kemelmacher-Shlizerman, Sankar, Shechtman, and Seitz
  - › ECCV 2010
- › Puppeteering a celebrity
- › Video or photo collection of celebrity
- › Image based
- › Local Binary Pattern (LBP)

