## Image Morphing



Computational Photography

## Project 3: Gradient Domain Editing

- General concept: Solve for pixels of new image that satisfy constraints on the gradient and the intensity
- Constraints can be from one image (for filtering) or more (for blending)


## 2D image transformations



| Name | Matrix | \# D.O.F. | Preserves: | Icon |
| :--- | :---: | :---: | :--- | :---: |
| translation | $[\boldsymbol{I} \mid \boldsymbol{t}]_{2 \times 3}$ | 2 | orientation $+\cdots$ | $\square$ |
| rigid (Euclidean) | $[\boldsymbol{R} \mid \boldsymbol{t}]_{2 \times 3}$ | 3 | lengths $+\cdots$ | $\square$ |
| similarity | $[s \boldsymbol{R} \mid \boldsymbol{t}]_{2 \times 3}$ | 4 | angles $+\cdots$ | $\square$ |
| affine | $[\boldsymbol{A}]_{2 \times 3}$ | 6 | parallelism $+\cdots$ | $\square$ |
| projective | $[\tilde{\boldsymbol{H}}]_{3 \times 3}$ | 8 | straight lines | $\square$ |

These transformations are a nested set of groups

- Closed under composition and inverse is a member


## Take-home Question

Suppose we have two triangles: $A B C$ and $A^{\prime} B^{\prime} C^{\prime}$. What transformation will map $A$ to $A^{\prime}$, $B$ to $\mathrm{B}^{\prime}$, and C to $\mathrm{C}^{\prime}$ ? How can we get the parameters?


## Today: Morphing

Women in art

http://youtube.com/watch?v=nUDIoN- Hxs
Aging


## Texturing in transformed coordinates



Given a coordinate transform $\left(x^{\prime}, y^{\prime}\right)=T(x, y)$ and a source image $f(x, y)$, how do we compute a transformed image $g\left(x^{\prime}, y^{\prime}\right)=f(T(x, y))$ ?

## Forward mapping



Send each pixel $f(x, y)$ to its corresponding location $\left(x^{\prime}, y^{\prime}\right)=T(x, y)$ in the second image

## Forward mapping

What is the problem with this approach?


Send each pixel $f(x, y)$ to its corresponding location

$$
\left(x^{\prime}, y^{\prime}\right)=T(x, y) \text { in the second image }
$$

Q: what if pixel lands "between" two pixels?
A: distribute color among neighboring pixels ( $x^{\prime}, y^{\prime}$ )

- Known as "splatting"


## Inverse mapping



Get each pixel $g\left(x^{\prime}, y^{\prime}\right)$ from its corresponding location

$$
(x, y)=T^{-1}\left(x^{\prime}, y^{\prime}\right) \text { in the first image }
$$

Q: what if pixel comes from "between" two pixels?

## Inverse mapping



Get each pixel $g\left(x^{\prime}, y^{\prime}\right)$ from its corresponding location $(x, y)=T^{-1}\left(x^{\prime}, y^{\prime}\right)$ in the first image
Q: what if pixel comes from "between" two pixels?
A: Interpolate color value from neighbors

- nearest neighbor, bilinear, Gaussian, bicubic
- E.g. interpolate.interp2 or ndimage.map_coordinates in Python scipy


## Bilinear Interpolation



## Forward vs. inverse mapping

Q: which is better?

A: Usually inverse—eliminates holes

- however, it requires an invertible warp function


## Morphing = Object Averaging



The aim is to find "an average" between two objects

- Not an average of two images of objects...
- ...but an image of the average object!
- How can we make a smooth transition in time?
- Do a "weighted average" over time t


## Averaging Points

What's the average of $P$ and $Q$ ?

Linear Interpolation
New point: $P+t^{*}(Q-P)$
Or equivalently: $(1-t) P+t Q$ $0<t<1$


Extrapolation: $\mathrm{t}<0$ or $\mathrm{t}>1$
$P+1.5 \mathrm{v}$
$=P+1.5(Q-P)$

$$
P+0.5 v \quad=-0.5 P+1.5 Q(t=1.5)
$$

$P$ and $Q$ can be anything:

- points on a plane (2D) or in space (3D)
- Colors in RGB (3D)
- Whole images (m-by-n D)... etc.


## Idea \#1: Cross-Dissolve



Interpolate whole images:
Image $_{\text {halfway }}=(1-t)^{*}$ Image $_{1}+\mathrm{t}^{*}$ image $_{2}$
This is called cross-dissolve in film industry

But what if the images are not aligned?

Idea \#2: Align, then cross-disolve


Align first, then cross-dissolve

- Alignment using global warp - picture still valid


## Dog Averaging



## What to do?

- Cross-dissolve doesn't work
- Global alignment doesn't work
- Cannot be done with a global transformation (e.g. affine)
- Any ideas?

Feature matching!

- Nose to nose, tail to tail, etc.
- This is a local (non-parametric) warp

Idea \#3: Local warp, then cross-dissolve


Morphing procedure For every frame $t$,

1. Find the average shape (the "mean dog" ${ }^{\oplus}$ )

- local warping

2. Find the average color

- Cross-dissolve the warped images


## Local (non-parametric) Image Warping



Need to specify a more detailed warp function

- Global warps were functions of a few $(2,4,8)$ parameters
- Non-parametric warps $u(x, y)$ and $v(x, y)$ can be defined independently for every single location $x, y$ !
- Once we know vector field u,v we can easily warp each pixel (use backward warping with interpolation)

Image Warping - non-parametric
Move control points to specify a spline warp
Spline produces a smooth vector field


## Warp specification - dense

How can we specify the warp?
Specify corresponding spline control points

- interpolate to a complete warping function


But we want to specify only a few points, not a grid

## Warp specification - sparse

How can we specify the warp?
Specify corresponding points

- interpolate to a complete warping function
- How do we do it?


How do we go from feature points to pixels?

## Triangular Mesh



1. Input correspondences at key feature points
2. Define a triangular mesh over the points

- Same mesh (triangulation) in both images!
- Now we have triangle-to-triangle correspondences

3. Warp each triangle separately from source to destination

- Affine warp with three corresponding points (just like takehome question)


## Triangulations

A triangulation of set of points in the plane is a partition of the convex hull to triangles whose vertices are the points, and do not contain other points.

There are an exponential number of triangulations of a point set.


An $\mathrm{O}\left(n^{3}\right)$ Triangulation Algorithm
Repeat until impossible:

- Select two sites.
- If the edge connecting them does not intersect previous edges, keep it.



## "Quality" Triangulations

Let $\alpha\left(T_{i}\right)=\left(\alpha_{i 1}, \alpha_{i 2}, \alpha_{i 3}, \ldots\right)$ be the vector of angles in the triangulation $T$ in increasing order:

- A triangulation $T_{1}$ is "better" than $T_{2}$ if the smallest angle of $T_{1}$ is larger than the smallest angle of $T_{2}$
- Delaunay triangulation is the "best" (maximizes the smallest angles)

good

bad


## Improving a Triangulation

In any convex quadrangle, an edge flip is possible. If this flip improves the triangulation locally, it also improves the global triangulation.


If an edge flip improves the triangulation, the first edge is called "illegal".

## Naïve Delaunay Algorithm

Start with an arbitrary triangulation. Flip any illegal edge until no more exist.

Could take a long time to terminate.


## Delaunay Triangulation by Duality

Draw the dual to the Voronoi diagram by connecting each two neighboring sites in the Voronoi diagram.

- The DT may be constructed in $\mathrm{O}(n \log n)$ time



## Image Morphing

How do we create a morphing sequence?

1. Create an intermediate shape (by interpolation)
2. Warp both images towards it
3. Cross-dissolve the colors in the newly warped images


## Warp interpolation

How do we create an intermediate shape at time t?

- Assume t = [0,1]
- Simple linear interpolation of each feature pair
- (1-t)*p1+t*p0 for corresponding features p0 and p1



## Morphing \& matting

Extract foreground first to avoid artifacts in the background

(c) $\alpha=0.0$


(d) $\alpha=0.2$

(g) $\alpha=0.8$
(h) $\alpha=1.0$

## Dynamic Scene



Black or White (MJ):
http://www.youtube.com/watch?v=R4kLKv5gtxc

Willow morph: http://www.youtube.com/watch?v=uLUyuWo3pG0

## Summary of morphing

1. Define corresponding points
2. Define triangulation on points

- Use same triangulation for both images

3. For each $t=0:$ step:1
a. Compute the average shape (weighted average of points)
b. For each triangle in the average shape

- Get the affine projection to the corresponding triangles in each image
- For each pixel in the triangle, find the corresponding points in each image and set value to weighted average (optionally use interpolation)
c. Save the image as the next frame of the sequence


## Next classes

Pinhole camera: start of perspective geometry

Single-view metrology: measure 3D distances from an image

