

Image Compositing and Morphing COS 426, Fall 2022

PRINCETON UNIVERSITY

Digital Image Processing

- Changing pixel values
 - Linear: scale, offset, etc.
 - Nonlinear: gamma, saturation, etc.
 - Histogram equalization
- Filtering over neighborhoods
 - Blur & sharpen
 - Detect edges
 - Median
 - Bilateral filter

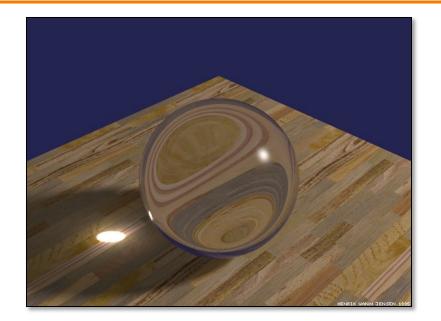
- Moving image locations
 - Scale
 - Rotate
 - Warp
- Combining images
 - Composite
 - Morph
- Quantization
- Spatial / intensity tradeoff
 Dithering

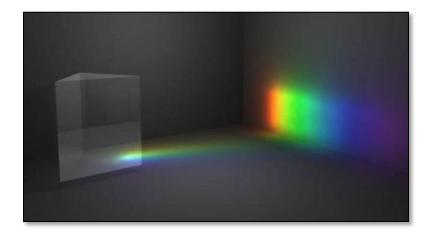


Types of Transparency

- Refraction
 - Light is bent as it goes through an object
 - Can focus light: caustics
 - Can be color-dependent: dispersion

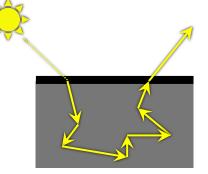






Types of Transparency

- Refraction
- Subsurface scattering
 - Light leaves at different position than it entered
 - Translucent materials









Types of Transparency

- Refraction
- Subsurface scattering
- Today: compositing

 Nonrefractive (partial) transparency
 - Separate image into layers with known order
 - *Pixelwise* combination: each pixel in each layer can be transparent, opaque, or somewhere in between





Smith & Blinn`84







Jurassic Park (1993)

Image Composition

- Issues:
 - Segmenting image into regions
 - Blending into single image seamlessly

Image Composition

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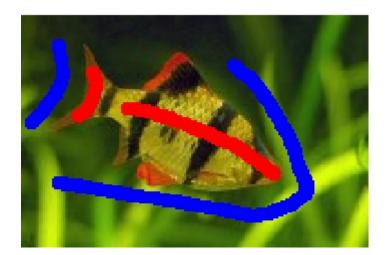
- Chroma keying (blue- or green-screen)
 - Photograph object in front of screen with known color

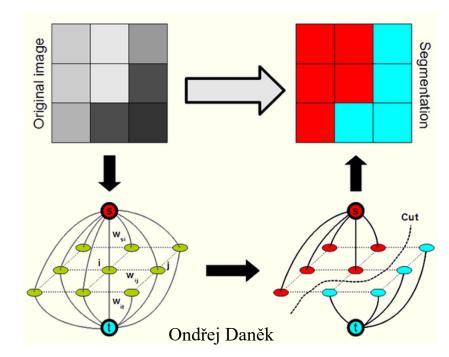


Rosco Spectrum



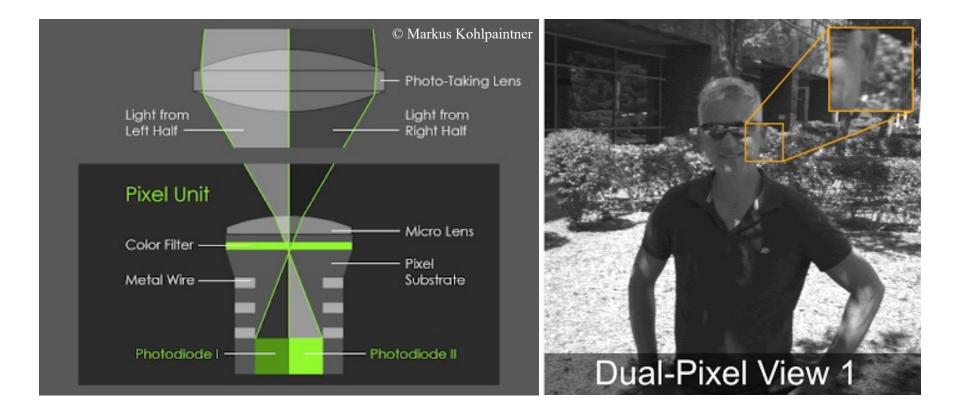
- Specify segmentation by hand
 - Purely manual: draw matte every frame
 - Semi-automatic: graph-cut (draw a few strokes)
 Implemented using min-cut algorithm: separate regions along minimal cuts (where edges measure differences between adjacent pixels)







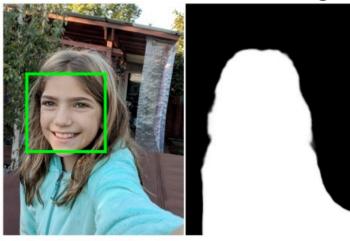
• Portrait mode in Google Pixel Phone

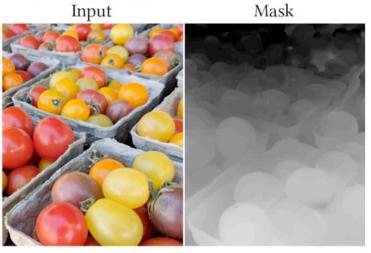


Wadhwa et al., 2018



Portrait mode blur in Google Pixel Phones





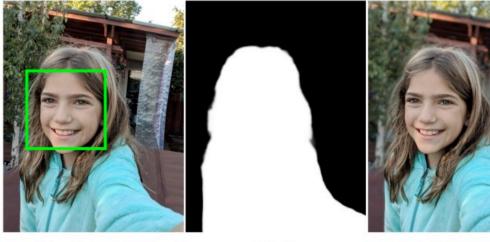
Disparity

Input

Wadhwa et al., 2018



Portrait mode blur in Google Pixel Phones



Input

Mask

Output





Wadhwa et al., 2018

Input

Disparity

Output

Image Composition



• Issues:

- Segmenting image into regions
- Blending into single image seamlessly

Image Blending

- Ingredients
 - Background image
 - Foreground image with blue background
- Method
 - Non-blue foreground pixels overwrite background







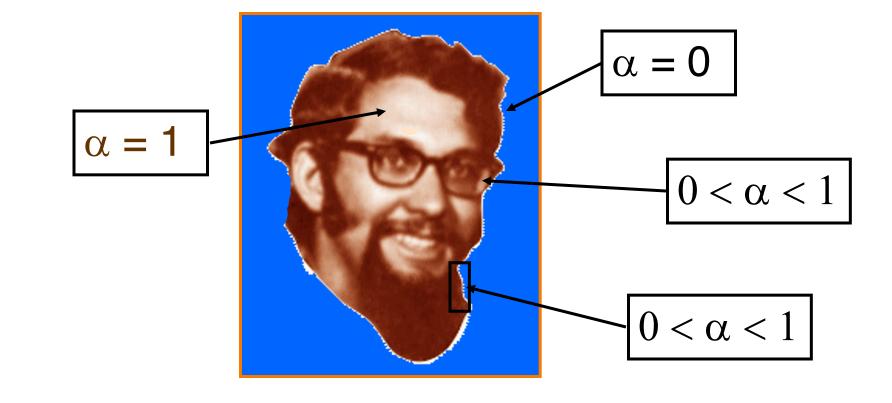
Blending with Alpha Channel

- Per-pixel "alpha" channel
 - Controls the linear interpolation between foreground and background pixels when elements are composited



Blending with Alpha Channel

- Per-pixel "alpha" channel
 - Controls the linear interpolation between foreground and background pixels when elements are composited

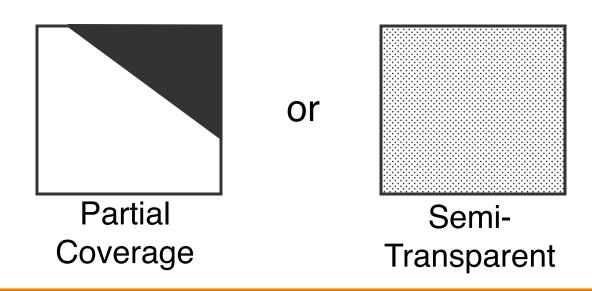


Alpha Channel

0



- Encodes pixel coverage information
 - $\alpha = 0$: no coverage (or transparent)
 - $\alpha = 1$: full coverage (or opaque)
 - $0 < \alpha < 1$: partial coverage (or semi-transparent)
- Example: $\alpha = 0.3$

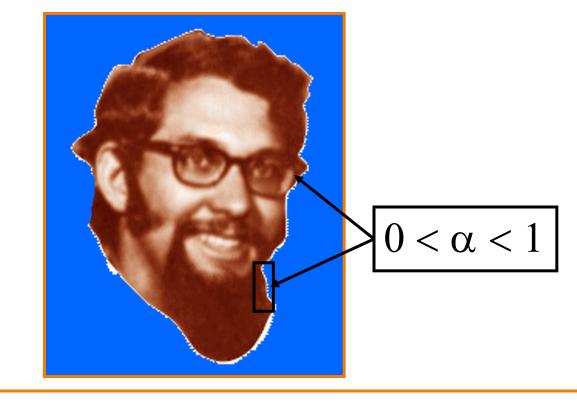


Alpha Blending: "Over" Operator



- If background B is opaque:
 - C = A over B
 - $C = \alpha_A A + (1 \alpha_A) B$

- If background B has its own α :
 - C = A over B
 - $C = \alpha_A A + (1 \alpha_A) \alpha_B B$
 - $\circ \ \alpha_{\rm C} = \alpha_{\rm A} + (1 \alpha_{\rm A})\alpha_{\rm B}$



• Suppose we put A over B over background G

• How much of B is blocked by A?

 α_A

B

G

• Suppose we put A over B over background G

```
• How much of B is blocked by A?
```

 α_A

B

G

• How much of B shows through A?

 $(1-\alpha_A)$



• Suppose we put A over B over background G

• How much of B is blocked by A?

 $\boldsymbol{\alpha}_{A}$

В

G

• How much of B shows through A?

$$(1-\alpha_A)$$

 $\circ\,$ How much of G shows through both A and B? $(1{-}\alpha_{\text{A}})(1{-}\alpha_{\text{B}})$

• Suppose we put A over B over background G



$$\alpha_A A + (1 - \alpha_A) \alpha_B B + (1 - \alpha_A)(1 - \alpha_B) G$$

Β

G

$$= \alpha_{A}A + (1 - \alpha_{A}) \left[\alpha_{B}B + (1 - \alpha_{B})G\right]$$

= A over [B over G]

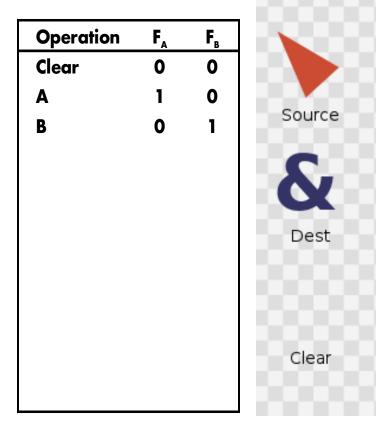
Must perform "over" back-to-front: right associative!





Composition algebra – 12 combinations

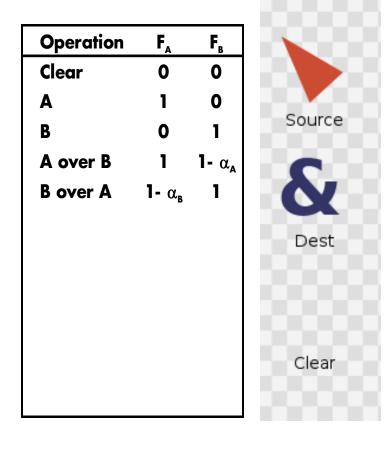
 $C' = F_A \alpha_A A + F_B \alpha_B B$





Composition algebra – 12 combinations

 $C' = F_A \alpha_A A + F_B \alpha_B B$

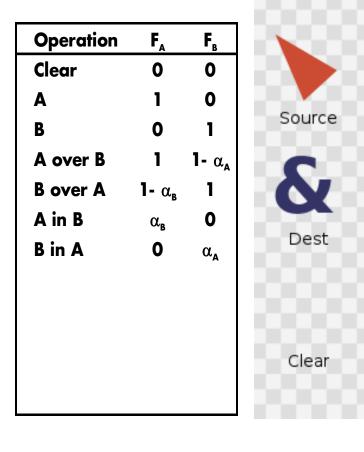






Composition algebra – 12 combinations

 $C' = F_A \alpha_A A + F_B \alpha_B B$

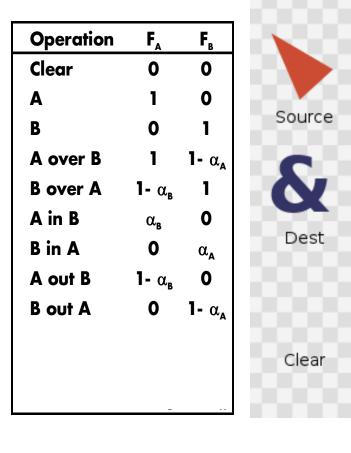


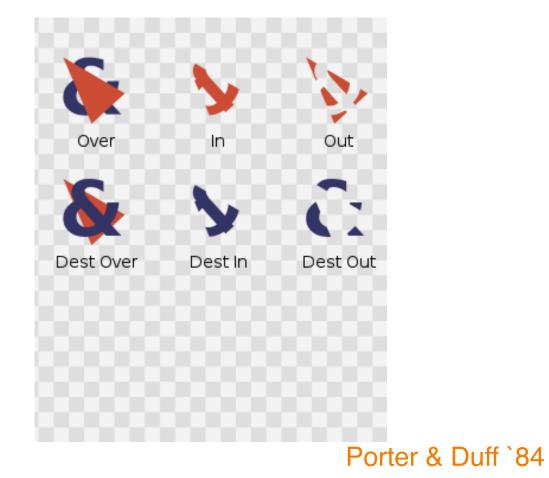




Composition algebra – 12 combinations

 $C' = F_A \alpha_A A + F_B \alpha_B B$

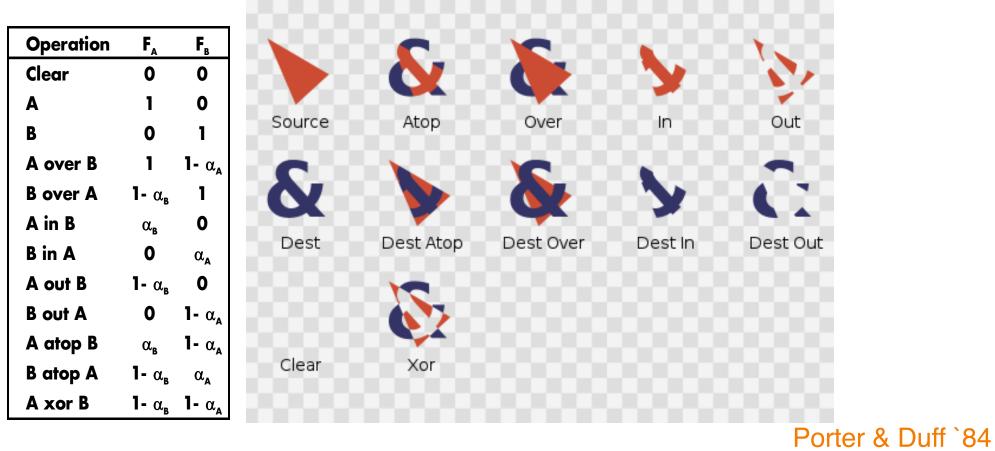




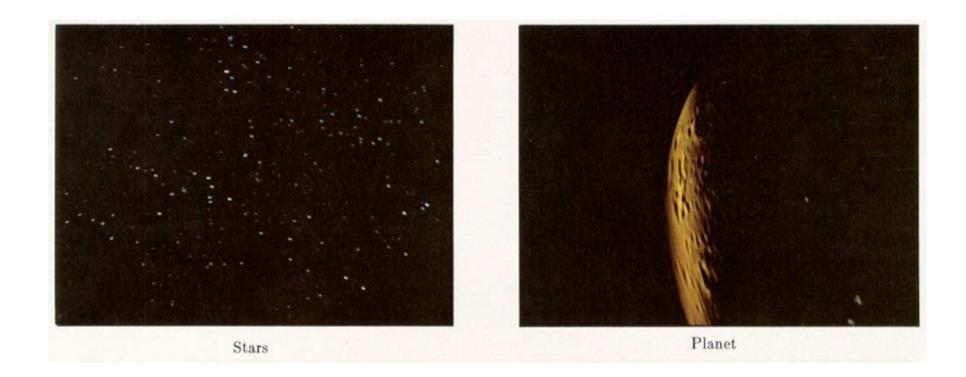


Composition algebra – 12 combinations

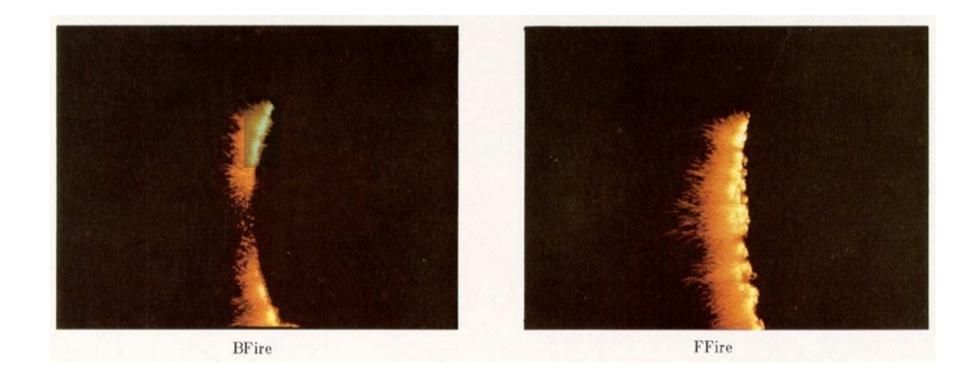
 $C' = F_A \alpha_A A + F_B \alpha_B B$



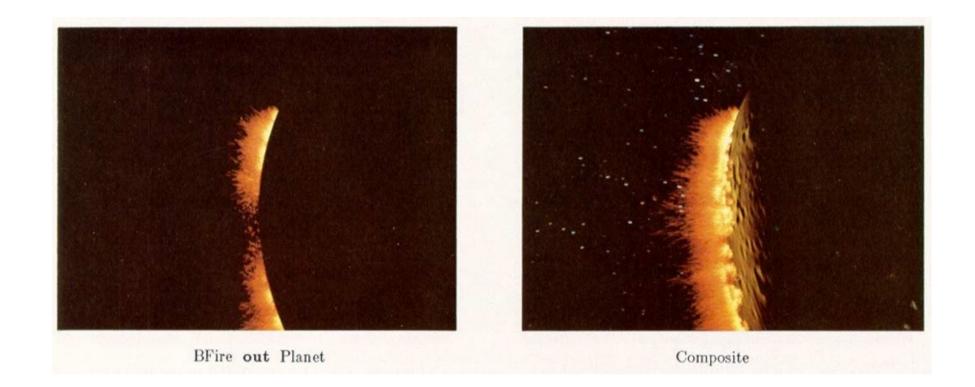




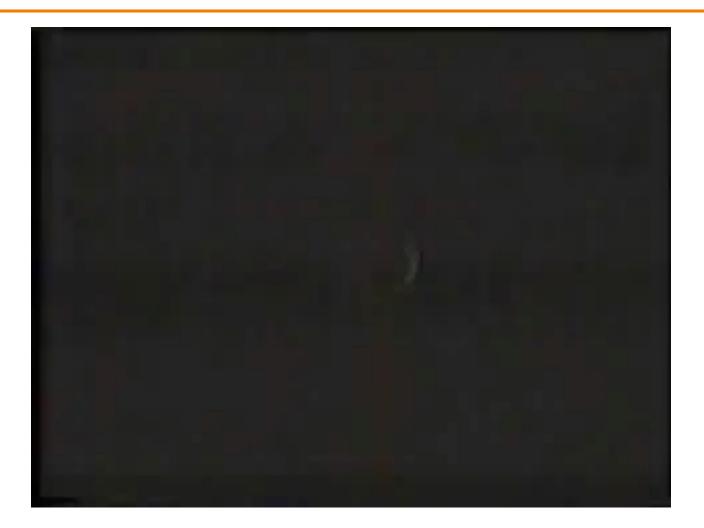








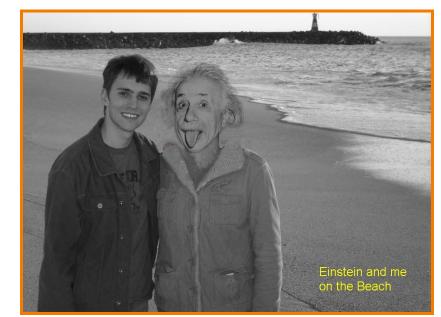




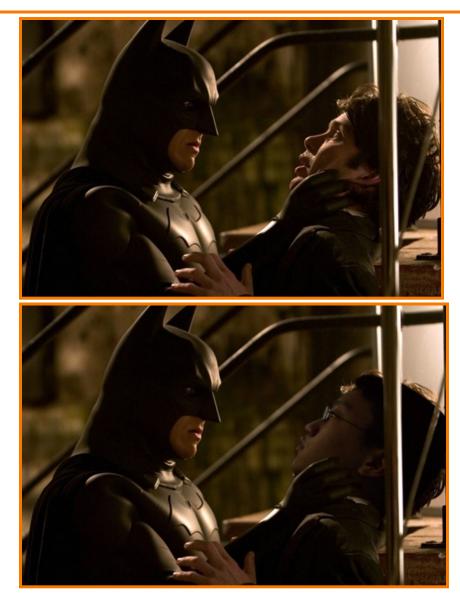
"Genesis" sequence from Star Trek II: The Wrath of Khan

COS426 Examples





Darin Sleiter

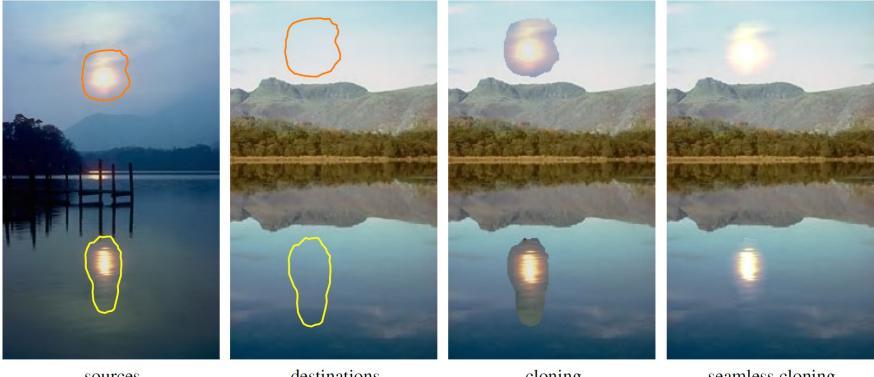


Kenrick Kin

Poisson Image Blending

Beyond simple compositing

• Solve for image samples that follow gradients of source subject to boundary conditions imposed by dest



sources

destinations



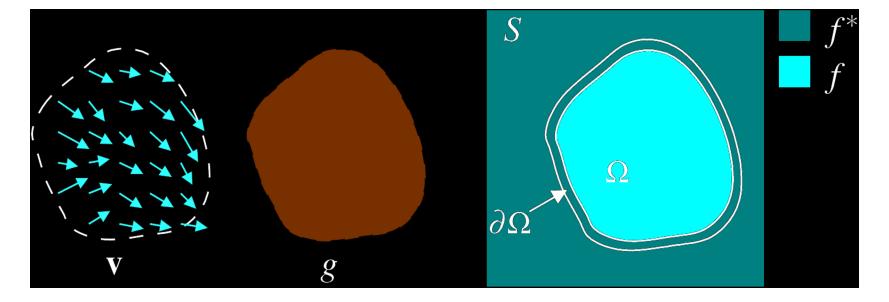




Poisson Image Blending

Beyond simple compositing

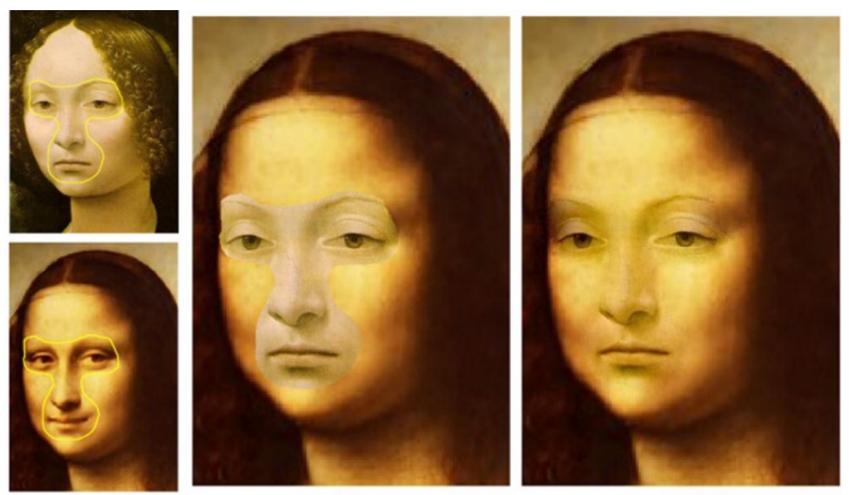
 Solve for image samples that follow gradients of source subject to boundary conditions imposed by dest



$$\min_{f} \iint_{\Omega} |\nabla f - \mathbf{v}|^2 \text{ with } f|_{\partial \Omega} = f^*|_{\partial \Omega}$$

Poisson Image Blending





source/destination

cloning

seamless cloning

Digital Image Processing

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Image Morphing



Animate transition between two images

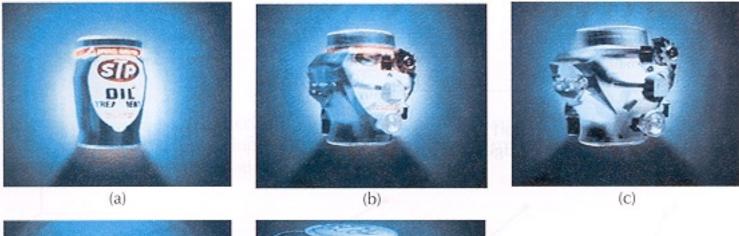






Figure 16-9 Transformation of an STP oil ca into an engine block. (Courtesy of Silicon Graphics, Inc.)

H&B Figure 16.9

Cross-Dissolving

- Blend images with "over" operator
 - alpha of bottom image is 1.0
 - $\circ~$ alpha of top image varies from 1.0 to 0.0 $\,$

blend(i,j) = $(1-t) \operatorname{src}(i,j) + t \operatorname{dst}(i,j)$ ($0 \le t \le 1$)

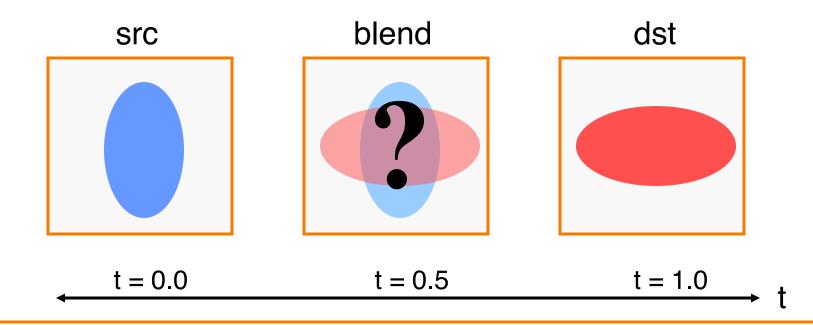
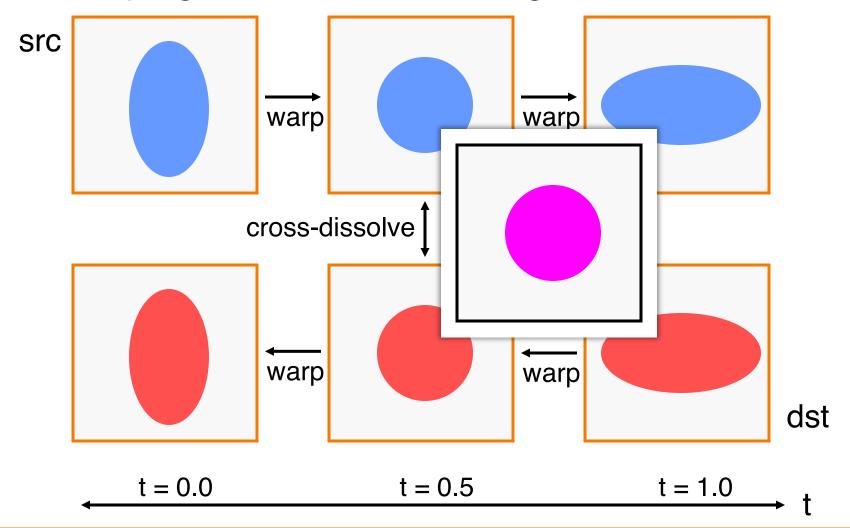


Image Morphing



Combines warping and cross-dissolving



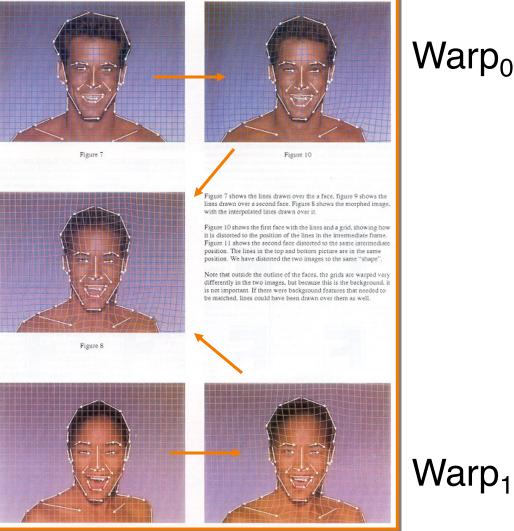
Beier & Neeley Example



Image₀

Result

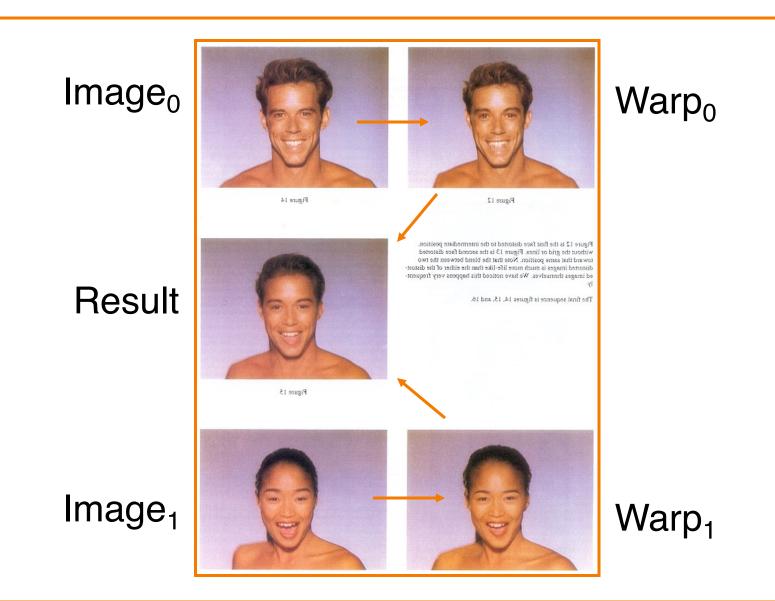
Image₁



Warp₀

Beier & Neeley Example



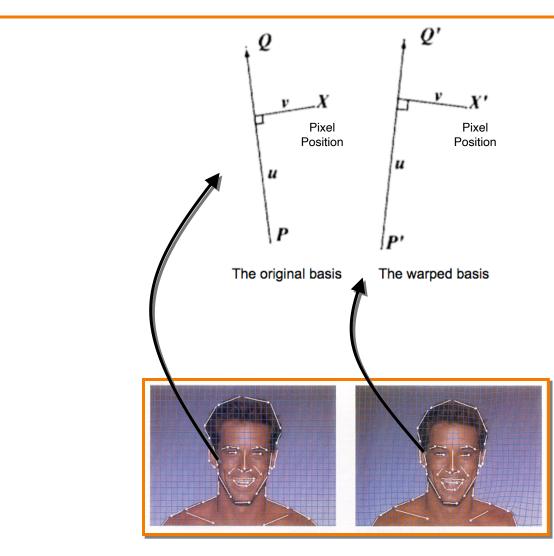


Beier & Neeley Example

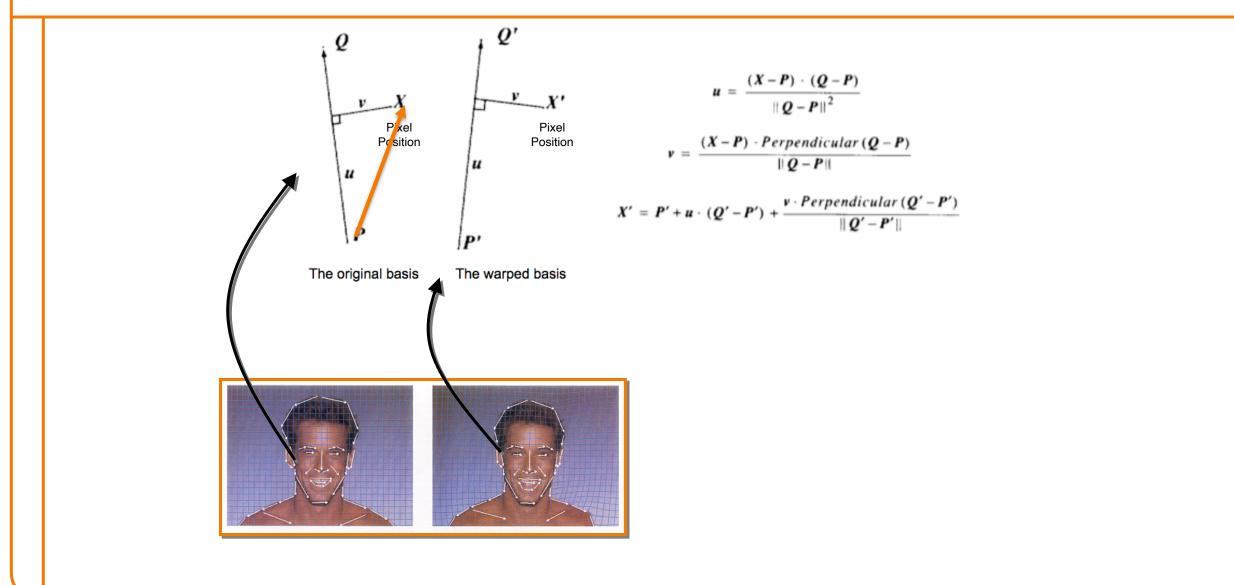


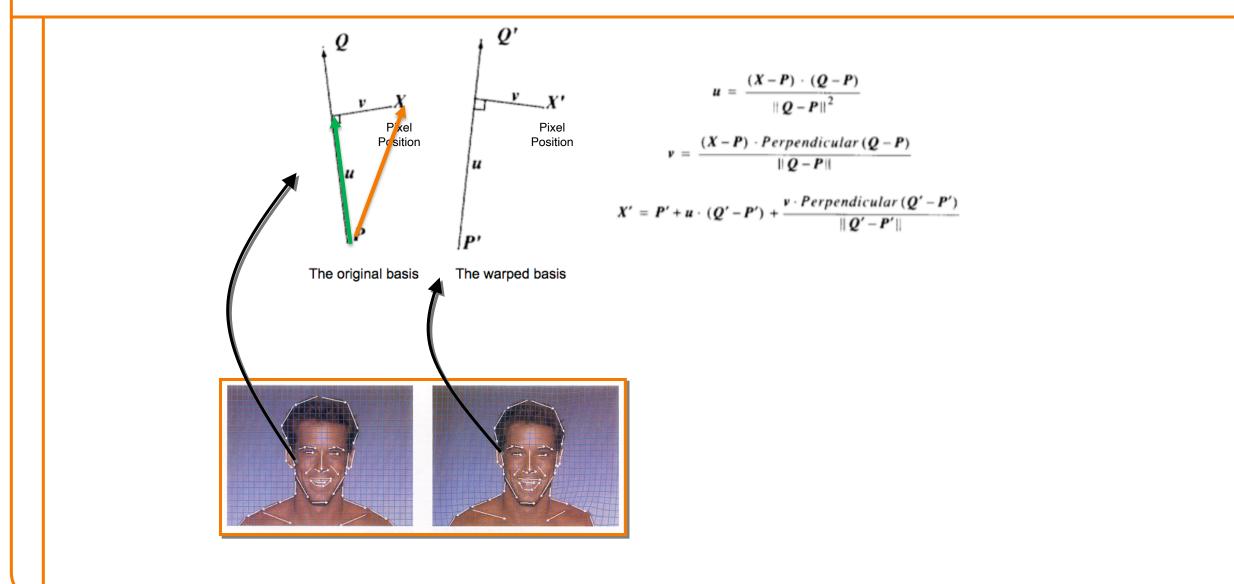


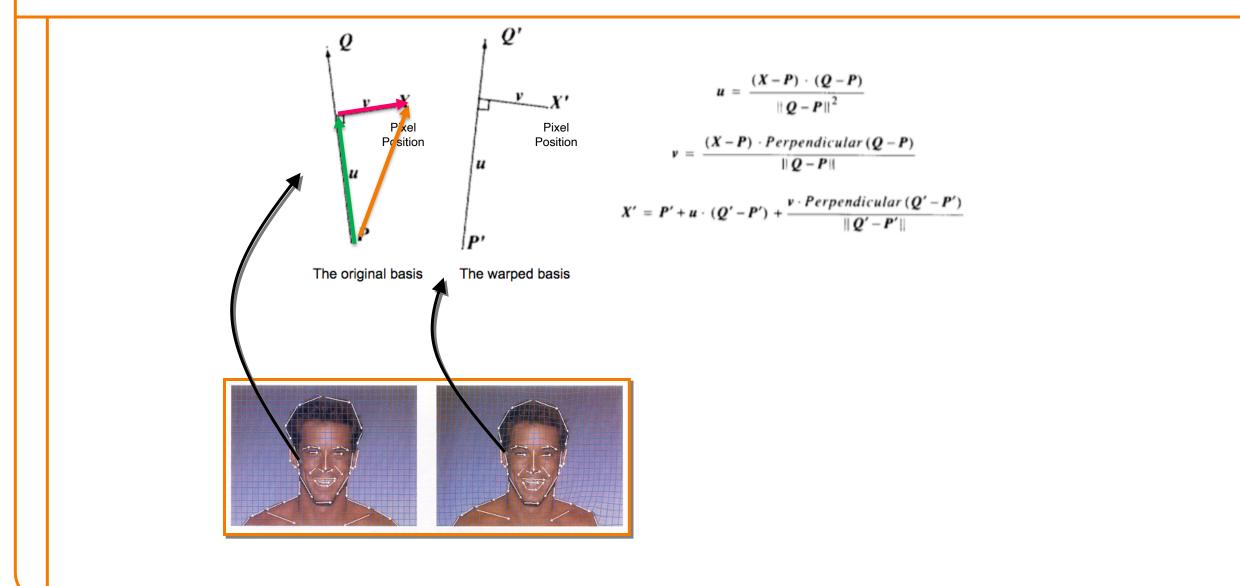
Black or White, Michael Jackson (1991)



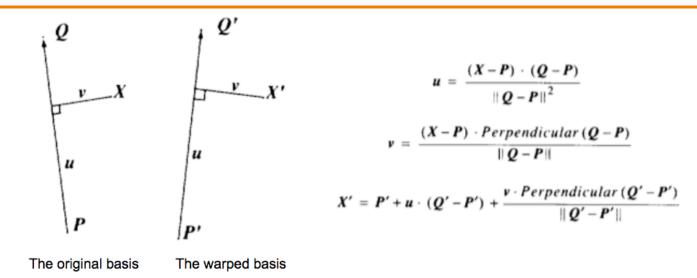












This generates one warp per line, each of which is a simple rotation and non-uniform scale (scaling is only done along the axis of the line). These warps must then be averaged to get the final warp. In the original paper, the weights for the average are tuned with the formula below. The *dist* variable is the distance of the point from the line segment, and the *length* variable is the length of the line segment.

$$weight = \left(\frac{length^p}{a+dist}\right)^b$$

The equations give several parameters to tune, and I got the best results when a = 0.001, b = 2, and p = 0. Ignoring the length of the line segments (by setting p to zero) gave better results than when the length was taken in to account. I used seven contours with 28 line segments to represent the features of each face. Nice implementation notes from Evan Wallace, Brown University

Nice implementation notes from Evan Wallace, Brown University http://cs.brown.edu/courses/csci1950-g/results/proj5/edwallac/

Warping Pseudocode



```
WarpImage(Image, L<sub>src</sub>[...], L<sub>dst</sub>[...])
begin
    foreach destination pixel p<sub>dst</sub> do
         psum = (0,0)
         wsum = 0
         foreach line L<sub>dst</sub>[i] do
             p_{src}[i] = p_{dst} transformed by (L_{dst}[i], L_{src}[i])
             psum = psum + p_{src}[i] * weight[i]
             wsum += weight[i]
         end
         p_{src} = psum / wsum
         \text{Result}(p_{dst}) = \text{Resample}(p_{src})
    end
end
```

Morphing Pseudocode



```
GenerateAnimation(Image<sub>0</sub>, L_0[...], Image<sub>1</sub>, L_1[...])
begin
    foreach intermediate frame time t do
        for i = 1 to number of line pairs do
            L[i] = line t^{th} of the way from L_0[i] to L_1[i]
        end
        Warp_0 = WarpImage(Image_0, L_0, L)
        Warp_1 = WarpImage(Image_1, L_1, L)
        foreach pixel p in FinalImage do
            \text{Result}(p) = (1-t) \text{Warp}_0 + t \text{Warp}_1
        end
    end
end
```







Amy Ousterhout

COS426 Examples





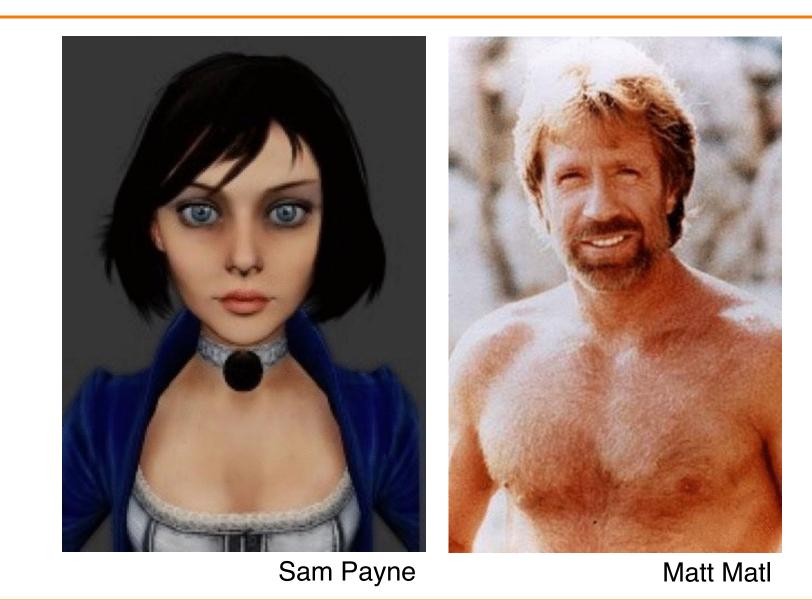
ckctwo

BELONA

Jon Beyer

COS426 Examples







- "Computational photography": new photographic effects that inherently use multiple images + computation
- Example: stitching images into a *panorama*



• Photo montage



• Stoboscopic images



• Extended depth-of-field





Scene Completion Using Millions of Photographs

James Hays and Alexei A. Efros

SIGGRAPH 2007

Slides by J. Hays and A. Efros







Image Completion

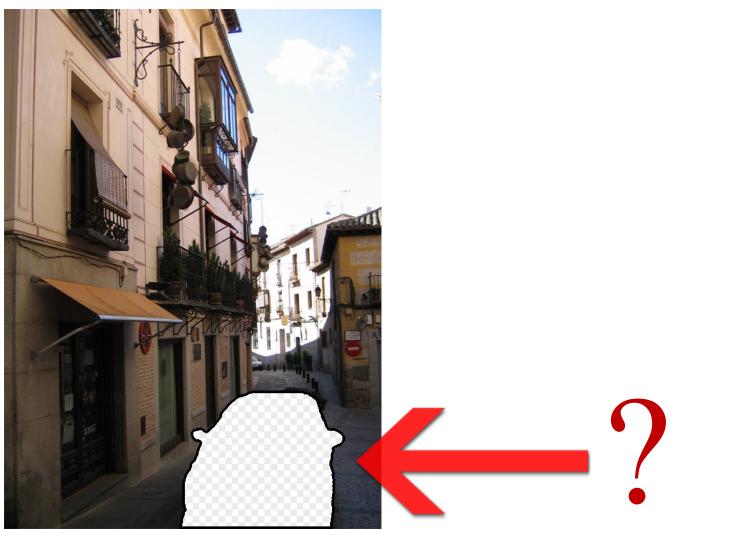


Image Completion

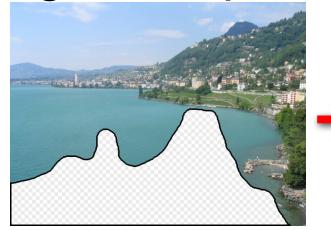
2.3 Million unique images from Flickr



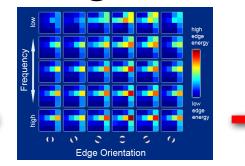


Scene Completion Result

Image Completion Algorithm



Input image





Scene Descriptor



Image Collection





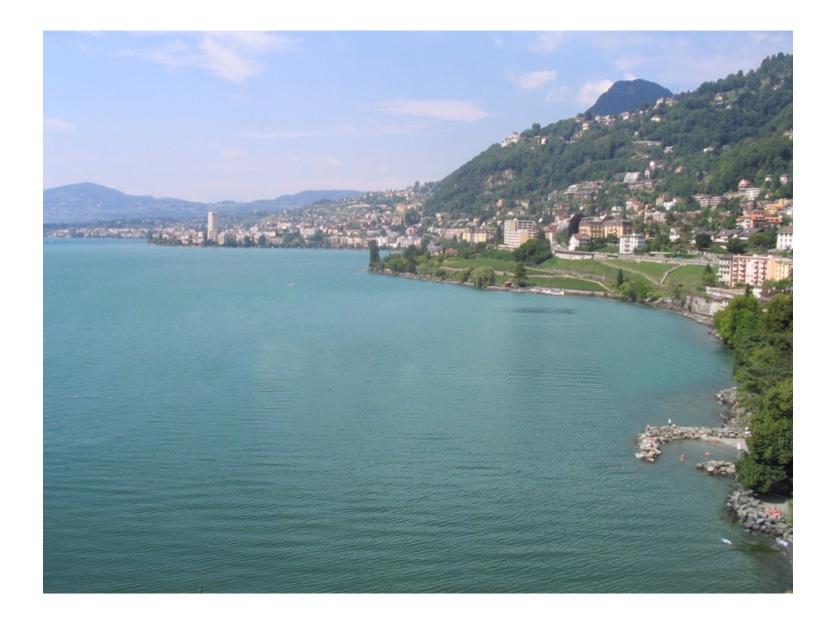
Mosaicing

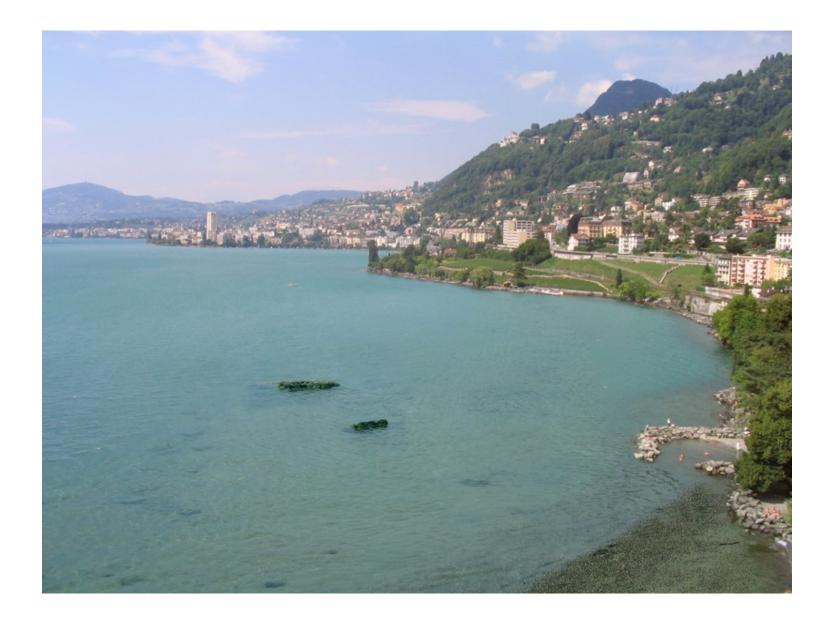


200 matches

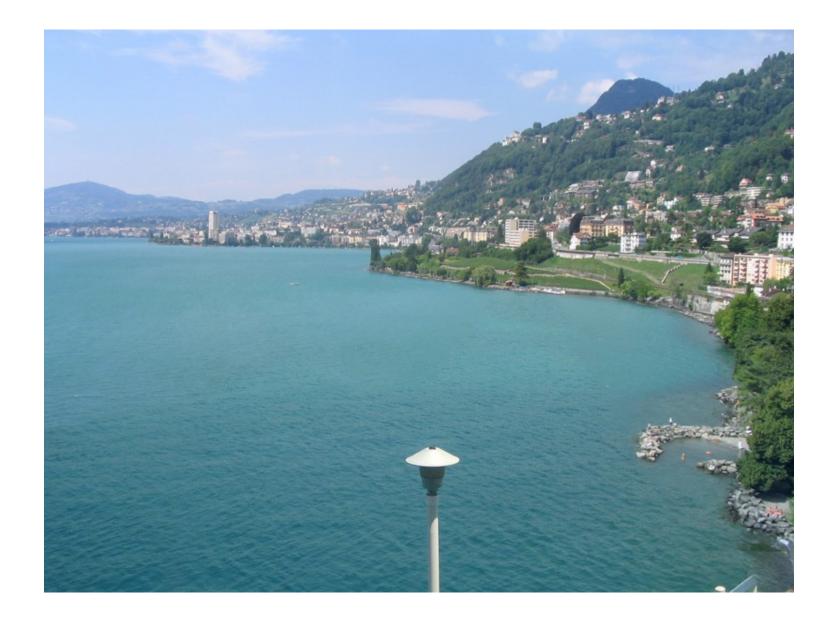
. . .

20 completions







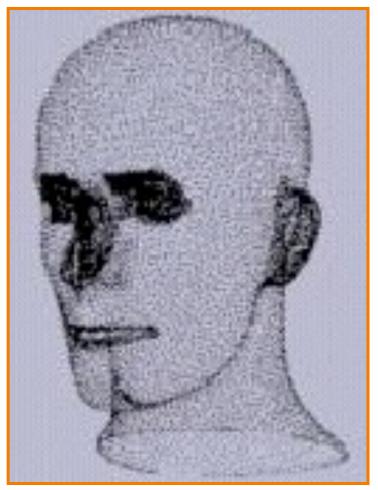




Summary

- Image compositing
 - Alpha channel
 - Porter-Duff compositing algebra
- Image morphing
 - Warping
 - Compositing
- Compositing in Computational Photography

Next Time: 3D Modeling



Hoppe

