

Face morphing generation and detection

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Outline

- 1 Introduction
- 2 Problem Statement
- 3 Results
- 4 Face Morphing Attack Detection

Introduction of image/face morphing

- Image morphing is one of the seminal problems in Computer Graphics
- Numerous artistic applications

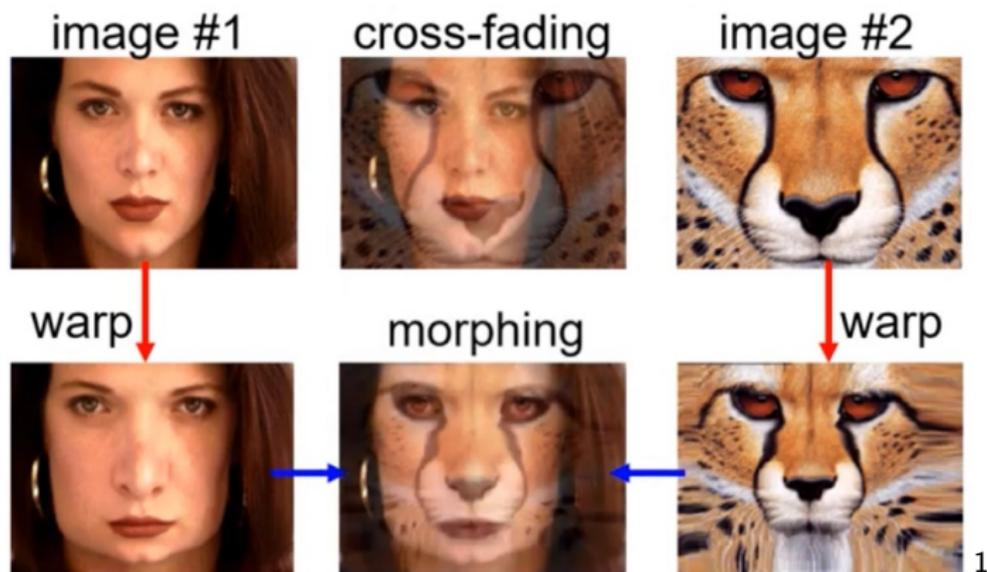


Figure: Image taken from: <https://www.cartoonbrew.com/vfx/oral-history-morphing-michael-jacksons-black-white-144015.html>

Problems of image/face morphing

- Cross-fades work only on previously aligned images
- If not aligned, we need to **warp** them for feature alignment
- And **blend** their colors for a smooth transition between images

Problems of image/face morphing



¹ Image taken from: Gomes, J, *Warping and Morphing of Graphical Objects*, 1998, Morgan Kaufmann.

Motivation

- The transformations up to this point are not **continuous**, or **smooth**
- Why is this important or useful?
- If the transformation is smooth, we can leverage its **differential properties** during the warping/blending
- Also, the transformation is a continuum between the images, instead of a finite number of states

Problem statement

- Coordinate-based neural networks were proposed as representations for signed distance functions in 3D
- They build a continuous and smooth representation of 2D and 3D objects^{2,3}

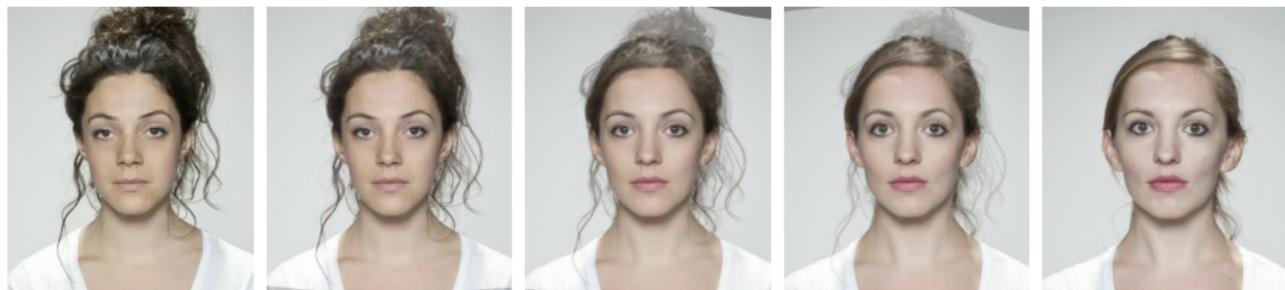


²Schirmer L, et al. Neural Networks for Implicit Representations of 3D Scenes. In: Proceedings of the 34th SIBGRAPI.

³Novello T, et al., Exploring differential geometry in neural implicits. In: Computers & Graphics, Volume 108, 2022.

Problem statement

- Can we leverage them to model warpings/blendings of (face) images?



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Assumptions and background

- We assume the images to be *neural images*, i.e., represented by sinusoidal multilayer perceptrons
 - $I_0, I_1 : \mathbb{R}^2 \rightarrow \mathcal{C}$, where \mathcal{C} is the color space
 - With this, we can calculate its derivatives using automatic differentiation
- We define a *neural warping* $\mathbf{T} : \mathbb{R}^2 \times \mathbb{R} \rightarrow \mathbb{R}^2$ to align the key-points $\{p_i, q_i\}$
 - The key-points may be extracted using face landmark detection (such as DLib 68 landmarks model⁴)
 - $\mathbf{T}(\mathbf{x})$ receives a pair of coordinates and outputs the warped coordinates
- We propose a *neural blending* $\mathcal{M}(\theta)$ to combine the warped images

⁴Kazemi, V. and Sullivan J., One millisecond face alignment with an ensemble of regression trees. In Proceedings of the IEEE CVPR 2014

Warping Loss

$$\mathcal{L}(\theta) = \mathcal{W}(\theta) + \mathcal{D}(\theta) + \mathcal{I}(\theta). \quad (1)$$

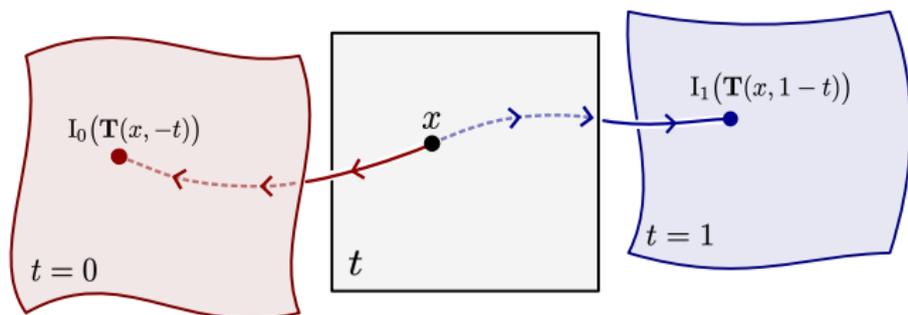
$\mathcal{W}(\theta)$, $\mathcal{D}(\theta)$, $\mathcal{I}(\theta)$ are the *warping*, *data*, *thin-plate* constraints

$$\mathcal{W}(\theta) = \underbrace{\int_{\mathbb{R}^2} \|\mathbf{T}(x, 0) - x\|^2 dx}_{\text{Identity constraint}} + \underbrace{\int_{\mathbb{R}^2 \times \mathbb{R}} \|\mathbf{T}(\mathbf{T}(x, t), -t) - x\|^2 dx dt}_{\text{Inverse constraint}}. \quad (2)$$

$$\mathcal{D}(\theta) = \int_{[0,1]} \|\mathbf{T}(p_i, t) - \mathbf{T}(q_i, 1 - t)\|^2 dt \quad (3)$$

$$\mathcal{I}(\theta) = \int_{\mathbb{R}^2 \times \mathbb{R}} \|\mathbf{Hess}(\mathbf{T})(x, t)\|_F^2 dx dt \quad (4)$$

An illustration of warping



Blending loss

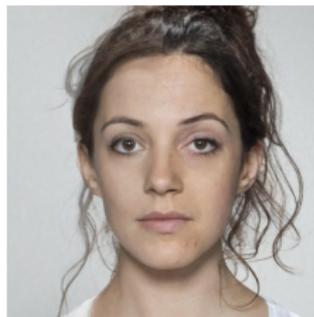
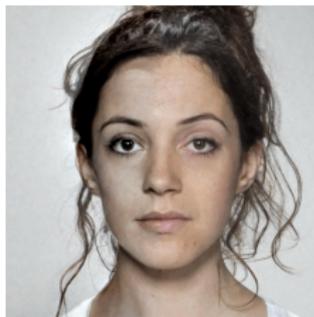
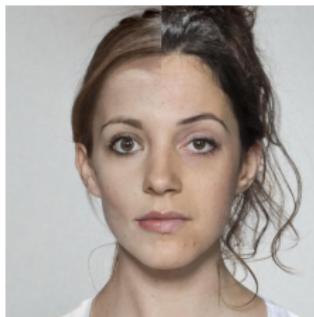
$$\mathcal{M}(\theta) = \underbrace{\int_{\Omega} \|\text{Jac}(\mathbf{I}) - U\|^2 dxdt}_{\mathcal{C}(\theta)} + \underbrace{\int_{S-\Omega} (\mathbf{I} - \mathbf{I}^*)^2 dxdt}_{\mathcal{B}(\theta)}. \quad (5)$$

where U is a matrix obtained by blending $\text{Jac}(\mathbf{I}_i)$, $\mathcal{C}(\theta)$ fits \mathbf{I} to the primitive of U in Ω , and $\mathcal{B}(\theta)$ is a *boundary constraint* to fit \mathbf{I} to \mathbf{I}^* in $S - \Omega$.

Reference images



Blending comparisons



No warping

seamless cloning

average cloning

mixed cloning

Varying gender and ethnicity



Interpolation 3D

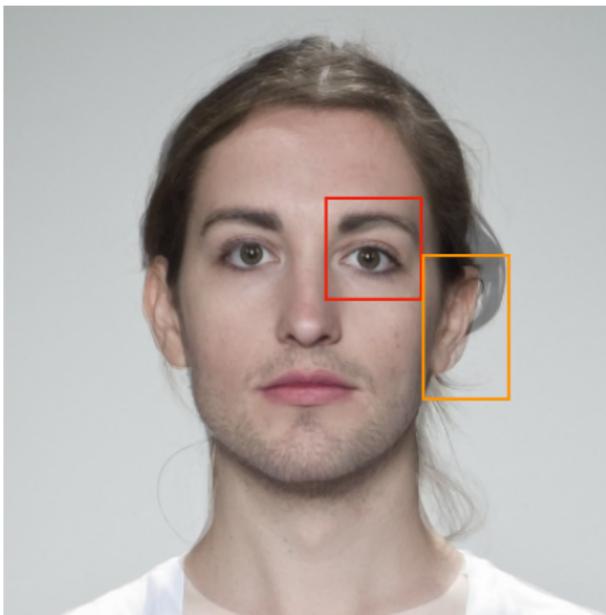


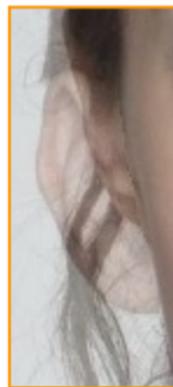
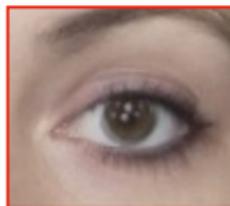
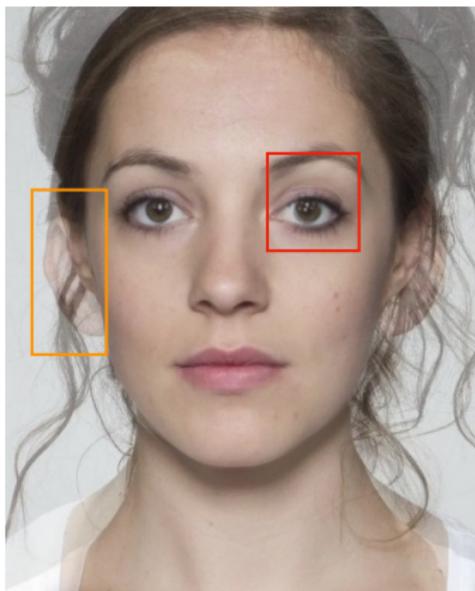
Warp 3D



How to detect morphings?

- Subtle changes in texture
- Misaligned features
- Ghost artifacts





- (Usually) based on convolutional neural networks for feature extraction
- Modelled as a classification task (is this a true face? or is it morphed?)
- More sophisticated networks were proposed (fused classification)⁵

⁵See: Medvedev, I., Shadmand, F., Gonçalves, N. MorDeepHy: Face Morphing Detection Via Fused Classification. International Conference on Pattern Recognition Applications and Methods

Questions?

- Novello T, da Silva V, Schardong G, Schirmer L, Lopes H, Velho L. Neural Implicit Surface Evolution using Differential Equations. Published online 2022. doi:10.48550/ARXIV.2201.09636, <https://arxiv.org/abs/2201.09636>
- Medvedev, I., Shadmand, F., Gonçalves, N. MorDeepHy: Face Morphing Detection Via Fused Classification. International Conference on Pattern Recognition Applications and Methods. Published online 2022. doi:10.48550/arXiv.2208.03110, <https://arxiv.org/abs/2208.03110>