



# Face modeling (part 2)

Jun-Yan Zhu

16-726 Learning-based Image Synthesis, Spring 2022

# Why Human Faces?

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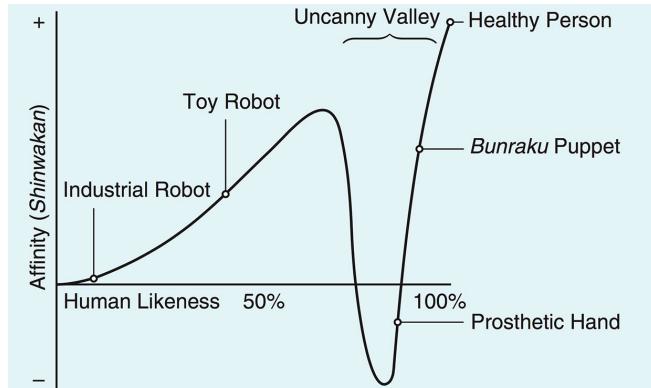
- Face is an important subject.
  - We are humans.
  - Many commercial applications.
- Lots of useful tools
  - 3D data: geometry-based synthesis.
  - 2D/3D Computer vision works for faces.



# Is Face Modeling Easy/Hard?

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- Face modeling is easy?
  - Plenty of aligned 3D face data.
  - 2D and 3D computer vision methods.
- Face modeling is hard?
  - Uncanny valley: Human eyes are extremely sensitive to any imperfections on faces.



# The Morphable face model (3D + PCA)

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Again, assuming that we have  $m$  such vector pairs in full correspondence, we can form new shapes  $\mathbf{S}_{model}$  and new appearances  $\mathbf{T}_{model}$  as:

$$\mathbf{S}_{model} = \sum_{i=1}^m a_i \mathbf{S}_i \quad \mathbf{T}_{model} = \sum_{i=1}^m b_i \mathbf{T}_i$$

$$s = \alpha_1 \cdot \text{face}_1 + \alpha_2 \cdot \text{face}_2 + \alpha_3 \cdot \text{face}_3 + \alpha_4 \cdot \text{face}_4 + \dots = \mathbf{S} \cdot \mathbf{a}$$

$$t = \beta_1 \cdot \text{face}_1 + \beta_2 \cdot \text{face}_2 + \beta_3 \cdot \text{face}_3 + \beta_4 \cdot \text{face}_4 + \dots = \mathbf{T} \cdot \mathbf{b}$$

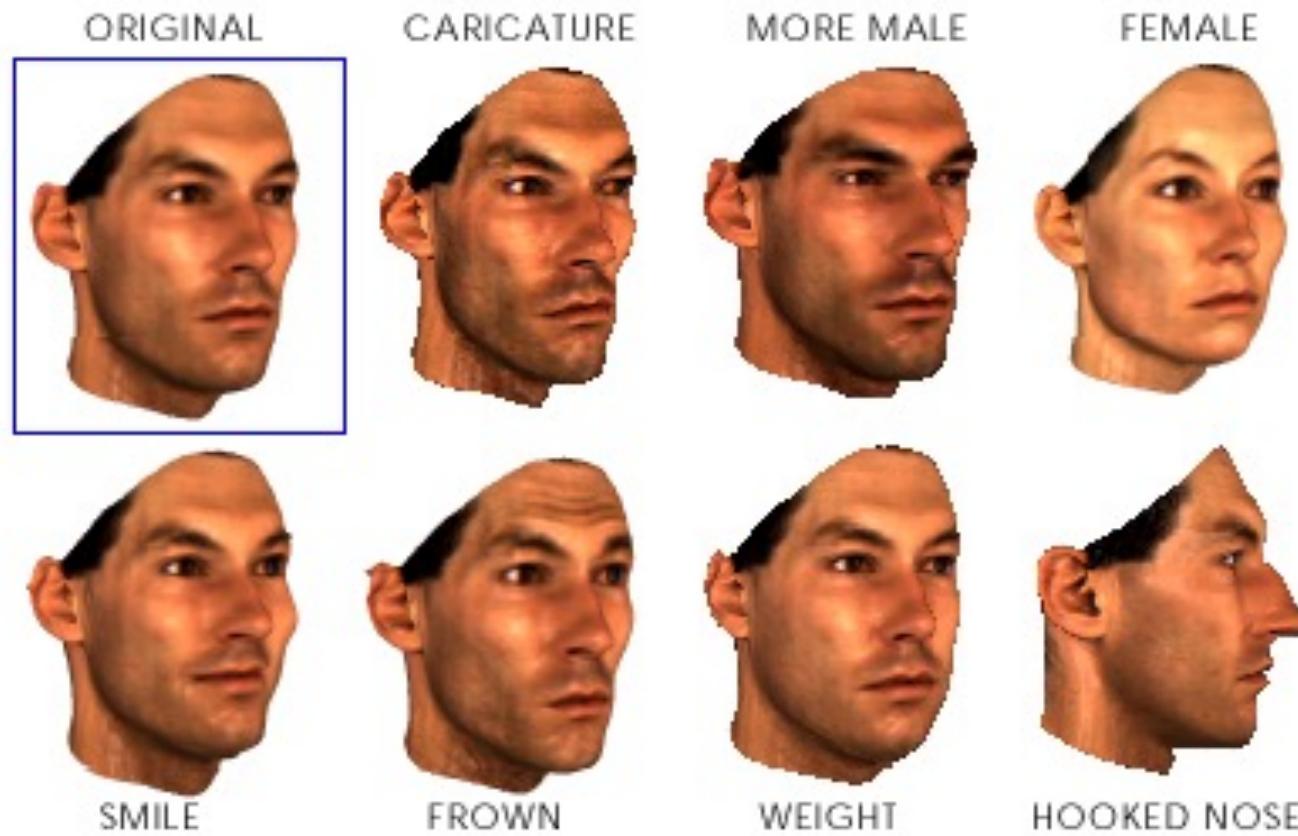
If number of basis faces  $m$  is large enough to span the face subspace then:

Any new face can be represented as a pair of vectors

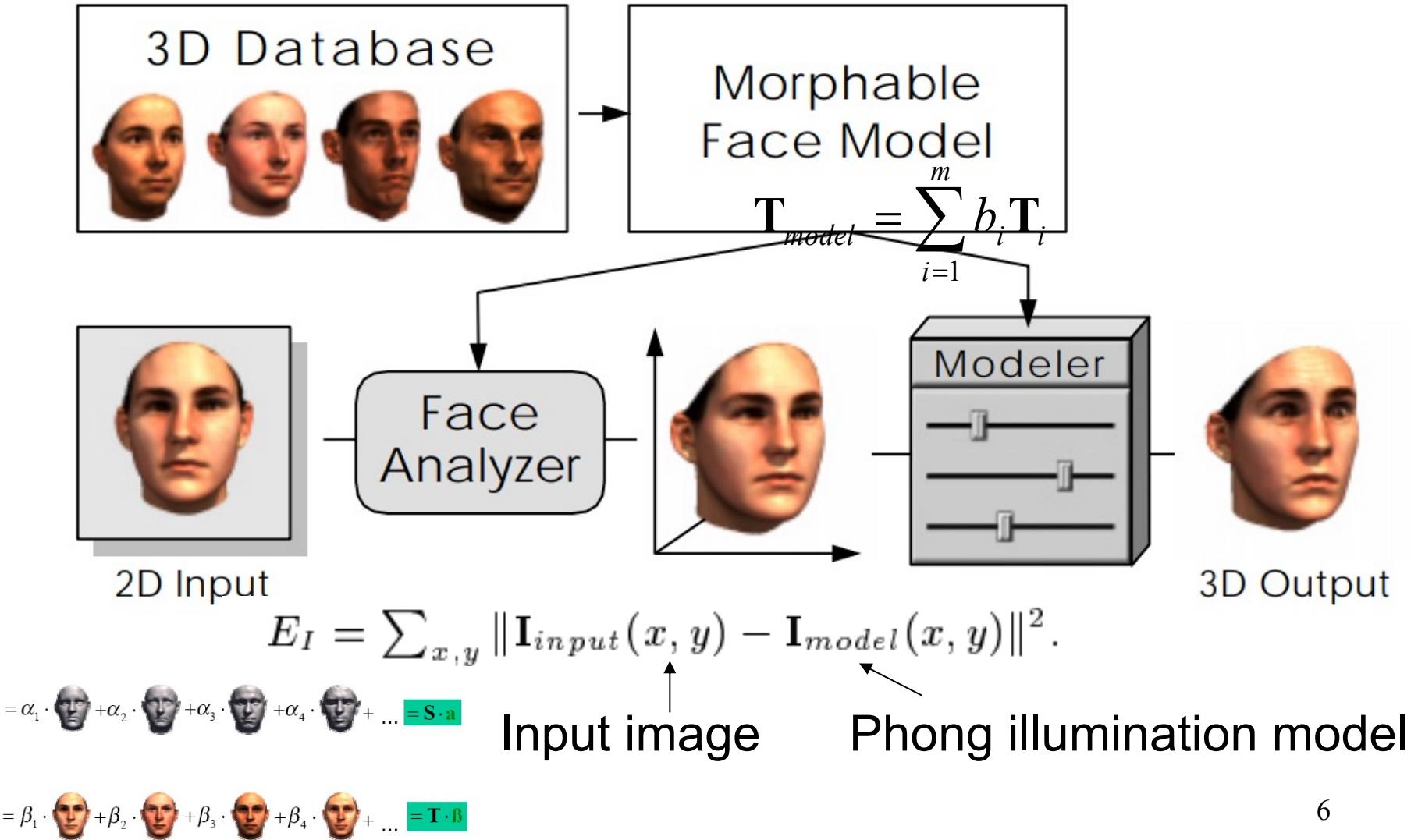
$(\alpha_1, \alpha_2, \dots, \alpha_m)^T$  and  $(\beta_1, \beta_2, \dots, \beta_m)^T$  !

# Using 3D Geometry: Blanz & Vetter, 1999

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# Using 3D Geometry: Blanz & Vetter, 1999

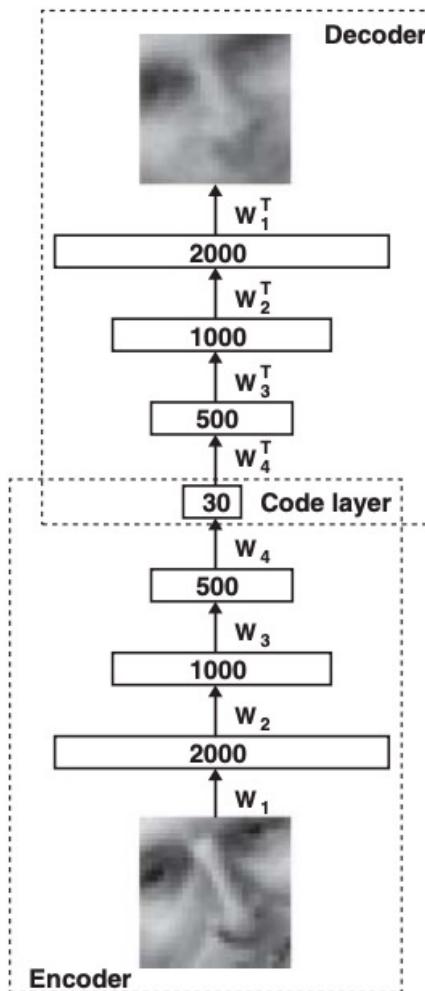


# How to Improve the results?

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- Using Deep Learning?
- But how?
- Deep learning vision methods:
  - 2D/3D landmark detection
  - 3D pose estimation
  - Face shape reconstruction
- Deep learning graphics models
  - generative models
  - 3D-aware generative models

# Autoencoder vs. PCA



Training objective: E encoder, G decoder/generator

$$\arg \min_{E,G} \mathbb{E}_x \|G(E(x)) - x\|_2$$



Top: Input. Middle: Autoencoder. Bottom: PCA

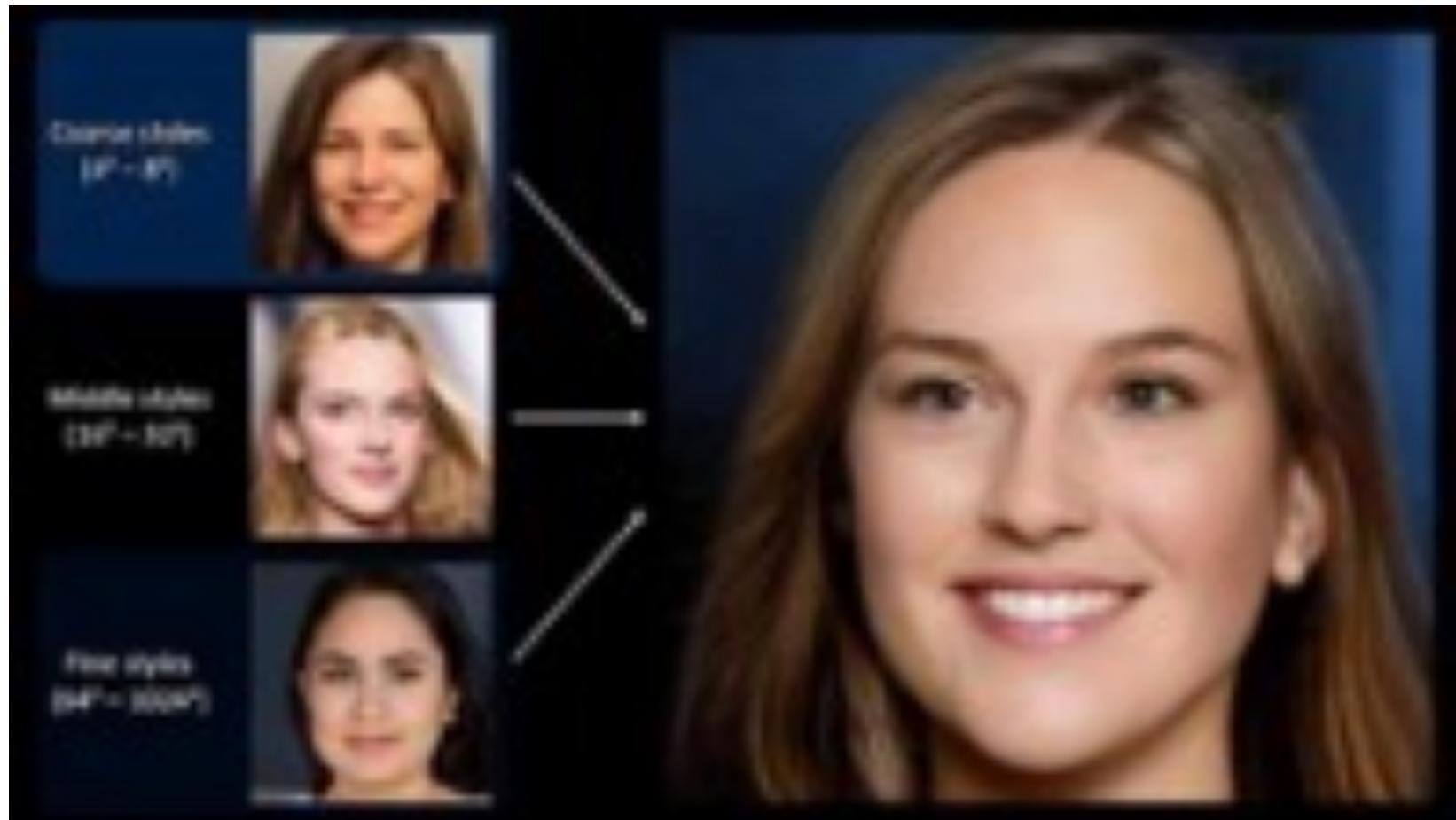
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# Deep learning method

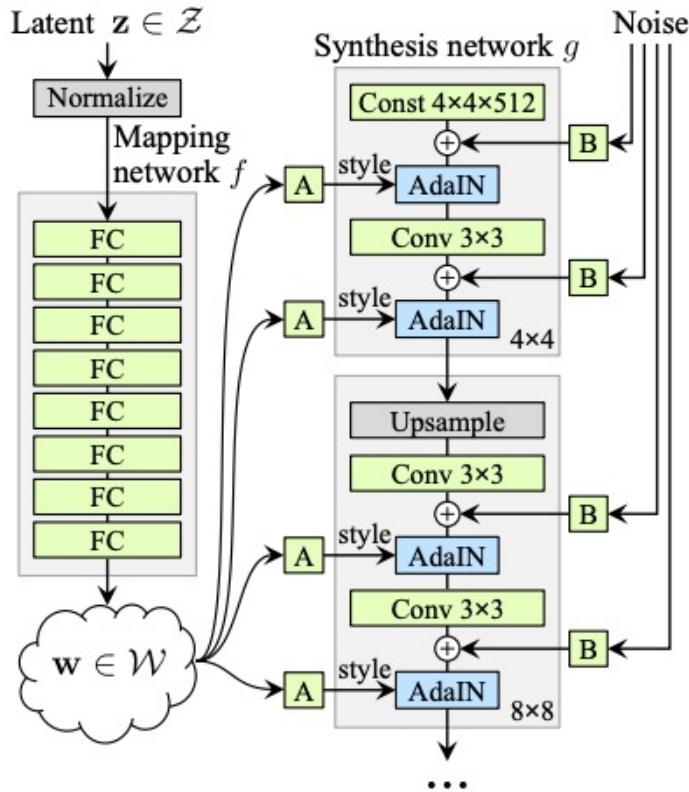
## PCA → Generative Model

# StyleGAN Face Results

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# Face Editing with GANs Projection



Optimizing the latent code

$$z^* = \arg \min_z \mathcal{L}(G(z), x)$$

Optimizing the style code

$$w^* = \arg \min_w \mathcal{L}(g(w), x)$$

Optimizing the extended style code

$$w_+^* = \arg \min_{w+} \mathcal{L}(g(w_+), x)$$

# Face Editing = latent space editing

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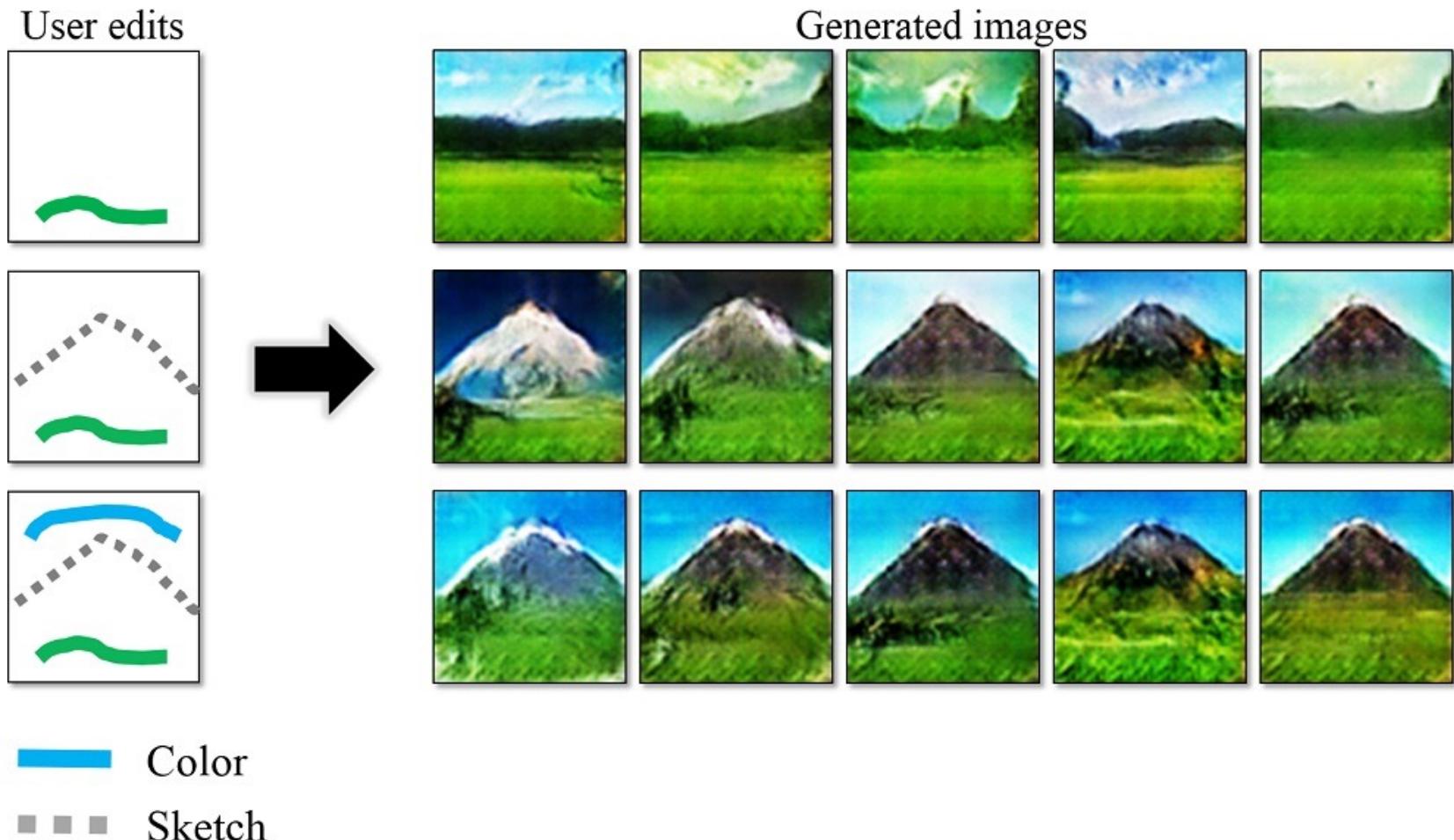
Interpolation between two faces in the  $w^+$  space

# Face Editing = latent space editing

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# Image Editing with GANs Projection



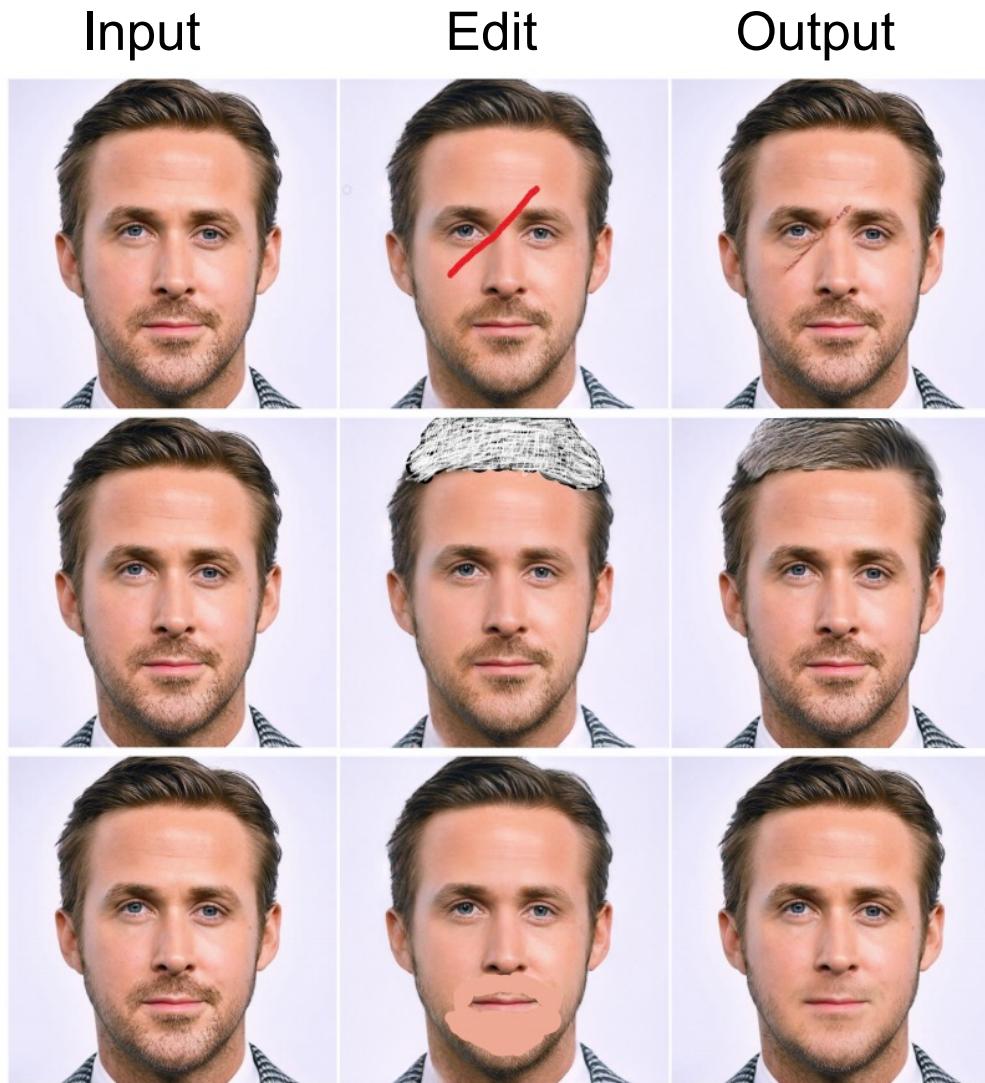
# Face Editing with GANs Projection

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# Face Editing with GANs Projection

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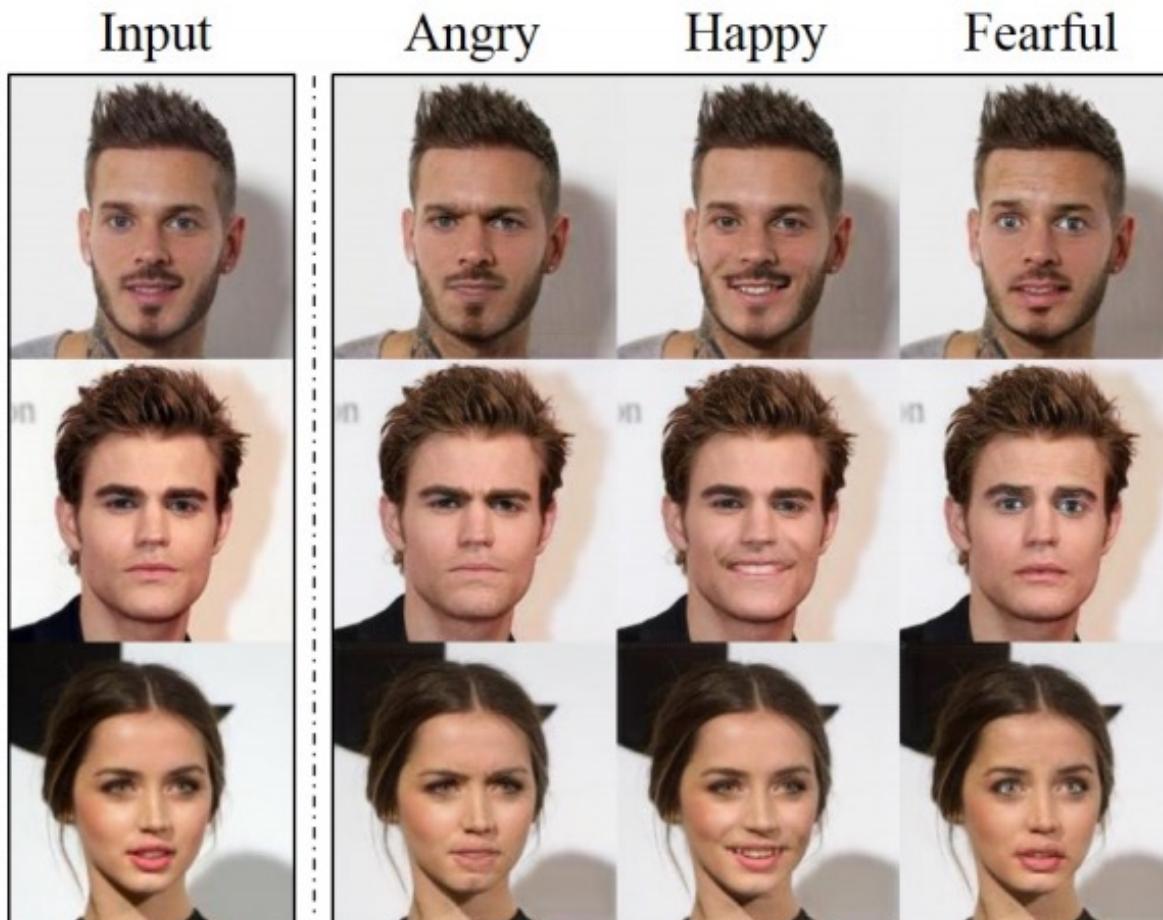
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# Deep learning method

## Image-to-Image Translation

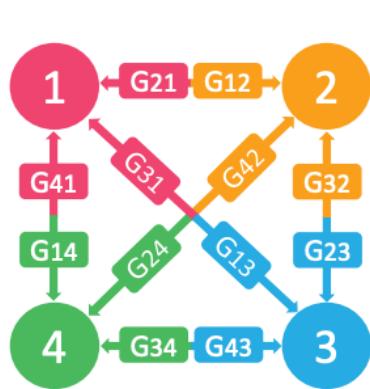
# Face Translation with StarGAN

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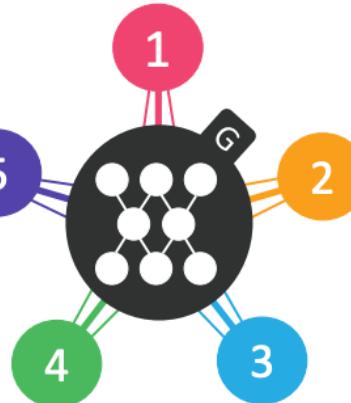


# Face Translation with StarGAN

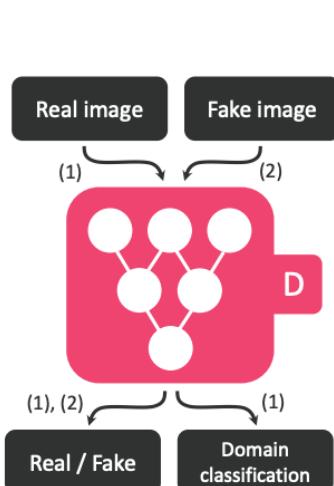
(a) Cross-domain models



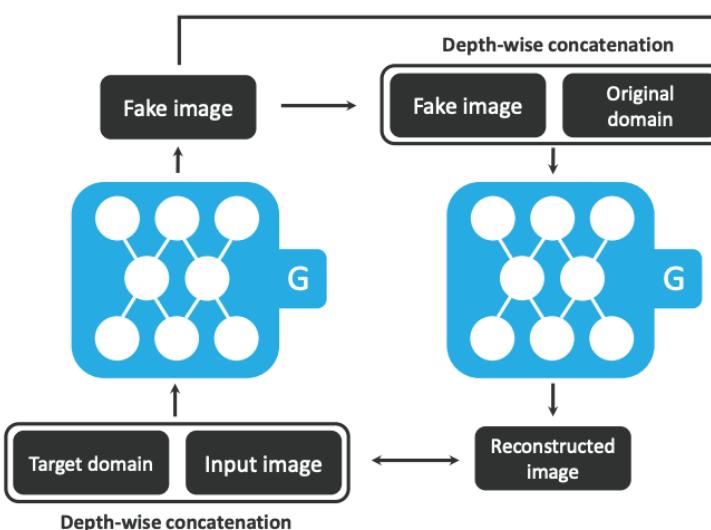
(b) StarGAN



(a) Training the discriminator

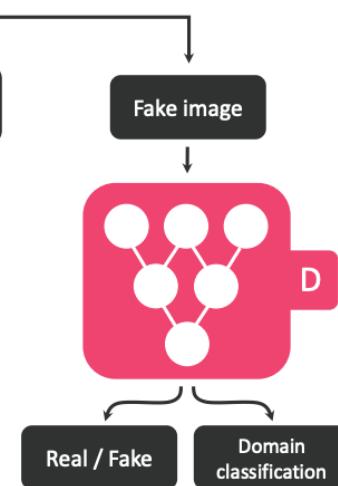


(b) Original-to-target domain



(c) Target-to-original domain

(d) Fooling the discriminator



# Face Translation with StarGAN v2

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Multi-modal synthesis; supports a reference image

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# 3D + Deep Learning

3D representation+ image-to-image

# CGI Face Editing

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Professional video

# CGI Face Editing

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Personal video

Video: ©

<https://www.youtube.com/watch?v=7Flvkn2quLY>

# Applications

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Original video

Pose editing

Expression editing

- Editing of head pose, rotation, face expression and eye gaze
- Combination of model-based face capture and CNN

# 3D + CNN

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## Model-based face capture and reenactment



Garrido et al., ToG 2016

Kemelmacher-Shlizerman et al., ECCV 2010

Shi et al., ToG 2014

Suwajanakorn et al., ICCV 2015

Thies et al., CVPR 2016

Averbuch-Elor et al., ToG 2017

Thies et al., SIGGRAPH 2018

## CNN-based methods



Karras et al., ICLR 2018

Goodfellow et al., NIPS 2014

Isola et al., CVPR 2017

Chen and Koltun, ICCV 2017

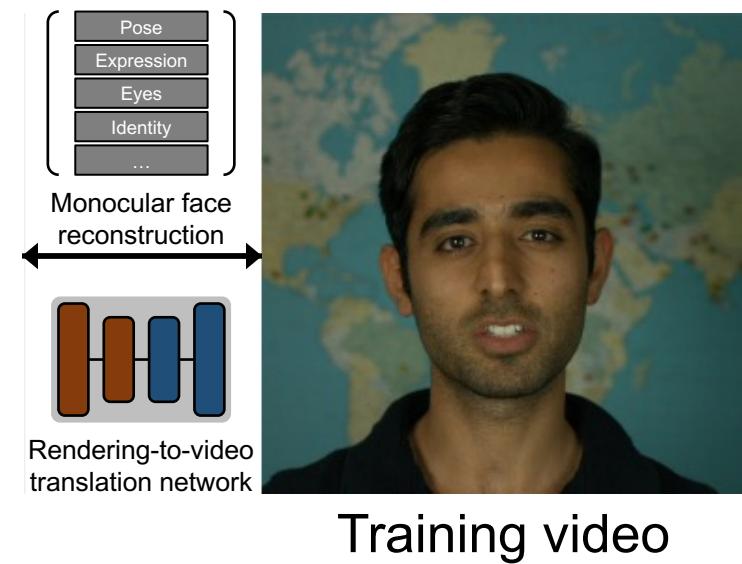
Tewari et al., ICCV 2017

Olszewski et al., ICCV 2018

Wang et al., CVPR 2018

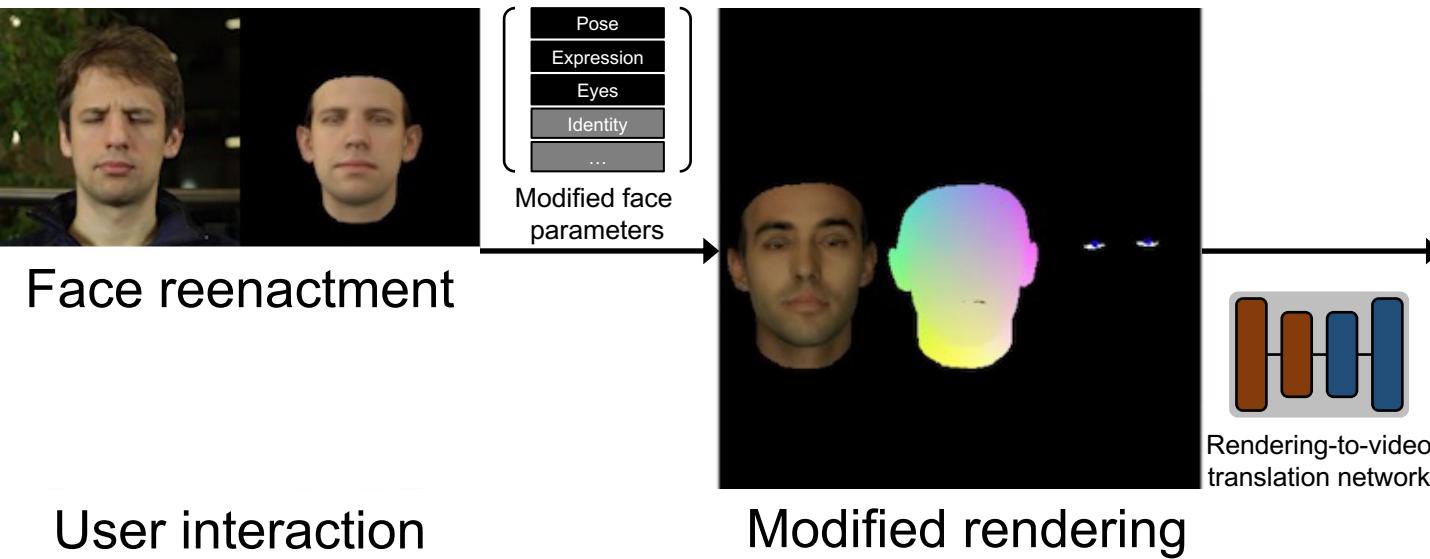
# Overview

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# Overview

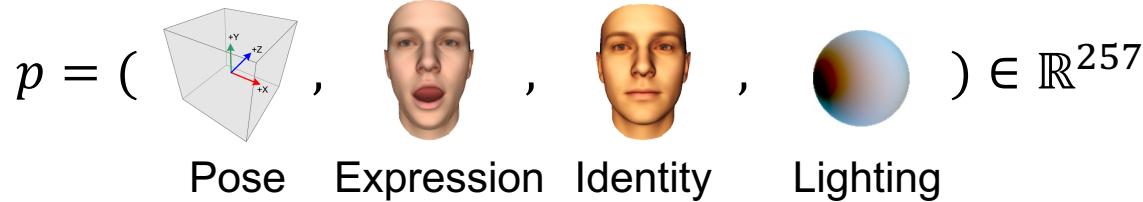
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# Monocular 3D Face Reconstruction

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- Parametric 3D face model

$$p = (\begin{array}{c} \text{Pose} \\ \text{Expression} \\ \text{Identity} \\ \text{Lighting} \end{array}, \quad \begin{array}{c} \text{Image 1} \\ \text{Image 2} \end{array}) \in \mathbb{R}^{257}$$


$$\min_p E(p) = E_{\text{photo}}(p) + E_{\text{land}}(p) + E_{\text{reg}}(p)$$

# Monocular 3D Face Reconstruction

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- Parametric 3D face model

$$p = (\text{Pose}, \text{Expression}, \text{Identity}, \text{Lighting}) \in \mathbb{R}^{257}$$

Pose   Expression   Identity   Lighting

$$\min_p E(p) = E_{\text{photo}}(p) + E_{\text{land}}(p) + E_{\text{reg}}(p)$$

$$\| \text{Image} - \text{Reconstructed Face} \|_2^2$$

# Monocular 3D Face Reconstruction

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- Parametric 3D face model

$$p = (\text{Pose}, \text{Expression}, \text{Identity}, \text{Lighting}) \in \mathbb{R}^{257}$$

Pose      Expression      Identity      Lighting

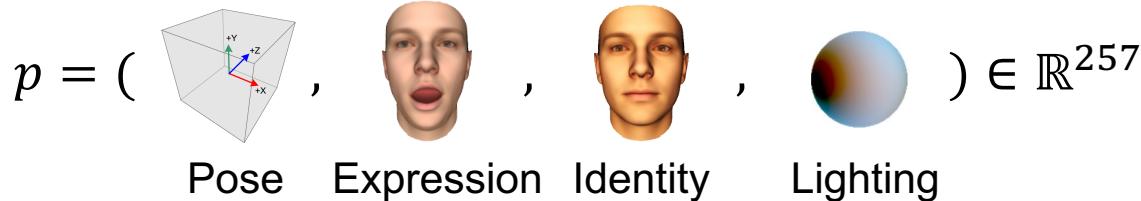
$$\min_p E(p) = E_{\text{photo}}(p) + E_{\text{land}}(p) + E_{\text{reg}}(p)$$

$$\| \text{Image} - \text{Reconstructed Face} \|_2^2 + \| \text{Landmarks} - \text{Reconstructed Landmarks} \|_2^2$$

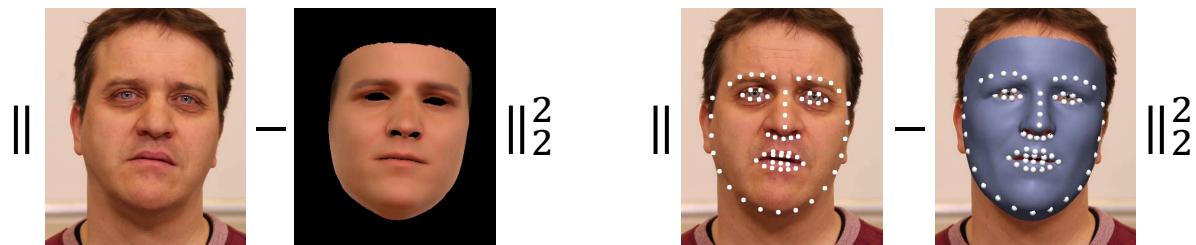
# Monocular 3D Face Reconstruction

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- Parametric 3D face model



$$\min_p E(p) = E_{\text{photo}}(p) + E_{\text{land}}(p) + E_{\text{reg}}(p)$$



Statistical and temporal  
regularization  
Garrido et al., ToG 2016

# Monocular 3D Face Reconstruction

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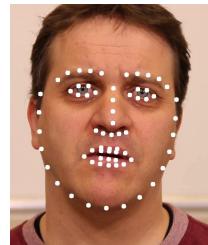
- Parametric 3D face model

$$p = (\begin{array}{c} \text{Pose} \\ \text{Expression} \\ \text{Identity} \\ \text{Lighting} \end{array}, \quad \begin{array}{c} \text{ } \\ \text{ } \\ \text{ } \\ \text{ } \end{array}) \in \mathbb{R}^{257}$$

$$\min_p E(p) = E_{\text{photo}}(p) + E_{\text{land}}(p) + E_{\text{reg}}(p)$$

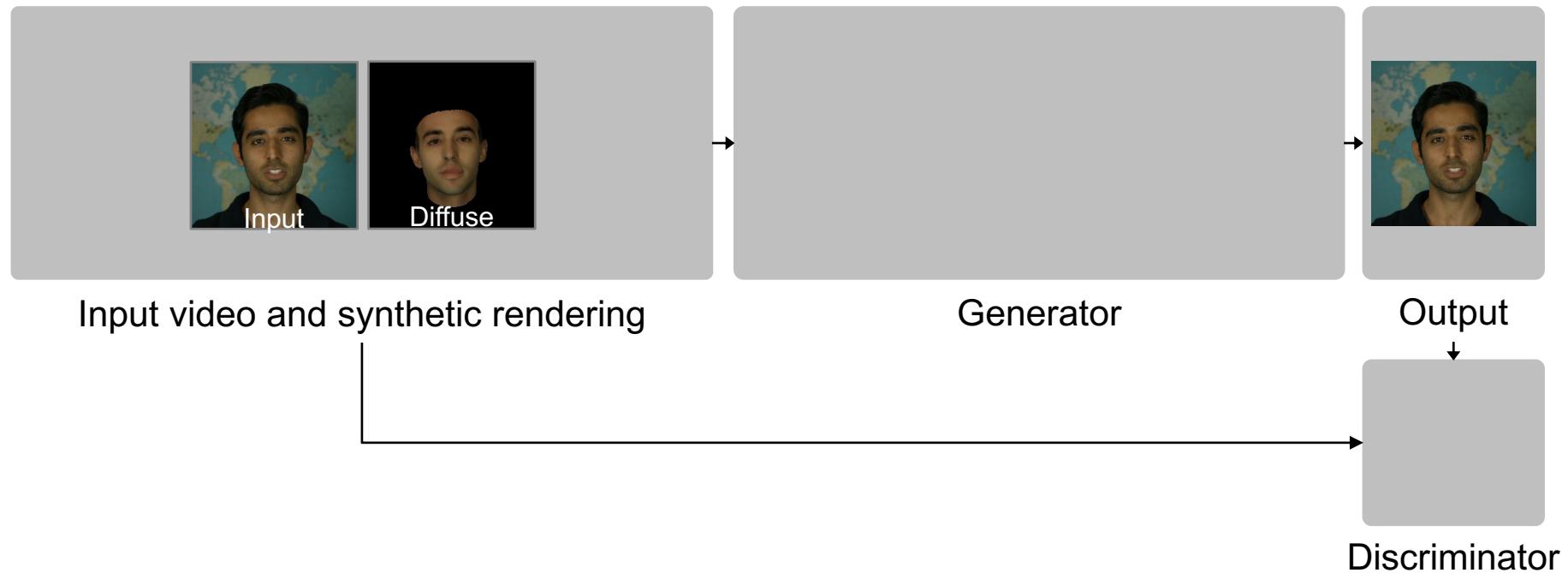
- Eye model

$$e = (\begin{array}{c} \text{ } \\ \text{ } \end{array}) \in \mathbb{R}^4$$



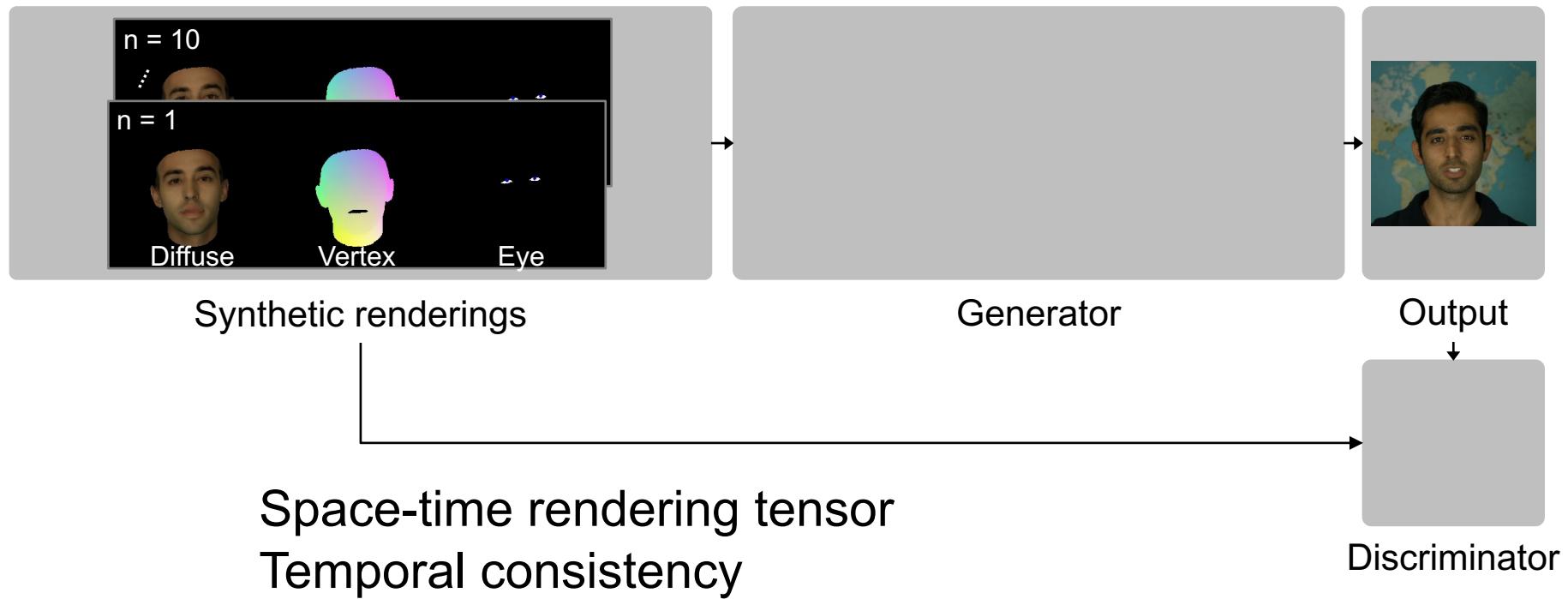
Saragih et al.,  
FG 2011

# Rendering-to-Video Translation Network

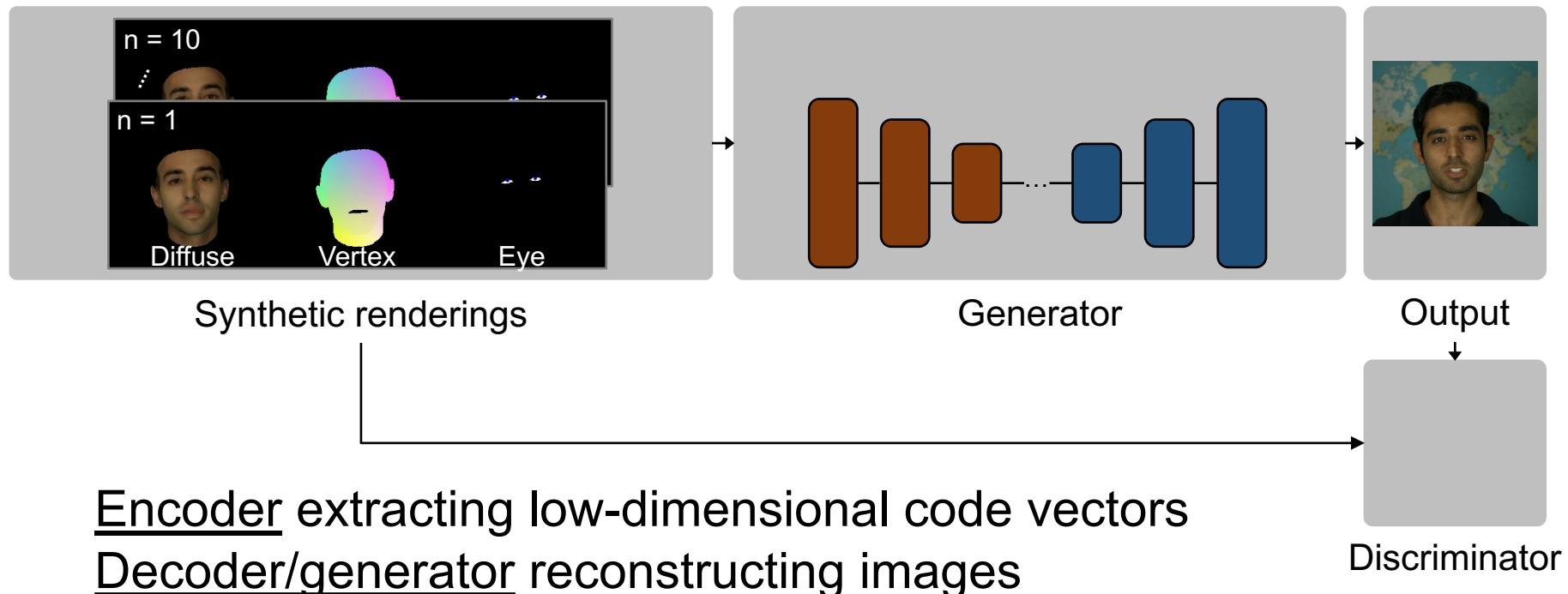


# Rendering-to-Video Translation Network

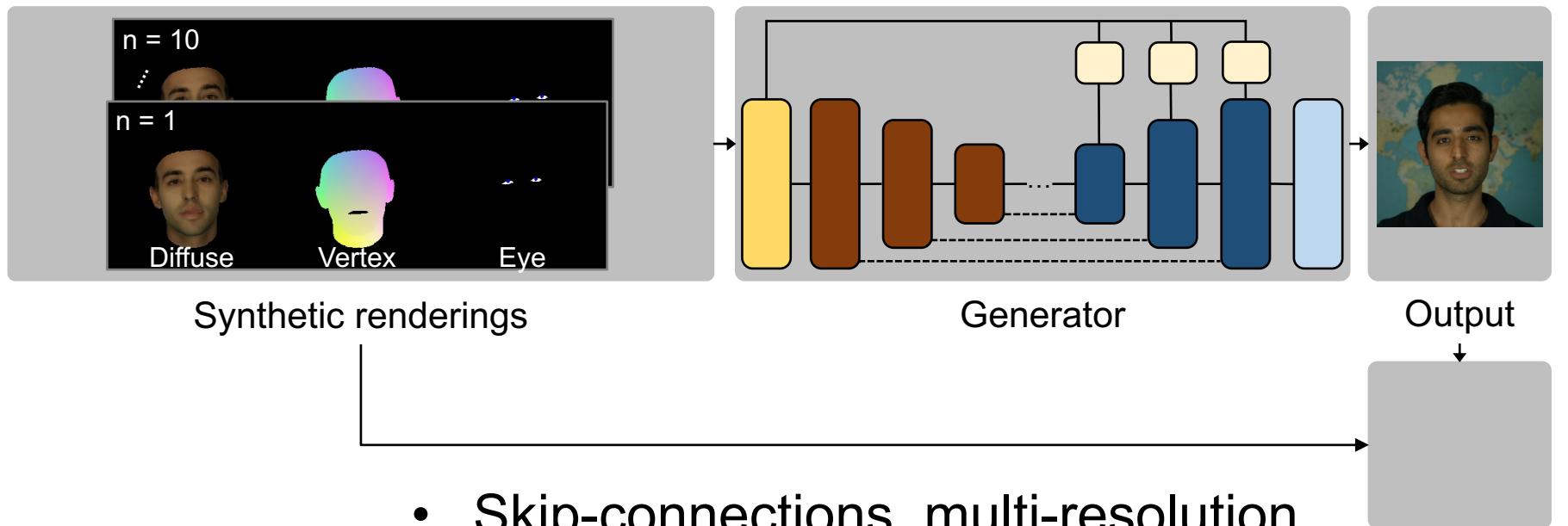
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# Rendering-to-Video Translation Network

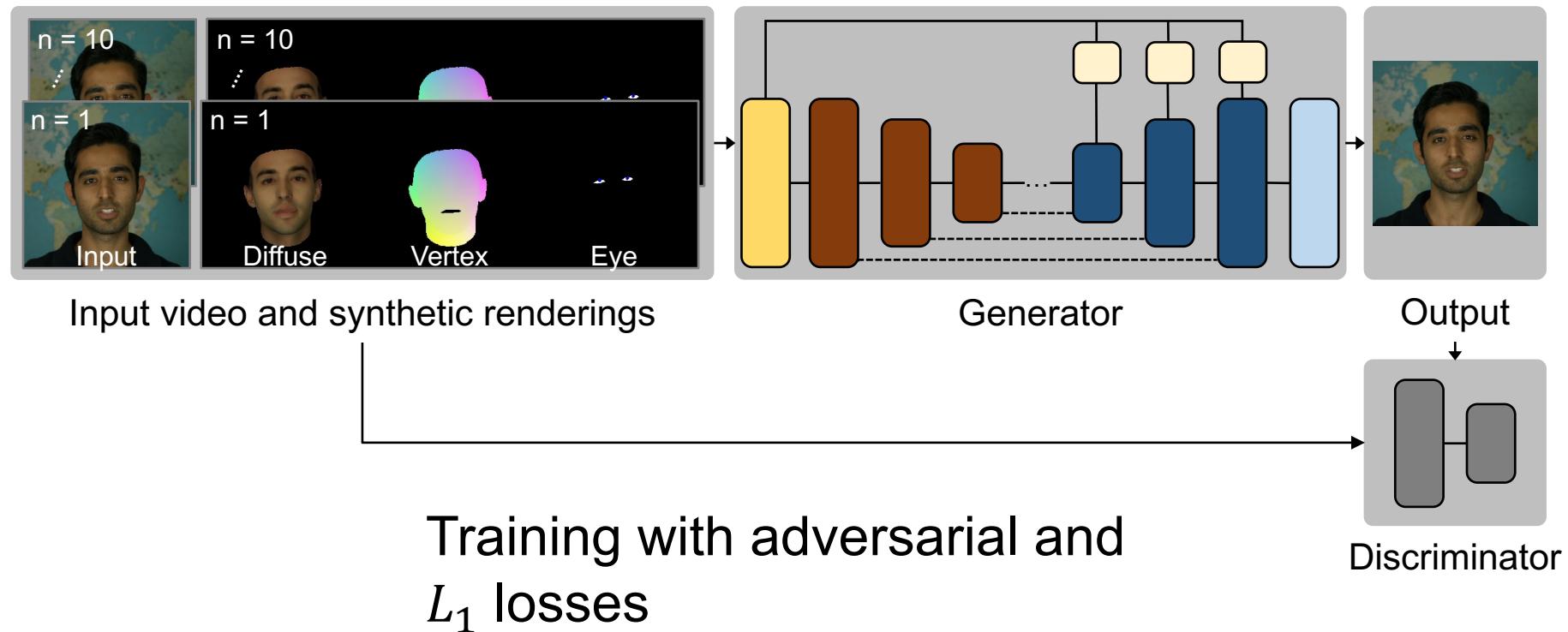


# Rendering-to-Video Translation Network



- Skip-connections, multi-resolution and refinement
- Fine-scale details

# Rendering-to-Video Translation Network



Training with adversarial and  
 $L_1$  losses

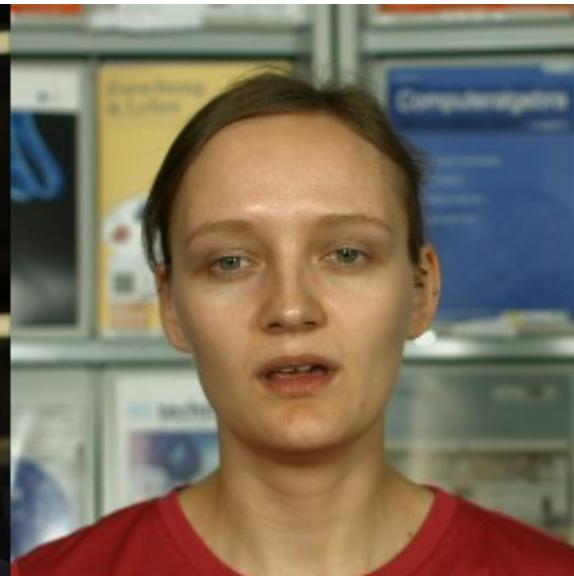
# Result: Facial Reenactment

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Retargeting portraits videos from source to target



Source

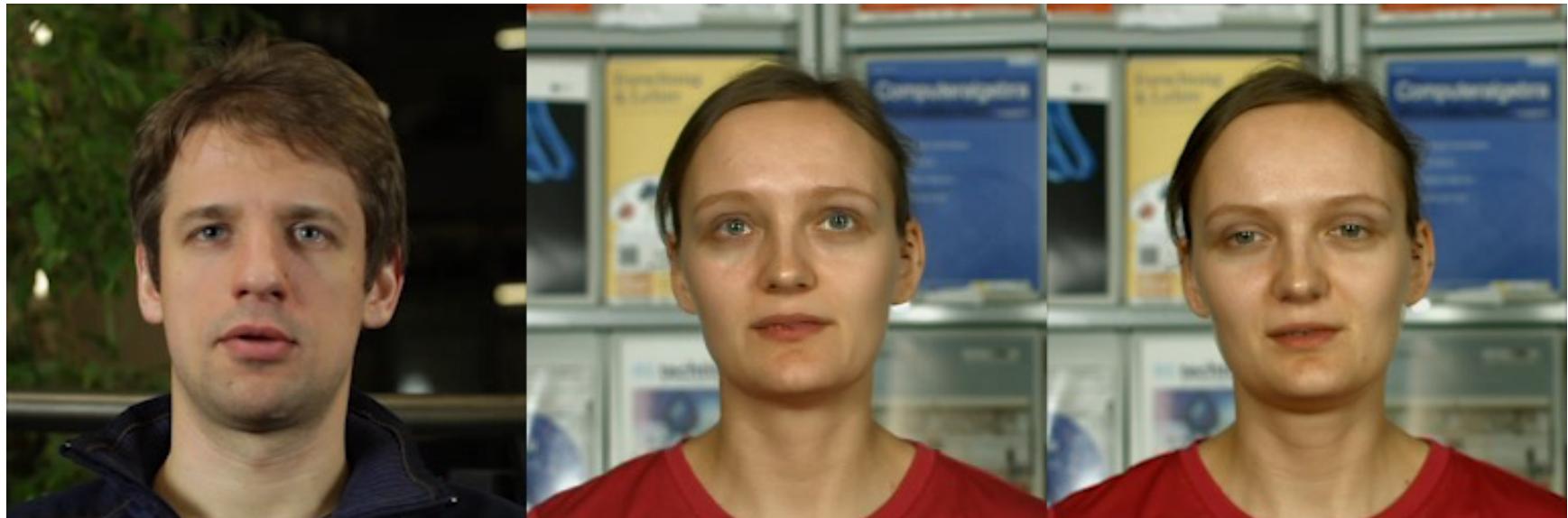


Result

# Result: Facial Reenactment

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Full reenactment of head pose, head rotation, face expression and eye gaze



Source

Result

Face2Face  
(Thies et al., 2016)

# Result: Facial Reenactment

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Source

Target

Result

# Result: Visual Dubbing

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Visual discomfort due to the discrepancy between video and audio tracks



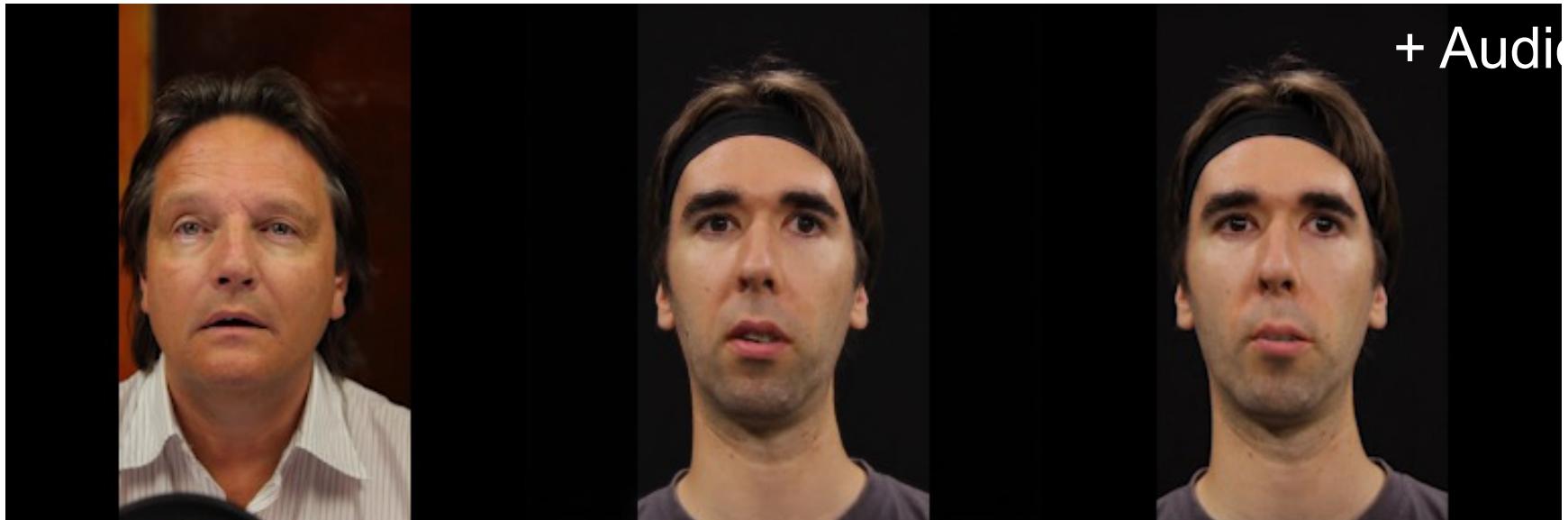
Dubbing actor video

Original video

# Result: Visual Dubbing

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Modification of mouth motion to match audio tracks



Dubbing actor video

Dubbed video

Garrido et al., 2015

# Result: Interactive Editing

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Pose

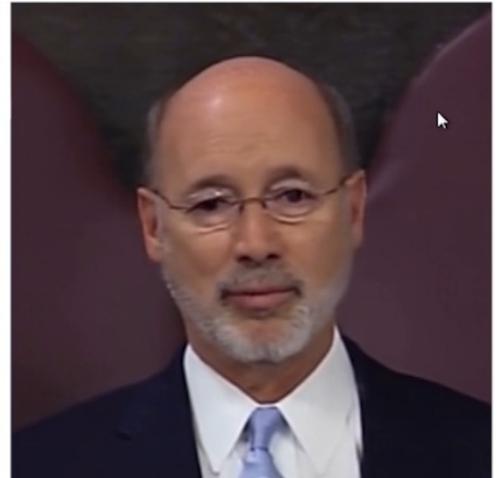
Expression

Shape

Approximately 9 fps

# Result: Interactive Editing

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YouTube videos

2× speed

Approximately 9 fps

Reagan video courtesy of NARA  
(public domain)

Obama video courtesy of the White  
House (public domain)

Wolf video courtesy of Tom Wolf  
(CC BY)

# Result: Post-Production

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Face reshaping

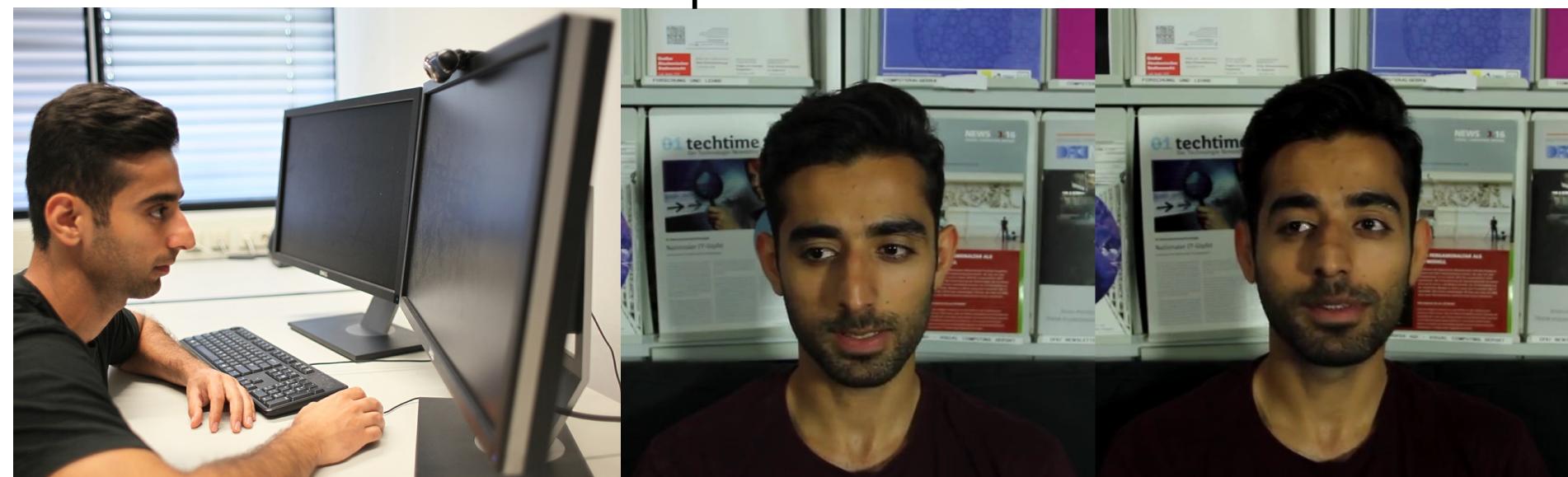
Subtle expression editing

*The Curious Case of Benjamin  
Button*  
video courtesy of Lola Visual Effects

# Result: Pose Correction in Teleconferencing

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Modification of head pose to match camera views



Setup

Camera view

Rotating up

# Result: Multi-View Teleconferencing

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Rotating up + side to side

Model-based video coding: 31 KB/s  
h.264 (e.g., Skype): 192 KB/s

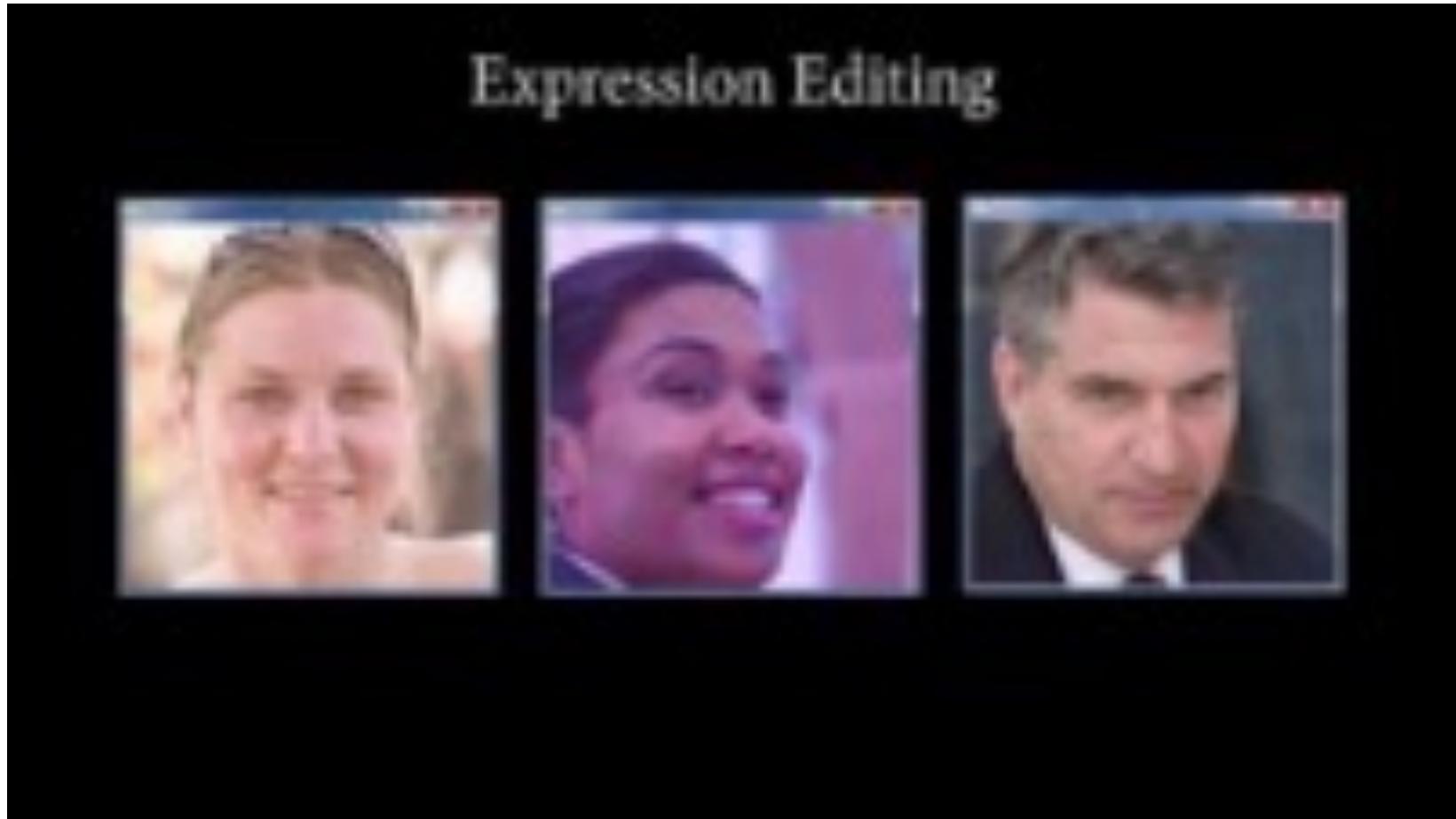
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# 3D + Deep Learning

## 3D representation+ StyleGAN

# StyleRig

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Face reconstruction network as encoder + StyleGAN as generator

# Danger and Ethical Concerns

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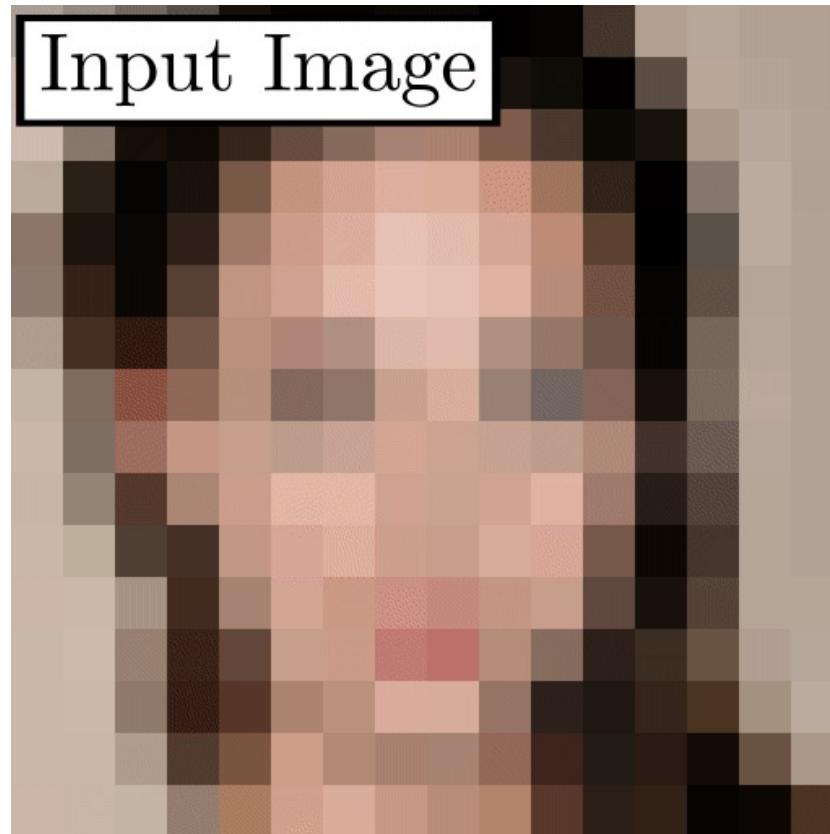


Image Super-resolution [PULSE, CVPR 2020]

# Danger and Ethical Concerns

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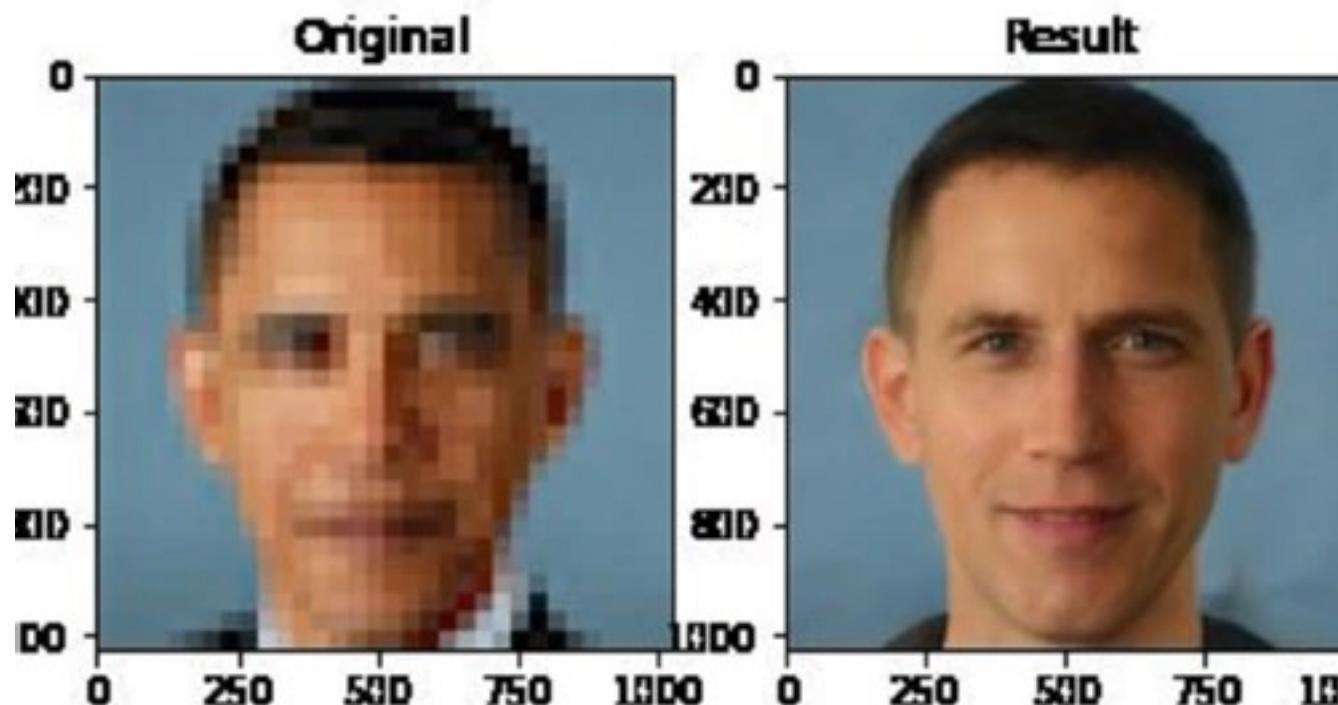
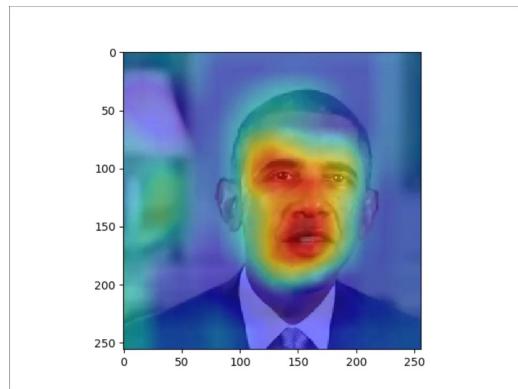


Image Super-resolution [PULSE, CVPR 2020]

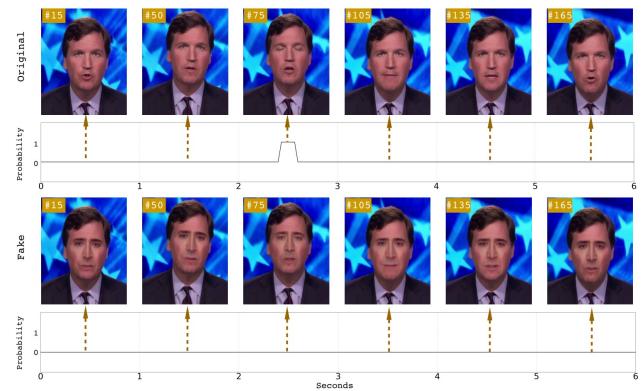
# Detecting Fake Faces



Ours



FaceForensics  
Rössler et al., arXiv 2018



In Ictu Oculi  
Li et al., arXiv 2018

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# Thank You!



16-726, Spring 2022

<https://learning-image-synthesis.github.io/sp22/>