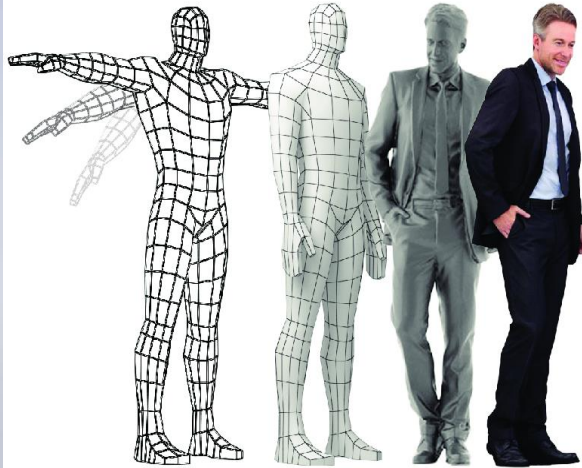


RECENT ADVANCES IN AVATARS

David Moreno, Peter Eisert, Fraunhofer HHI



Vision & Imaging Technologies

Innovative research along the entire video processing chain

Imaging Technologies

Computer Vision & Visual Computing

Imaging Technologies

Capturing Systems

Image/ Video Analysis & Understanding

Generative Models & Animation

Display & visualization technology

Capturing & Processing

Modelling & Digital Representations

Visualization & Rendering

Reasoning

Interaction & HCI

Cognitive & Interactive Systems

Fields of Application



Industry & Construction



Medicine



Multimedia



Security

Avatars as Intuitive Interfaces

Applications



Virtual therapists



Employee- and communication training



Virtual teaching



Entertainment



Medicine & therapy



Virtual assistants



Digital sign language specialist

Realism vs Interactivity



Models

- + Highly interactive, modifications possible
- Realism is expensive (motion & appearance)

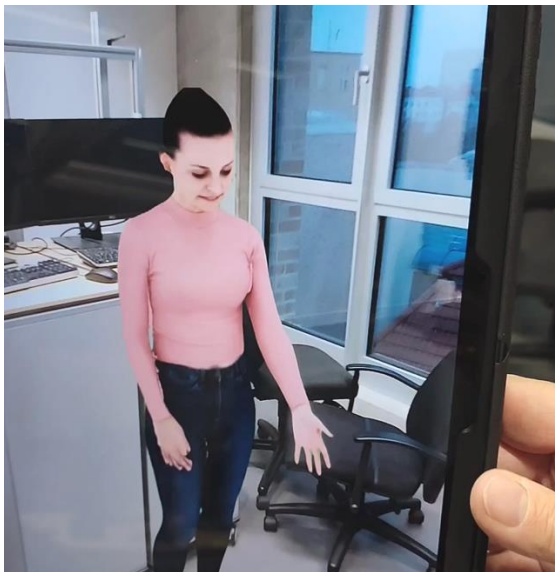
CG Data



Images, videos, volumetric video...

- + Highly detailed and realistic
- No modification possible

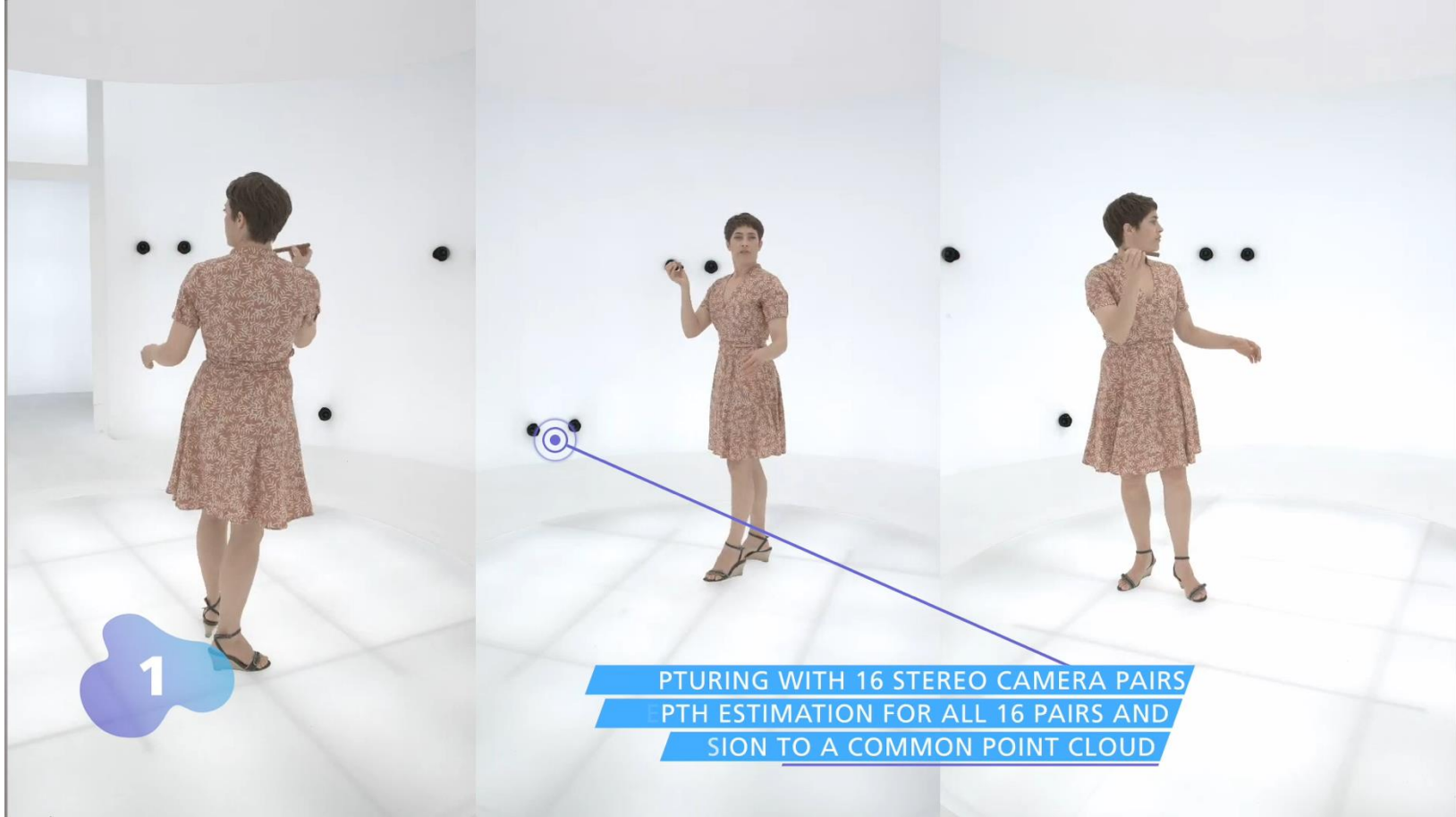
Real World Data



- Creation of highly realistic representations of virtual humans that can naturally interact with users via text and speech
- Personalization of avatars to new target persons
- Style, semantic content and emotion adapted animation of virtual character

Volumetric Video Studio @ Fraunhofer HHI





A. Hilsmann et al, *IET Comp. Vision*, 2020

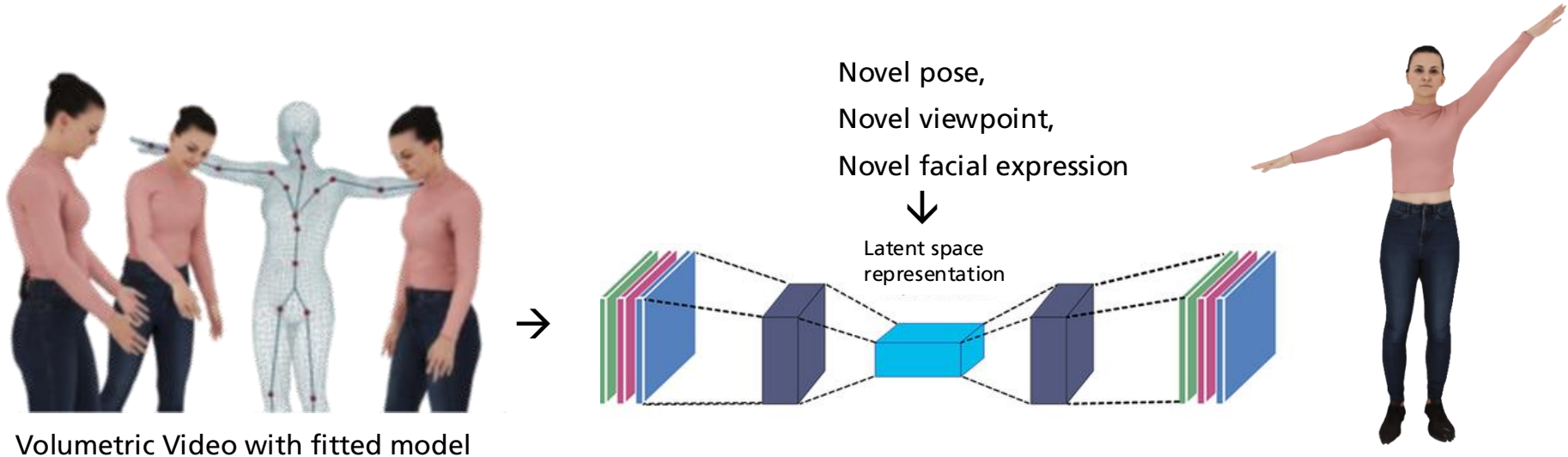
Kinematic Animation of Volumetric Video

Head Rotation: VR Experience Eva Umlauf- Her Testimony



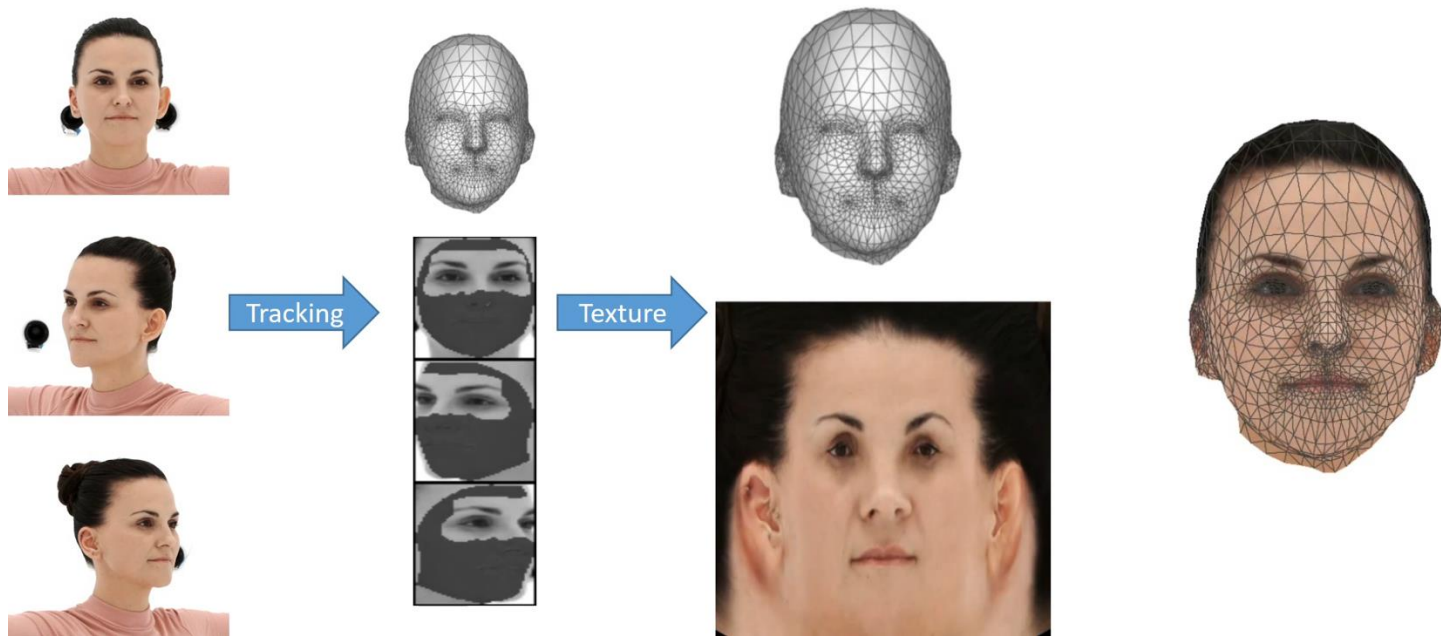
From Volumetric Video to Virtual Humans

Learning animatable representations from volumetric video



Neural Face Representation

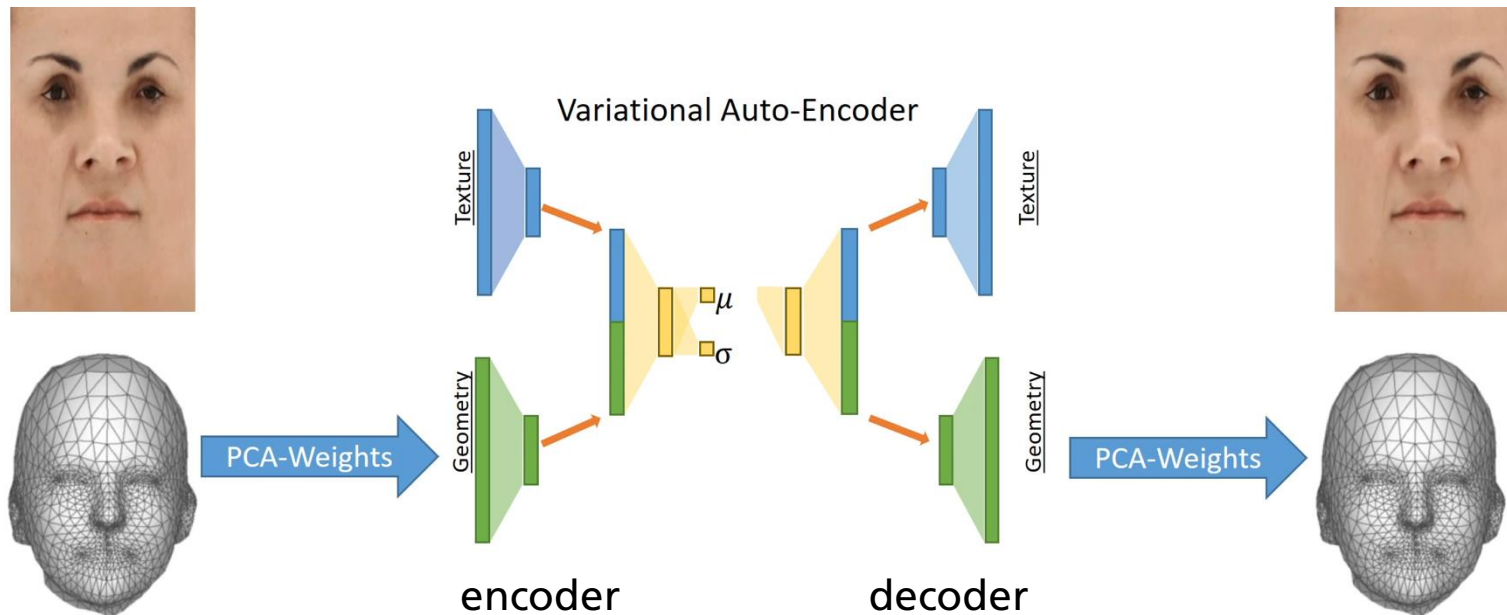
Hybrid face capture & representation learning



W. Paier et al., CVMP 2020
W. Paier et al., IET CV 2020

Neural Face Representation

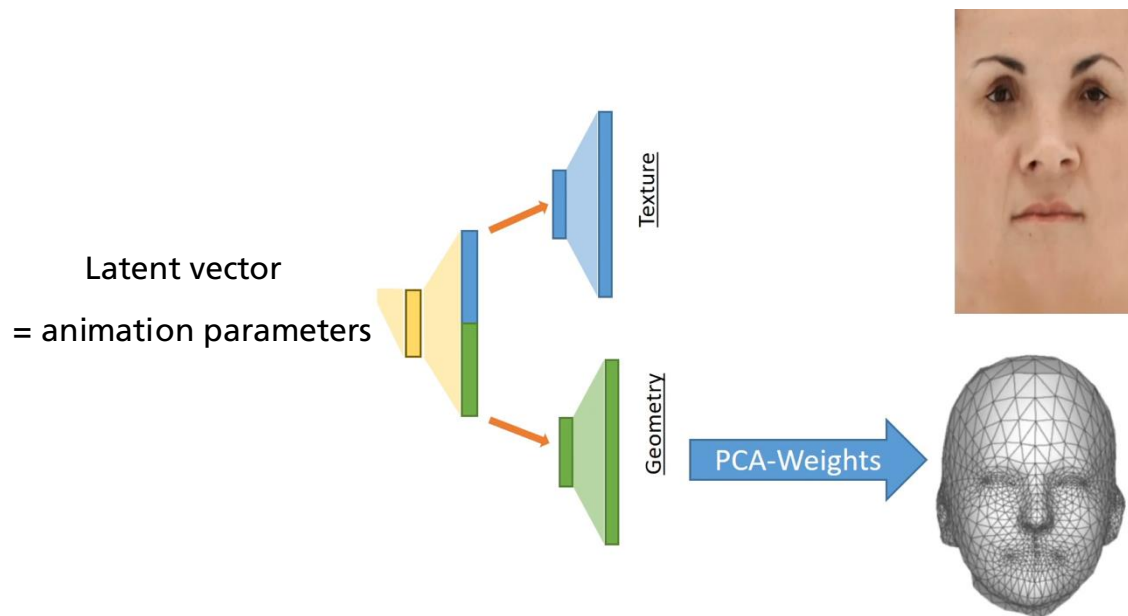
Variational Autoencoder



W. Paier et al., IEEE CGA 2021

Neural Face Representation

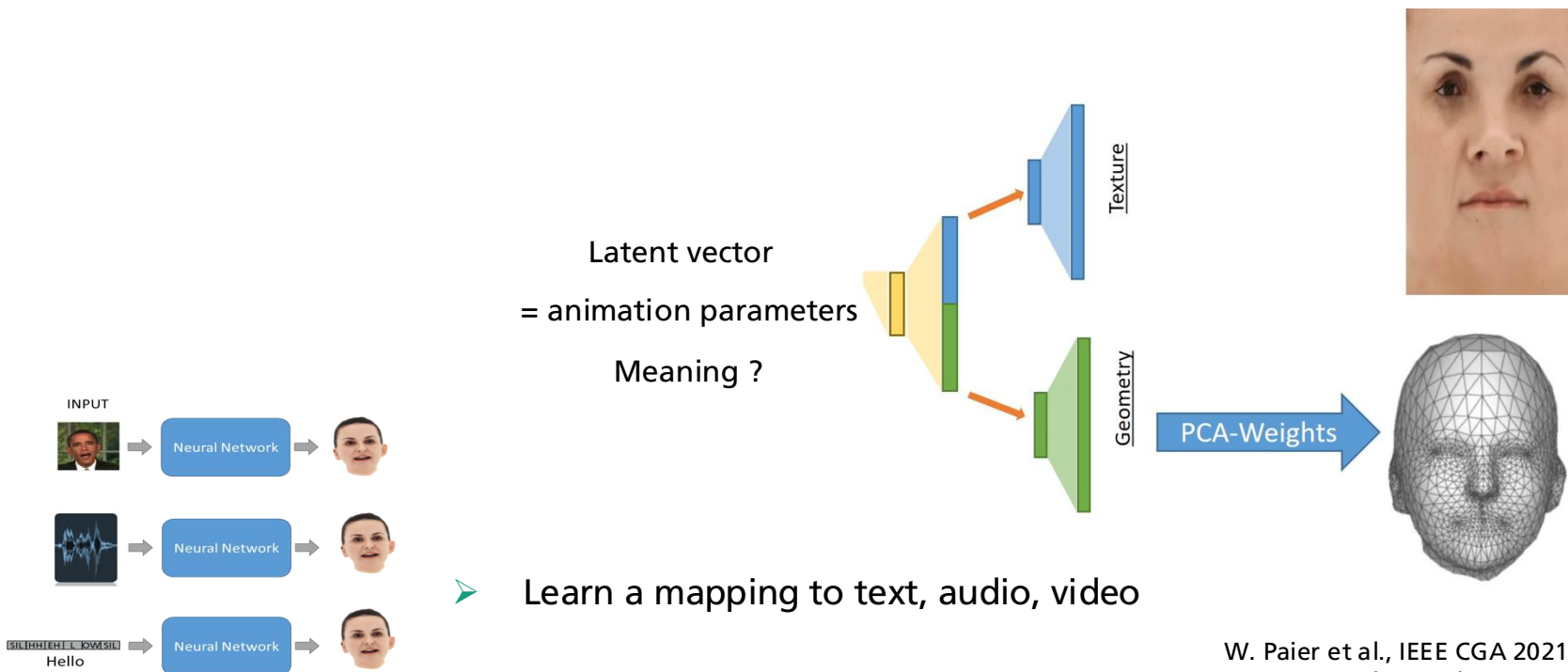
Variational Autoencoder



W. Paier et al., IEEE CGA 2021

Neural Face Representation

Variational Autoencoder



➤ Learn a mapping to text, audio, video

W. Paier et al., IEEE CGA 2021
W. Paier et al., Graphics & Visual Computing

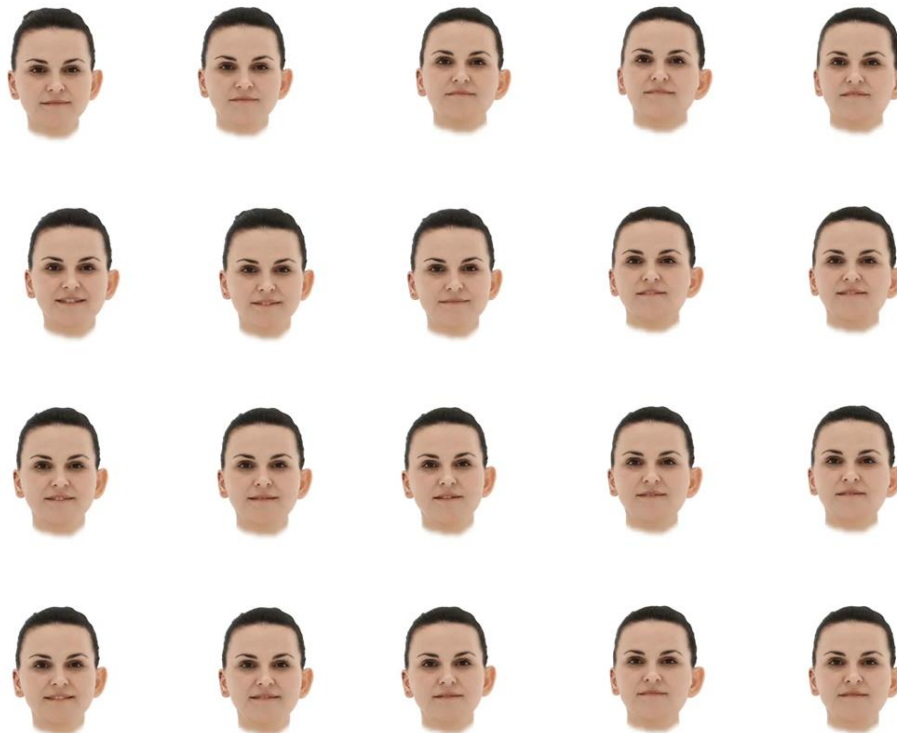
Neural Face Animation

Speech synthesis results



Neural Face Animation

Speech synthesis results – different styles



Animation Control



Driving video



Style 1

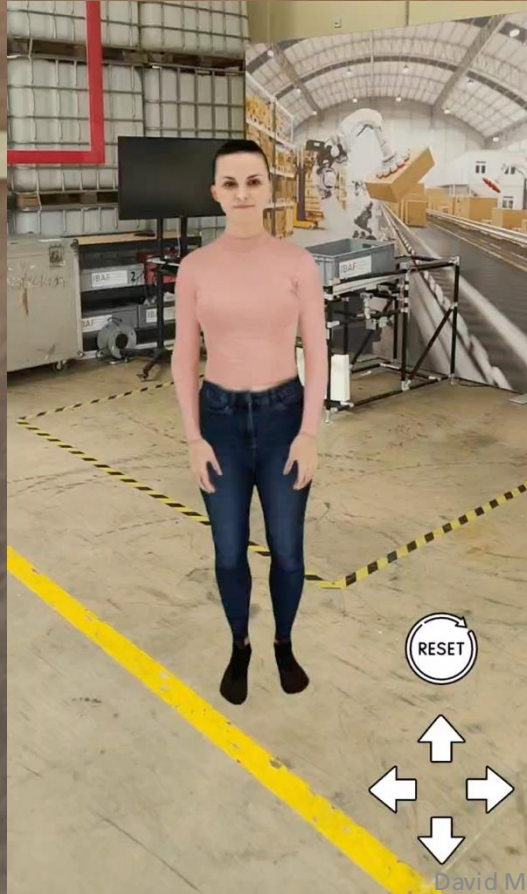


Style 2



Style 3

Integrated Avatars

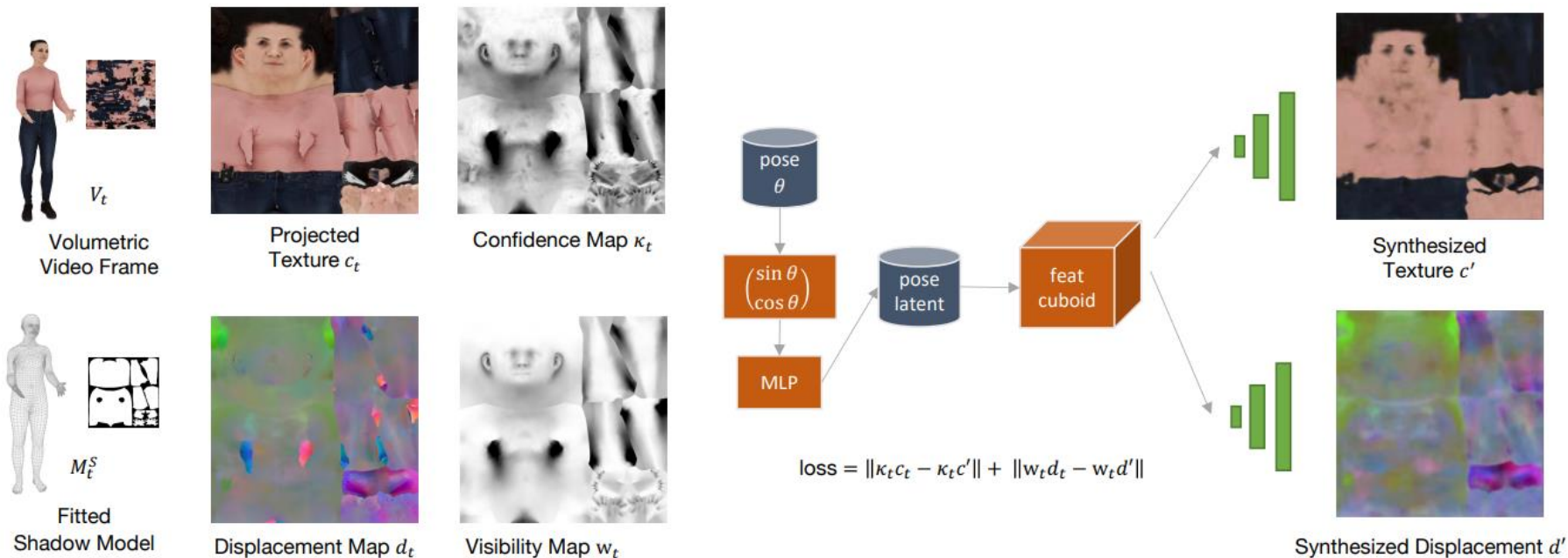


David Moreno – EuroXR

CyberLink
Power

Learning a Generative Model for Virtual Humans

Training a pose-dependent decoder



Morgenstern et al., IEEE VR 2024

Learning a Generative Model for Virtual Humans

Results

synthesized
model

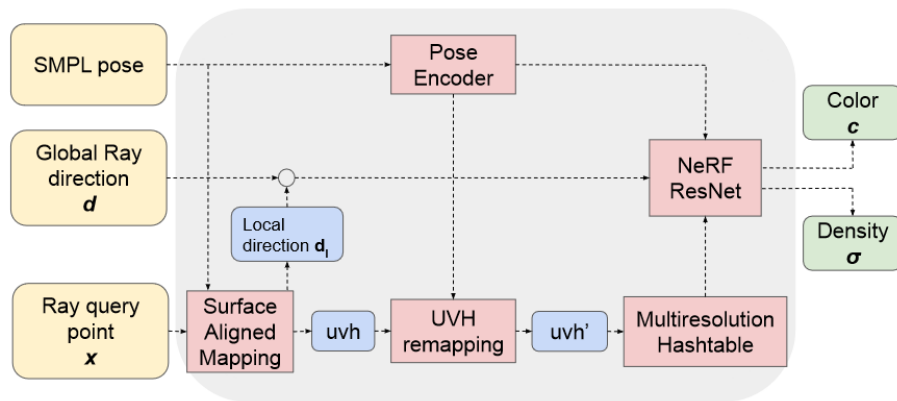


volumetric
video



SMPL Aligned NeRF

[Knoll VISAPP 2024]

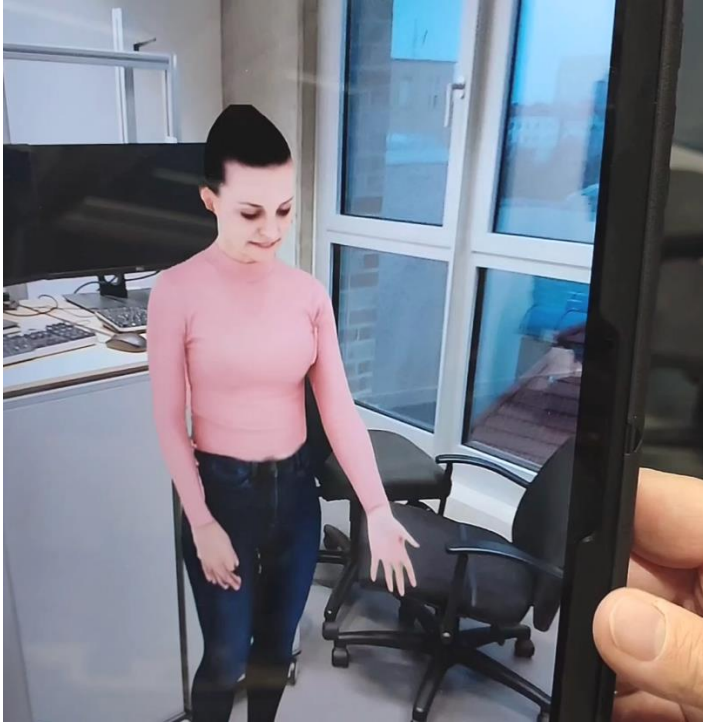


- Extended Surface aligned NeRF [XFM22]
- Mapping of canonical space to uv-height coordinates of SMPL model
- Remapping of coordinates with learned frame embeddings to account for alignment errors
- Can be animated via skeleton to unseen poses



Learning a Generative Model for Virtual Humans

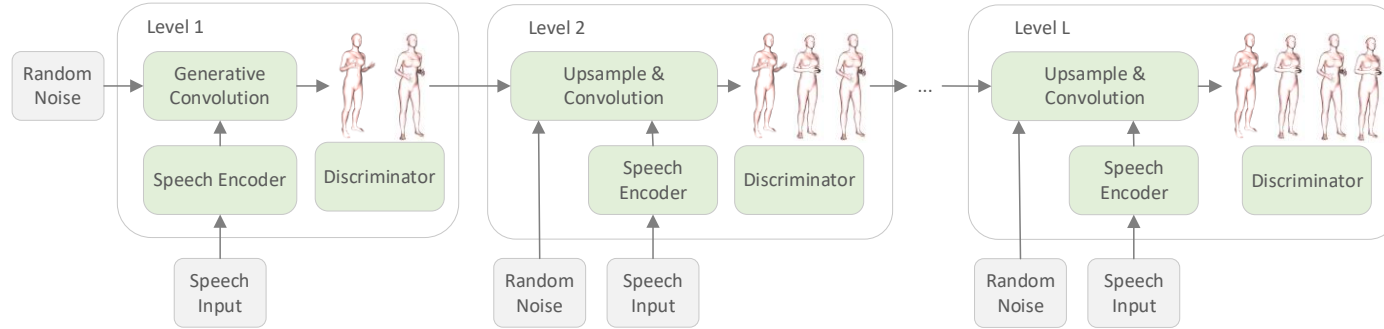
AR Visualization



[Morgenstern IEEE VR 2024]

Human Animation

Learning Co-Speech Gesture Generation

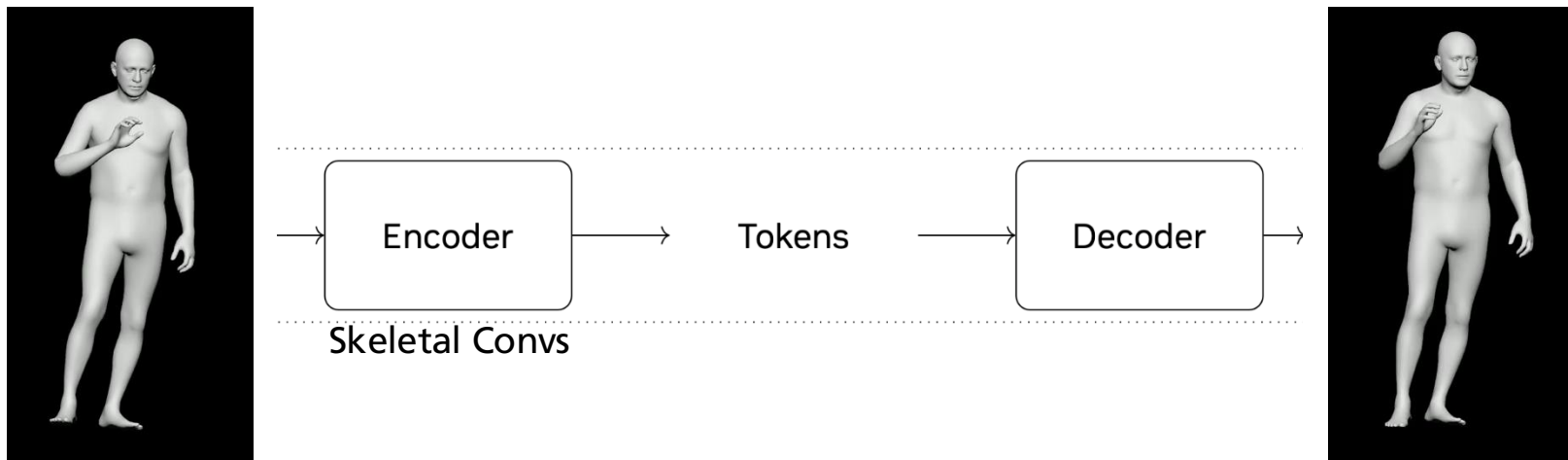


Moreno et al., CVMP 2024

Gesture Animation

New approach

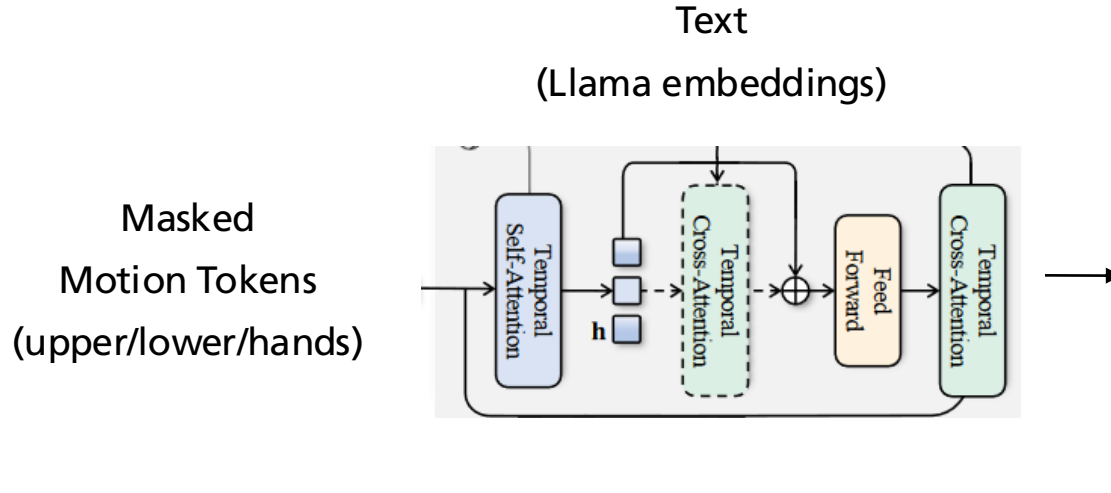
- We include skeletal-aware convolution layers in different motion tokenizers
- Each motion tokenizer learns a compact motion representation of the upper body, lower body and hands



Gesture Animation

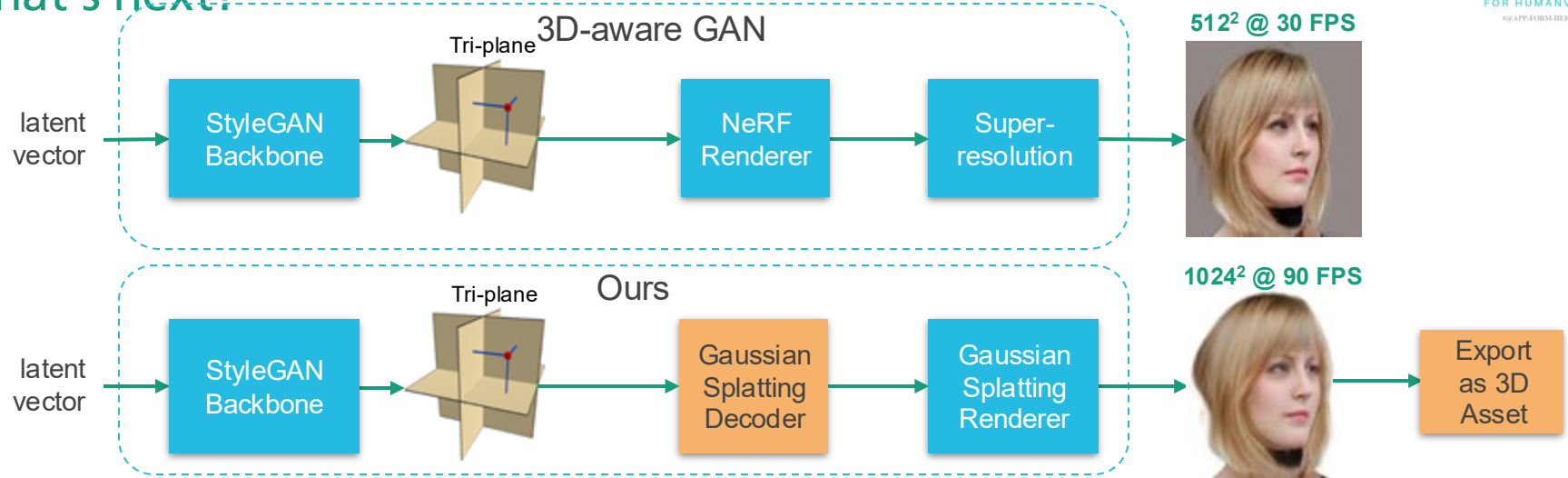
Tokenized Motion

- A transformer network predicts motion tokens based on semantic text embeddings (Llama embeddings)



From Volumetric Video to Realistic Avatars

What's next?



- Gaussian Splatting Renderer, standard rendering possible, e.g. Unity, 150fps
- Novel mapping network synthesizing Gaussian properties
- End to end training without superresolution network

[Barthel CVPR 2024]

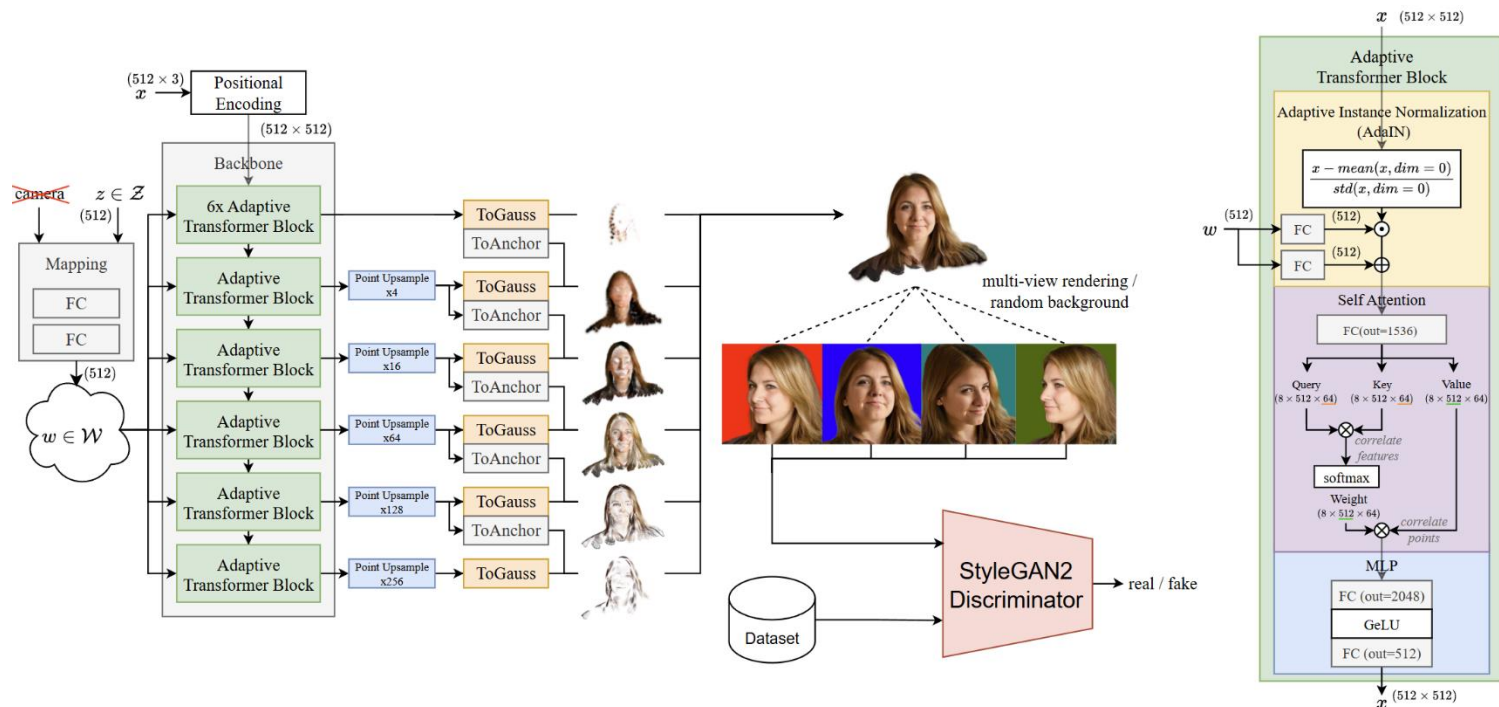
From Volumetric Video to Realistic Avatars

What's next?



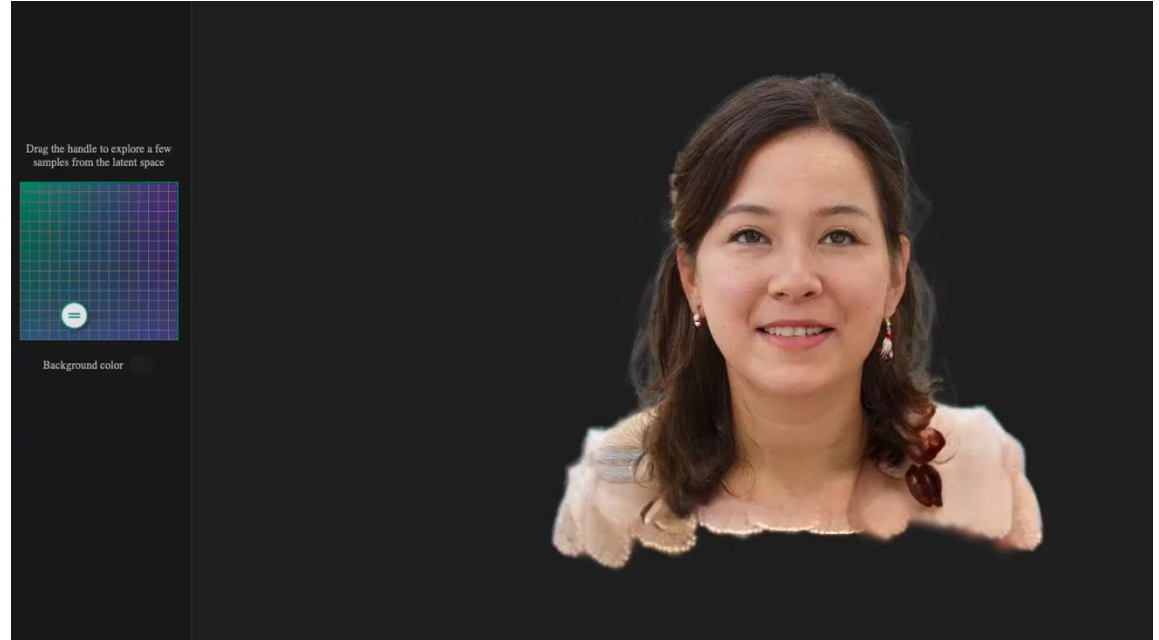
From Volumetric Video to Realistic Avatars

What's next?



F. Barthel et al., Arxiv 2025

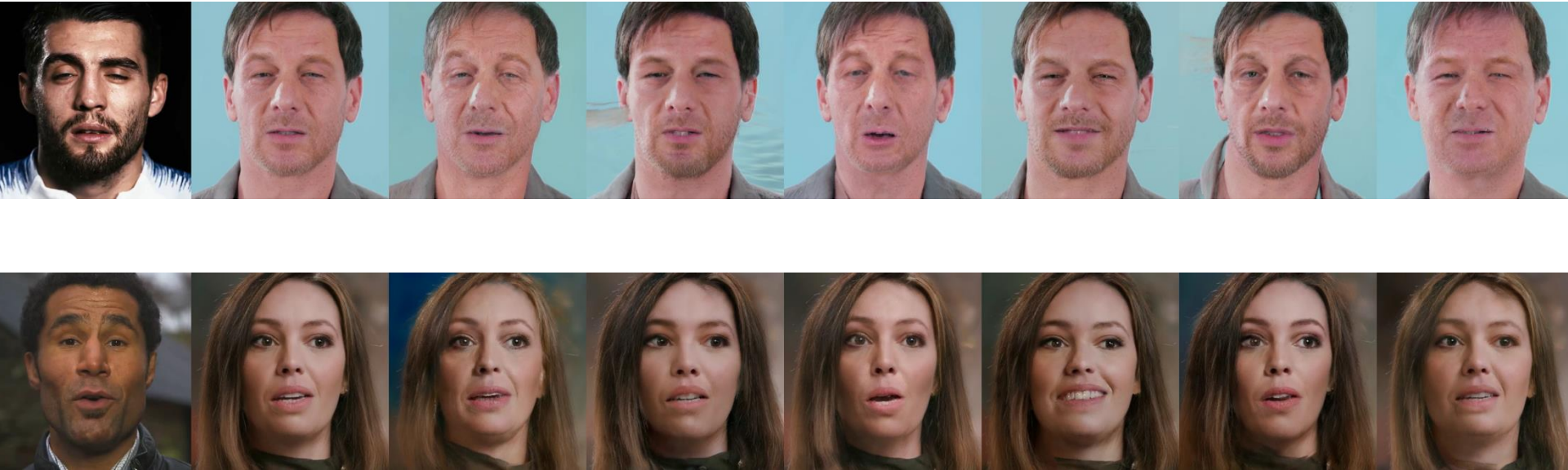
- Gaussian Splatting representation for high quality and real-time rendering
- Easy personalization by GAN structure
- synthesizes arbitrary faces
- GAN inversion adjusts model to 1 or more pictures of a person



Submitted to NeurIPS

F. Barthel et al., Arxiv 2025

Editing of Face Properties Through Latent Space Offsets



Fraunhofer-Institut für Nachrichtentechnik, Heinrich-Hertz-Institut, HHI

WE PUT SCIENCE INTO ACTION.



Contact:

David Moreno
david.moreno@hhi.fraunhofer.de

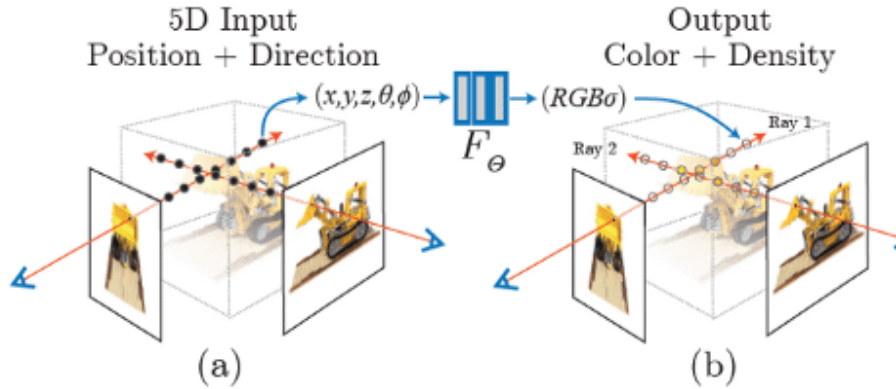


Peter Eisert
peter.eisert@hhi.fraunhofer.de

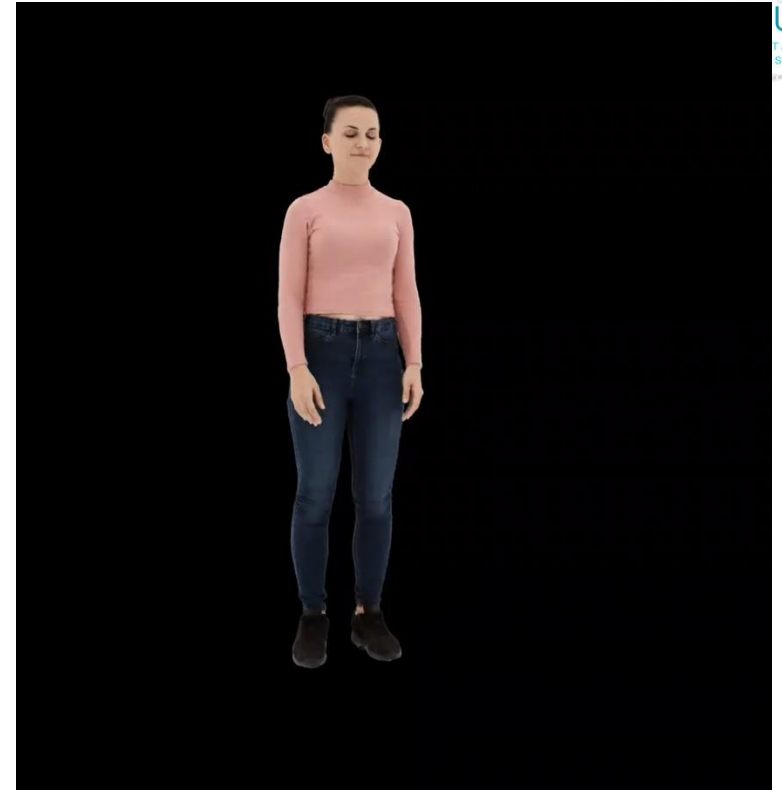
Einsteinufer 37
10587 Berlin

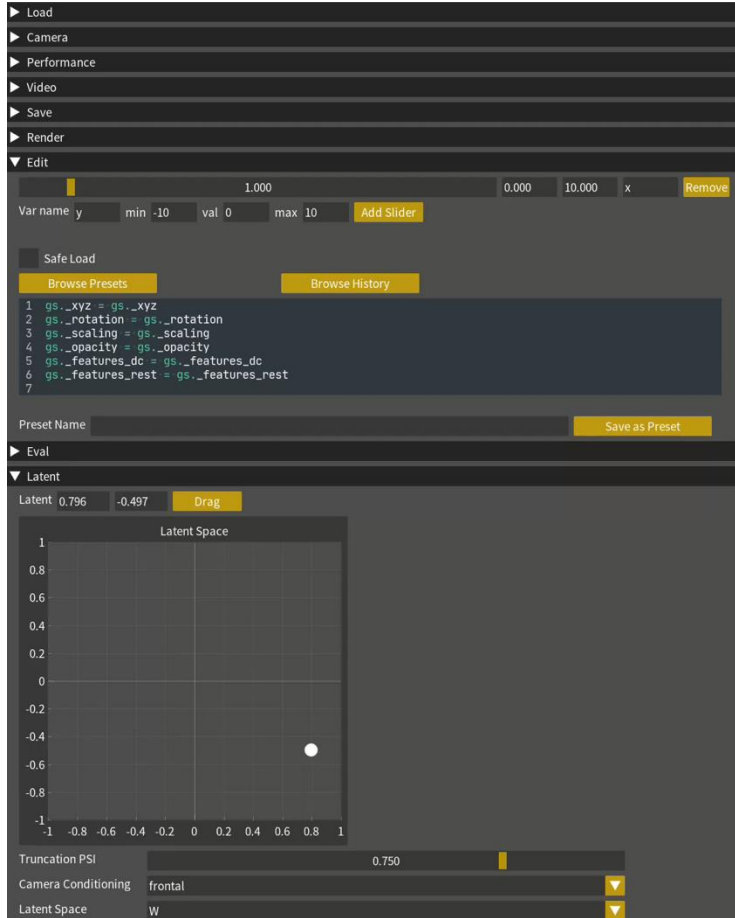


Neural Radiance Fields (NeRF)



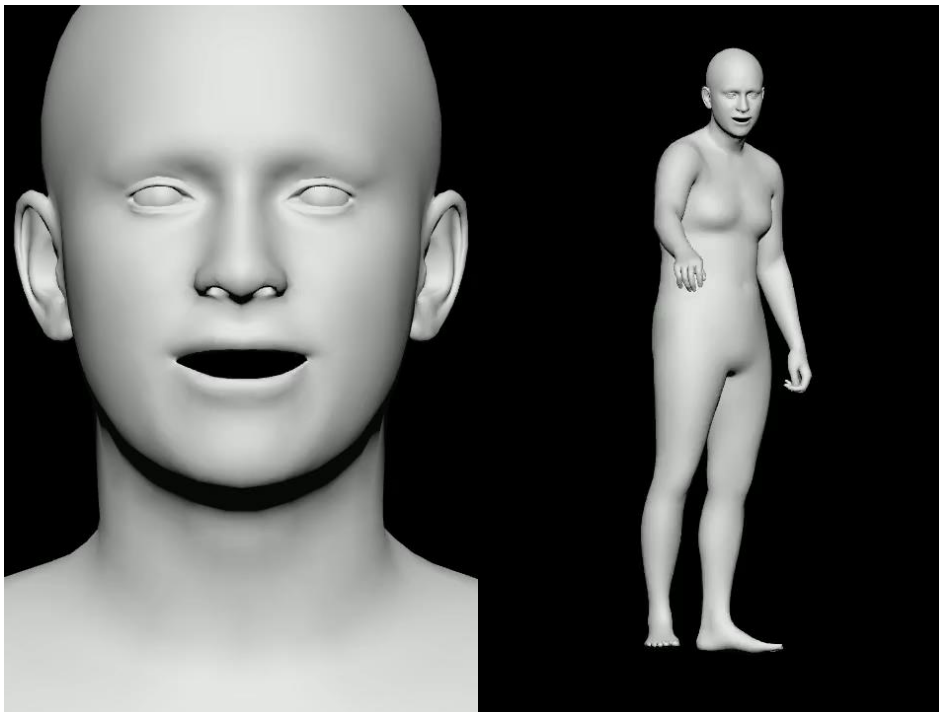
- MLP that reconstructs RGBA for a 3D point x
- Volume rendering by accumulating line of sight
- Enables static 3D object rendering
- Instant NGP for fast training



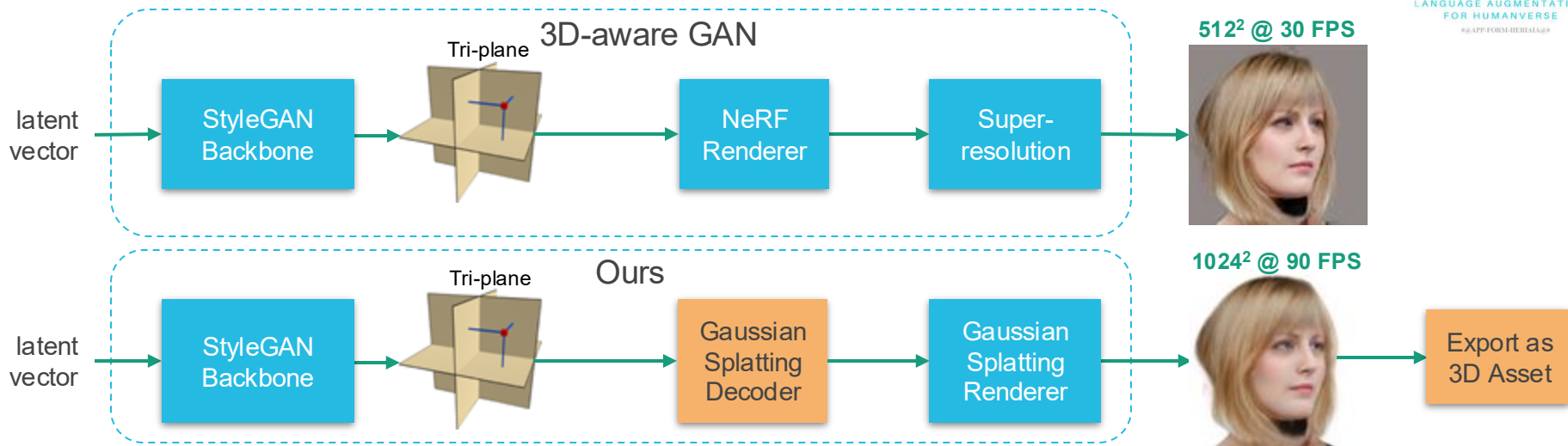


F. Barthel et al., Arxiv 2025

Gesture Animation – Preliminary Results



GAN Synthesis of Gaussian Splatting Faces



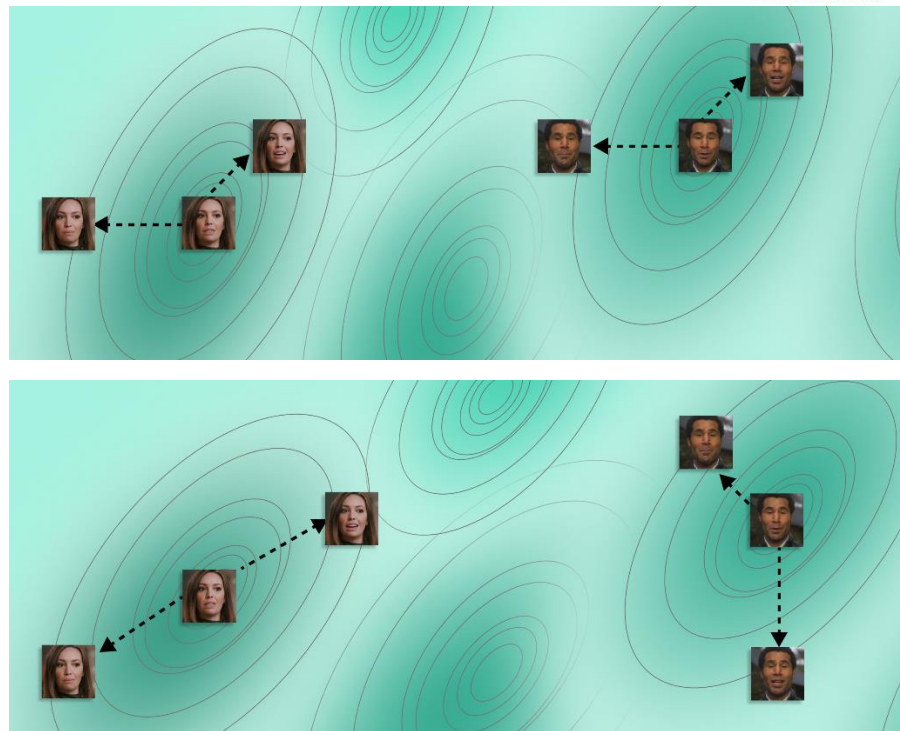
- Presented @ CVPR 2024
- Mapping network that creates properties of Gaussian Splats (size, color, opacity, orient.)
- Fast rendering with standard libraries
- No need for super-resolution step

Editable 3D Renderings



Lip Animation Through GAN Inversion

- GAN inversion for frames of a video sequence
- Create latent vectors defining motion space
- For different individuals, mimic directions can differ in latent space
- Retraining of generator to reshape GAN space such that same motion of different individuals is aligned



Expression Transfer



Compression of Gaussian Splatting Scenes

- Gaussian Splatting scenes show high quality results but are very large in size
- Efficient coding and storage necessary
- Our codec compresses exploiting
 - high dimensional sorting
 - scene refinement
- Size reduction of a factor 15-25 at almost same quality

[Morgenstern et al. ECCV 2024]

3D Gaussian Splatting

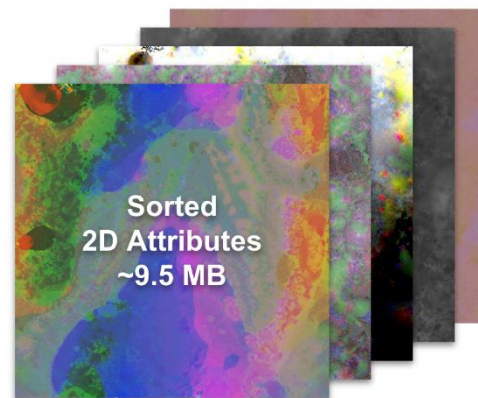
Millions of Gaussians
~174 MB

Index	Position	Color	Scale	Opacity	...
0	(0.148, 0.601, 0.294)	(0.871, 0.917, 0.886)	(0.499, 0.639, 0.471)	0.722	...
1	(0.481, 0.319, 0.554)	(0.983, 0.799, 0.379)	(0.576, 0.653, 0.913)	0.912	...
2	(0.773, 0.833, 0.396)	(0.207, 0.164, 0.421)	(0.693, 0.551, 0.672)	0.074	...
3	(0.838, 0.267, 0.669)	(0.475, 0.478, 0.521)	(0.192, 0.459, 0.771)	0.667	...
4	(0.135, 0.054, 0.808)	(0.308, 0.882, 0.829)	(0.394, 0.532, 0.026)	0.503	...
5	(0.148, 0.601, 0.295)	(0.871, 0.917, 0.886)	(0.499, 0.639, 0.471)	0.722	...
...	...	...	...	...	...
2.5e7	(0.027, 0.134, 0.02)	(0.204, 0.669, 0.816)	(0.877, 0.553, 0.416)	0.587	...



PSNR 24.89dB

Ours



PSNR 24.83dB

Fraunhofer-Institut für Nachrichtentechnik, Heinrich-Hertz-Institut, HHI

WE PUT SCIENCE INTO ACTION.

Contact:

Peter Eisert
peter.eisert@hhi.fraunhofer.de
+49 30 31002 614

Einsteinufer 37
10587 Berlin

