

Dress-Up: Generating Animatable Clothed 3D Humans via Latent Modeling of 3D Gaussian Texture Maps



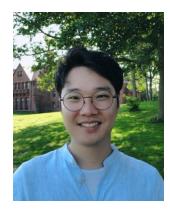
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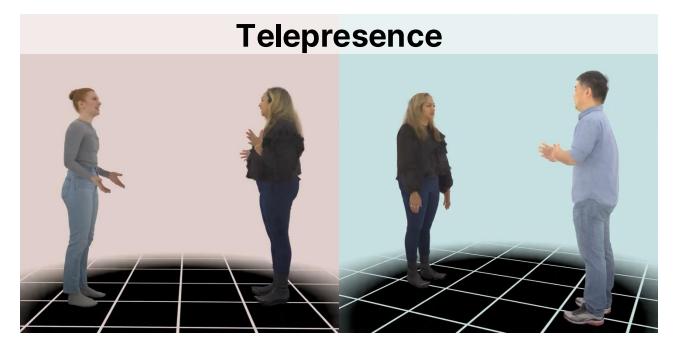








Photorealistic, animatable, clothed 3D humans

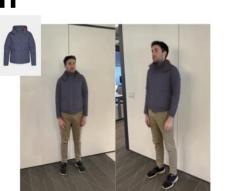


3D virtual try-on









Synthetic data

Gaming & content creation





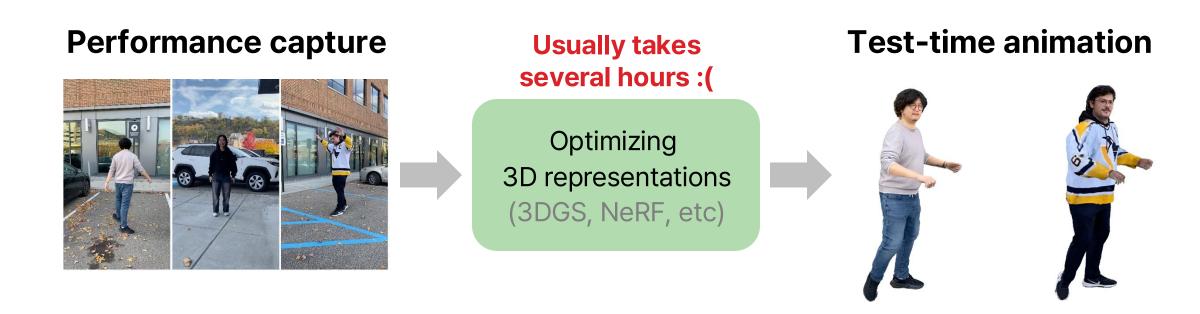
Agrawal et al., "Seamless Interaction: Dyadic Audiovisual Motion Modeling and Large-Scale Dataset," arXiv preprint 2506.22554. Patel et al., "AGORA: Avatars in Geography Optimized for Regression Analysis," CVPR 2021. Cao et al., "GS-VTON: Controllable 3D Virtual Try-on with Gaussian Splatting," arXiv preprint 2410.05259

Conventional avatar creation pipeline

Stage 1) Real-world performance captures

Stage 2) Optimize animatable 3D representations

Stage 3) Animate avatars at test-time



Our goal: Scalable avatar creation pipeline

We build a feed-forward, unconditional generative model for creating **Photorealistic**, **Animatable**, **Clothed** 3D human avatars



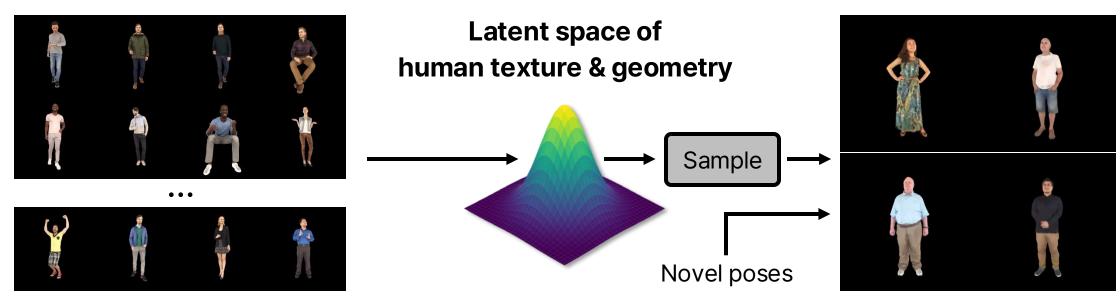
Dress-Up: Generative modeling of 3D humans

Learn latent space from real human scans

- Build a compact latent space of <u>diverse human</u>
 <u>textures and geometries</u>
- Encoder-free latent space modeling, i.e., autodecoding, for efficient training

Learn to sample from the latent space

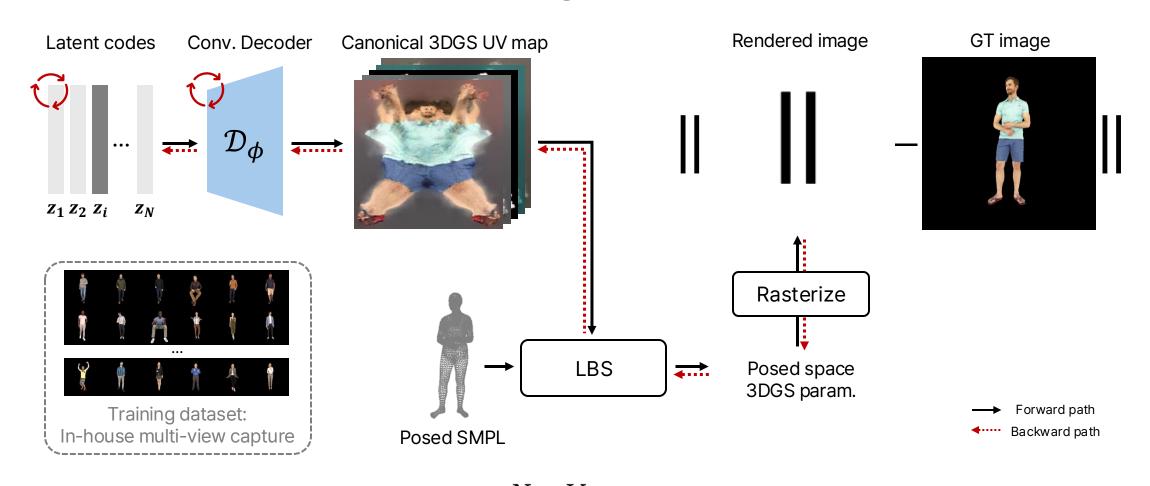
- Train a <u>latent diffusion model (LDM)</u> using the built latent codes as a dataset
- LDM maps randomly sampled noise into the valid latent codes, later decoded into 3DGS avatars



Multi-view real-human scans

Generated 3DGS human avatars

Latent space modeling of 3D clothed humans



Training objective

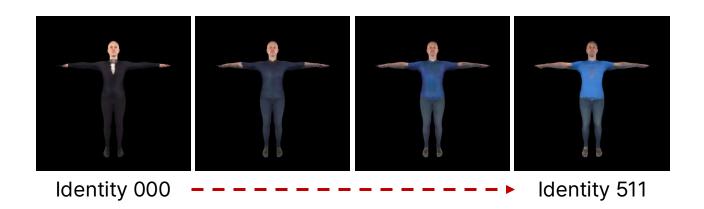
$$\mathcal{L}_{\mathrm{AD}} = \sum_{i=1}^{N} \sum_{v=1}^{N} \lVert \mathbf{I}_{i,v} - \mathcal{R}\left(\boldsymbol{ heta}, \mathcal{D}_{\phi}(\mathbf{z}_i)
ight)
Vert_1 + \lVert \mathbf{z}_i
Vert_2^2$$
Recon. loss Regularization

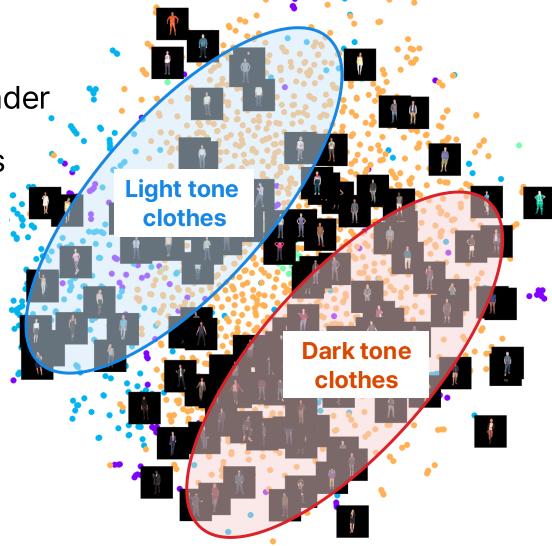
Learned latent space of 3D clothed humans

We trained a latent space with 2K identities

The learned latent space inherently finds
 the semantic clusters, e.g., cloth colors, gender

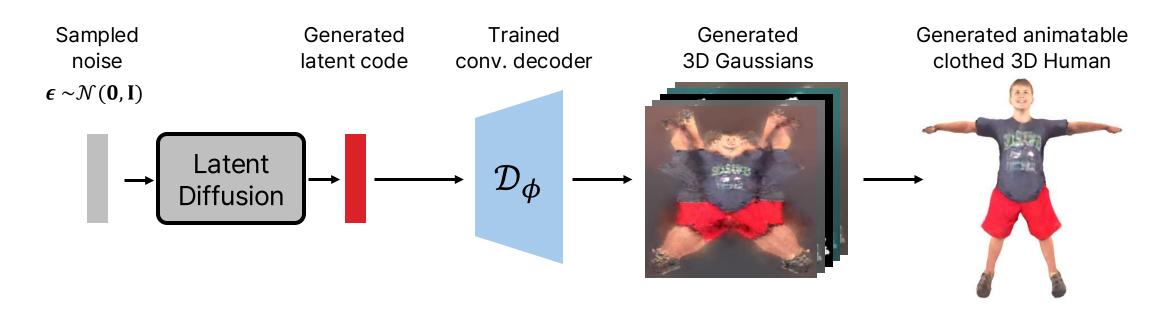
 Interpolation across the learned latent codes are possible → Already can create new IDs





Latent diffusion for generating 3DGS avatars

- Train a latent diffusion model (LDM) using the learned per-ID latent codes
- LDM learns the complex data manifold of the clothed canonical 3DGS avatars
- LDM can sample novel codes, and these codes will be decoded into
 3DGS parameters in SMPL UV coordinates



Randomly generated canonical 3DGS avatars



Animating the generated avatars at test-time



Current limitations & future directions

- Lack of high-quality multi-view human scan datasets
 - → Can benefit from synthetic image datasets, paired with 3D skinned meshes^[1]
- Do not model the dynamic properties of the clothes
 - → Can extend with Physics-integrated 3D Gaussians^[2]
- Lack of appearance controllability (e.g., text control)
 - → Can add text or other modalities as a condition for LDM, to add controllability (we have some early results!)



Future direction: Trinity for virtual 3D humans

- A virtual 3D human, which satisfies the following:



- 2) **Interact seamlessly** with the 3D objects and scenes
- 3) Make lifelike **emotional expressions**

More ways to go!

Could pave the way toward the next-generation of human-centric content creation.

Photorealism



Physical interaction



Emotional expression







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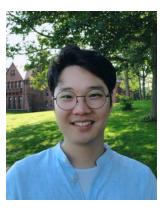
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