



Task & Motivation

Inputs

Outputs

Task: for outfit personalization, combine User's identity with Model's outfit, preserve high-fidelity 3DGS appearance

Why is this hard?

- 2D-to-3D try-on can degrade garment texture and view consistency [1].
- Layered garment & body transfer can introduce intersections [2].

Failure example of 2D-to-3D try-on [1]: red dotted boxes show view-inconsistent garment details.

Motivation & Key Idea

- To avoid view inconsistency and intersection artifacts, **compose** reusable avatar's **head & body** instead of regenerating outfits.
- To **preserve identity** beyond the face, match the clothed body to the User's **physique** and **skin tone**.
- To **reduce hallucinated** changes in high-quality regions, **refine only seams and reshaping artifacts**.

Contributions

- Cross-avatar **composition** for 3DGS outfit personalization.
- GSReshape for body-shape-aware **clothed-body retargeting**.
- SeamFix + FullbodyFix for localized and full-body **artifact repair**.

Method

Mesh-based Avatar Reconstruction (§3.1)

Body Alignment & Reshaping (§3.2)

3D Gaussian Update (§3.3)

Reference: SMPL in user's physique and model's pose

Training Artifact Refiners

Mesh-based Avatar Reconstruction

Reconstruct mesh based 3DGS avatars from multi-view images [3,4], with semantic segmentation and SMPL fitting.

Cross-Avatar Composition

Align the User's head & neck to the Model pose, then **combine** with Model's clothed body. **Transfer User's skin-tone** to body for identity preservation.

GSReshape

Retarget the clothed body to the user's physique by **jointly deforming** the mesh and the attached Gaussians [5].

Seam/FullbodyFix

Repair **seam and reshaping artifacts**, then use the refined views to update the composed 3DGS avatar [6].

Key Takeaways:

- Cross-avatar compositions **lack pixel-aligned GT** for training artifact refiners.
- Two head-swaps** synthesize pairs: A→B introduces artifacts; B→A returns to A's layout for GT alignment.
- Seam/FullbodyFix learn artifact removal; refined multi-view renders then update the composed 3DGS avatar.

Results

Quantitative Results on THuman2.0 [9]

Method	Outfit DINO ↑ [7]	Head + Neck DINO ↑ [7]	Warp RMSE ↓ [8]	User Preference ↑
VTON360 [1]	0.633	0.786	0.0276	8.7%
TIP-Editor [10]	N/A	0.356	0.0388	2.6%
Ours	0.883	0.818	0.0175	88.7%

Additional Qualitative Results

Ablations

Stress Test: Cross-Gender Pair

Key Takeaways

- Compose first, repair locally.**
- Mesh-based Gaussians make **geometric identity transfer** practical.
- Diffusion refinement acts as **cleanup**, instead of full outfit generation.

Limitations / Future work

- Requires high-quality mesh-based Gaussian avatars as input.
- FullbodyFix is currently triggered by visual inspection.
- Animatable transfer remains future work.

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References

[1] He et al. VTON 360: High-Fidelity Virtual Try-On from Any Viewing Direction. CVPR, 2025.
 [2] Lin et al. LayGA: Layered Gaussian Avatars for Animatable Clothing Transfer. SIGGRAPH, 2024.
 [3] Kerbl et al. 3D Gaussian Splatting for Real-Time Radiance Field Rendering. ACM TOG, 2023.
 [4] Shao et al. SplattingAvatar: Realistic Real-Time Human Avatars with Mesh-Embedded Gaussian Splatting. CVPR, 2024.
 [5] Huang et al. Intersection-Free Garment Retargeting. SIGGRAPH, 2025.
 [6] Wu et al. DIFIX3D+: Improving 3D Reconstructions with Single-Step Diffusion Models. CVPR, 2025.
 [7] Oquab et al. DINOv2: Learning Robust Visual Features without Supervision. TMLR, 2024.
 [8] Asim et al. MET3R: Measuring Multi-View Consistency in Generated Images. CVPR, 2025.
 [9] Yu et al. Function4D: Real-Time Human Volumetric Capture from Very Sparse Consumer RGBD Sensors. CVPR, 2021.
 [10] Zhuang et al. TIP-Editor: An Accurate 3D Editor Following Both Text-Prompts and Image-Prompts. ACM TOG, 2024.