



SurFhead

Affine Rig Blending for Geometrically Accurate 2D Gaussian Surfel Head Avatars

ICLR 2025

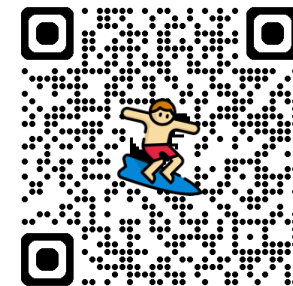
Jaeseong Lee^{1*} Taewoong Kang^{1*} Marcel C. Bühler² Min-Jung Kim¹
Sungwon Hwang¹ Junha Hyung¹ Hyojin Jang¹ Jaegul Choo¹

* Indicates equal contribution

summertight.github.io/SurFhead



ICLR
International Conference On
Learning Representations



Teaser



Geometrical Head Avatars

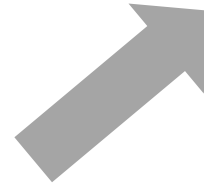


High Fidelity Normals & Depths

Geometrical Head Avatars



High Fidelity Normals & Depths



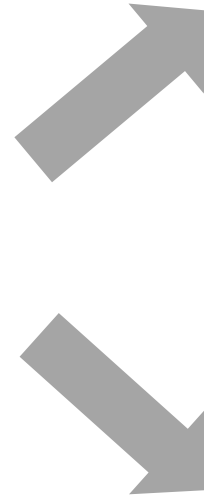
1 Relighting



Geometrical Head Avatars



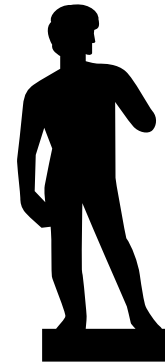
High Fidelity Normals & Depths



1 Relighting

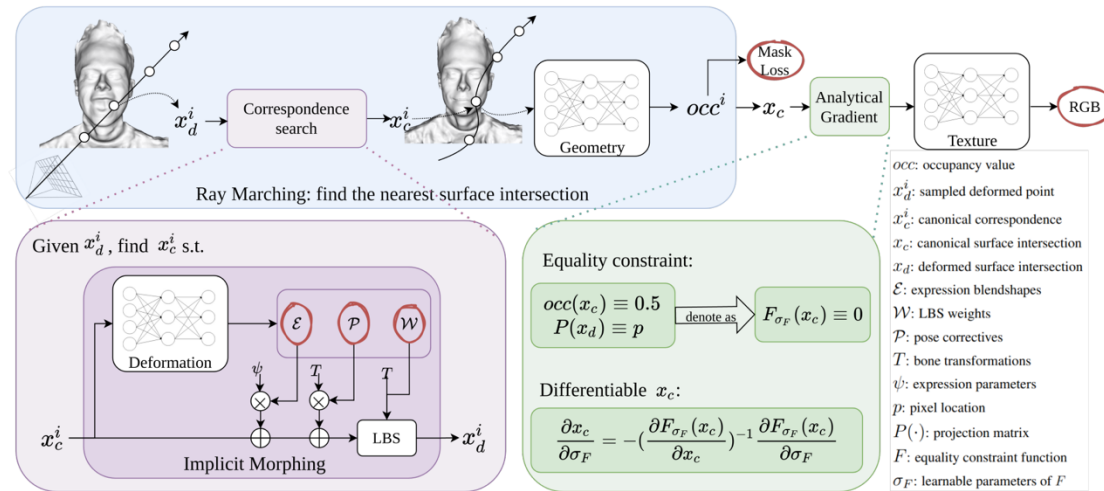


2 Meshing

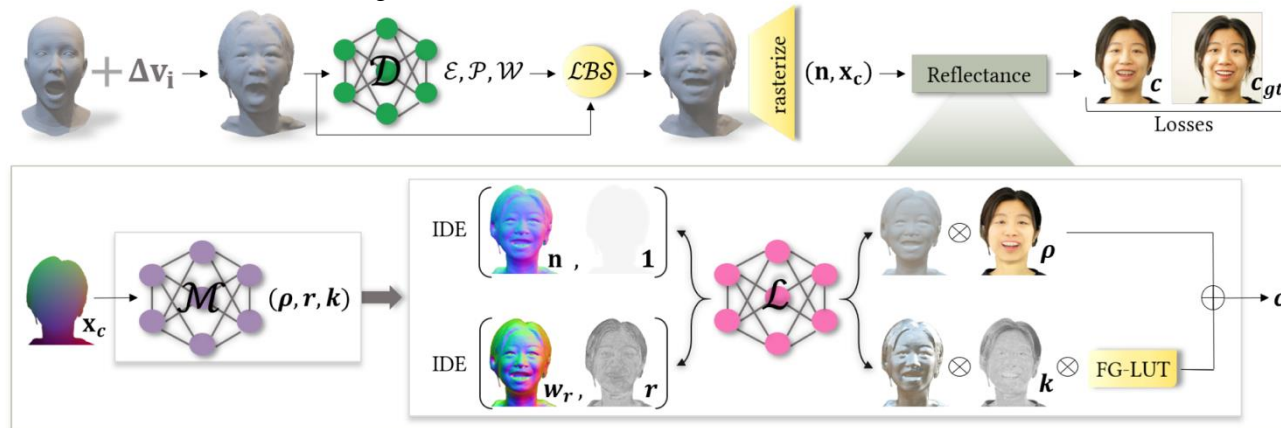


Geometrical Head Avatars – Previous art

IMAvatar (Zheng et al., CVPR 2022)

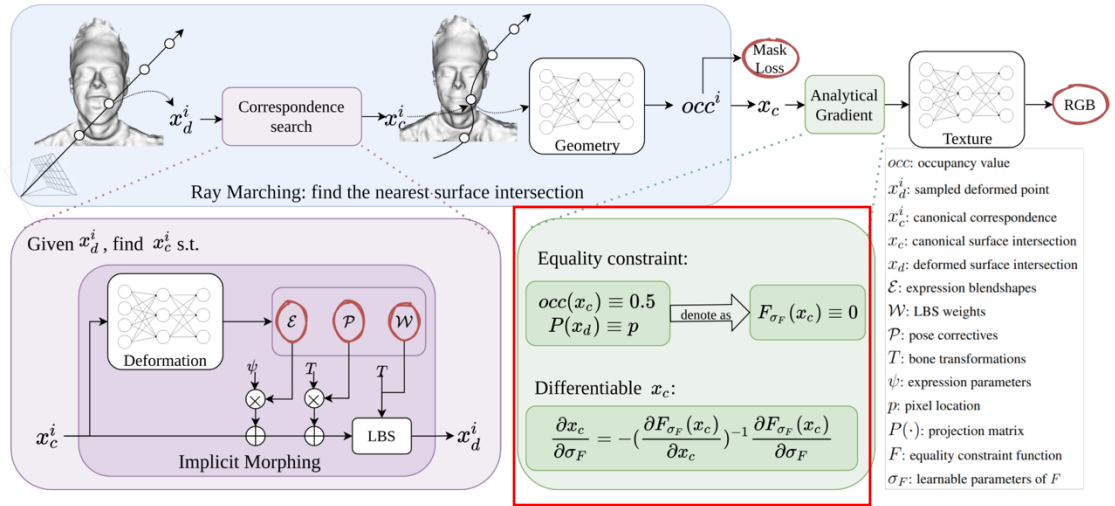


FLARE (Bharadwaj et al., TOG 2023)



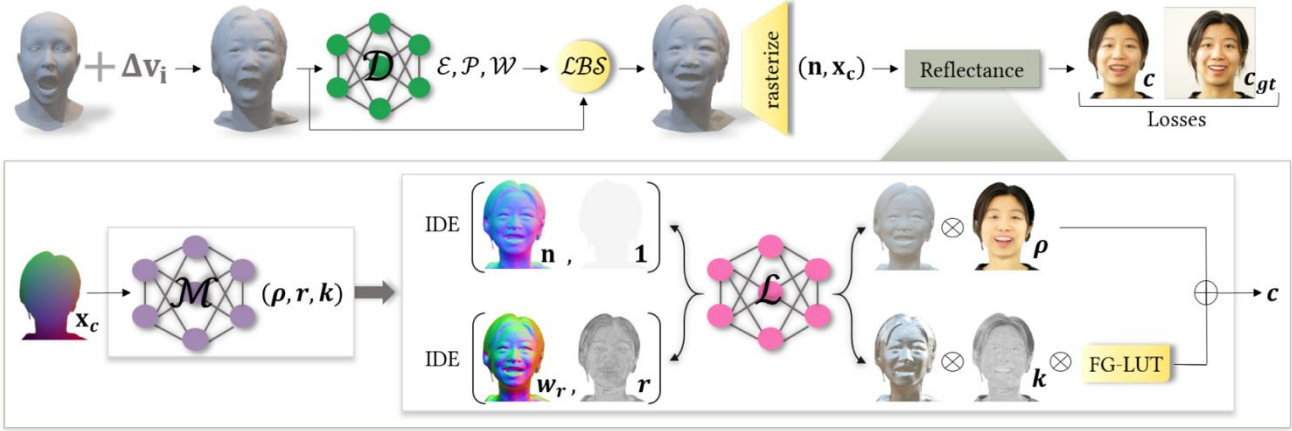
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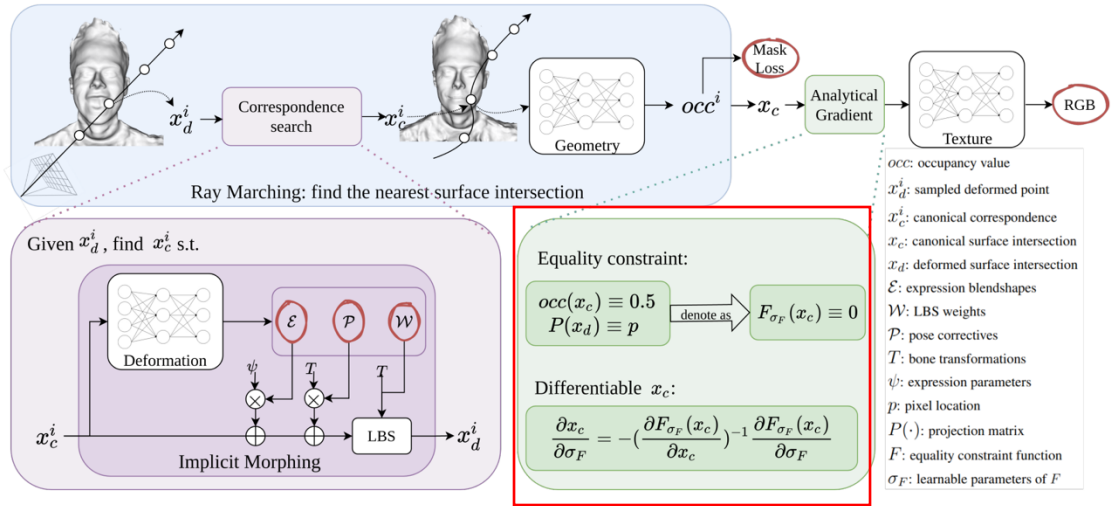
⊗ Exhaustive Numerical Surface Search

FLARE (Bharadwaj et al., TOG 2023)



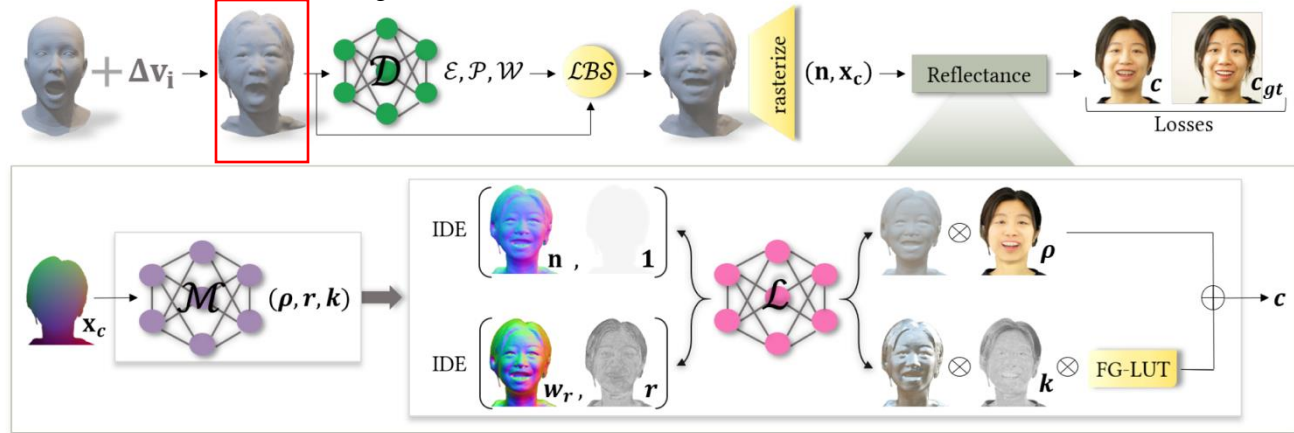
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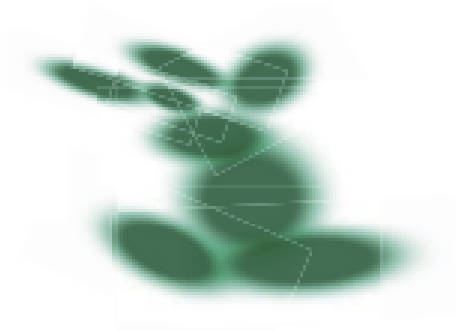
FLARE (Bharadwaj et al., TOG 2023)



⊗ Unstable Training-time Remeshing



How to design 'geometrical' & 'dynamic' **Gaussian** head avatars?



3D Gaussians

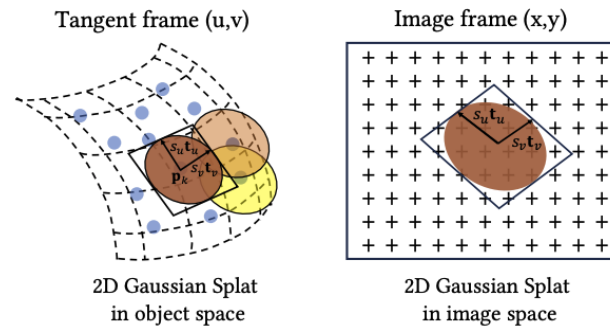
3DGS (Kerbl et al., SIGGRAPH 2023)



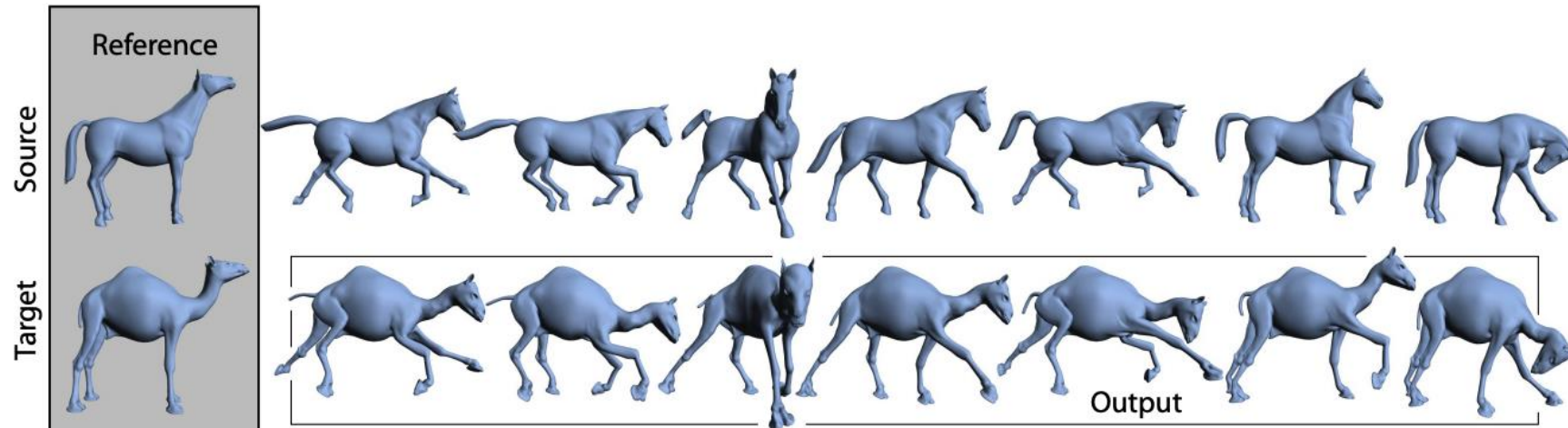
How to design 'geometrical' & '**dynamic**' Gaussian head avatars?

To achieve accurate normal and depth, **we utilize 2D surfels**

Also accurate rigging states, inspired by DTF, **rigging Gaussians with Jacobian gradients.**



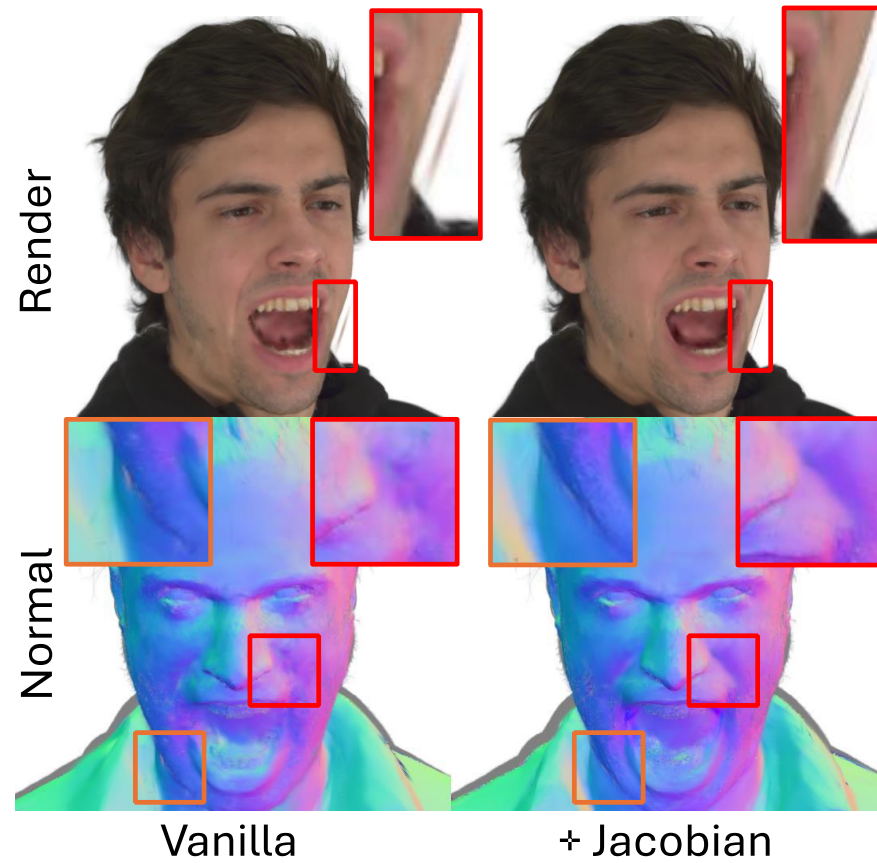
2DGS (SIGGRAPH 2024, Huang et al.)



DTF (SIGGRAPH 2000, Sumner and Popovic)

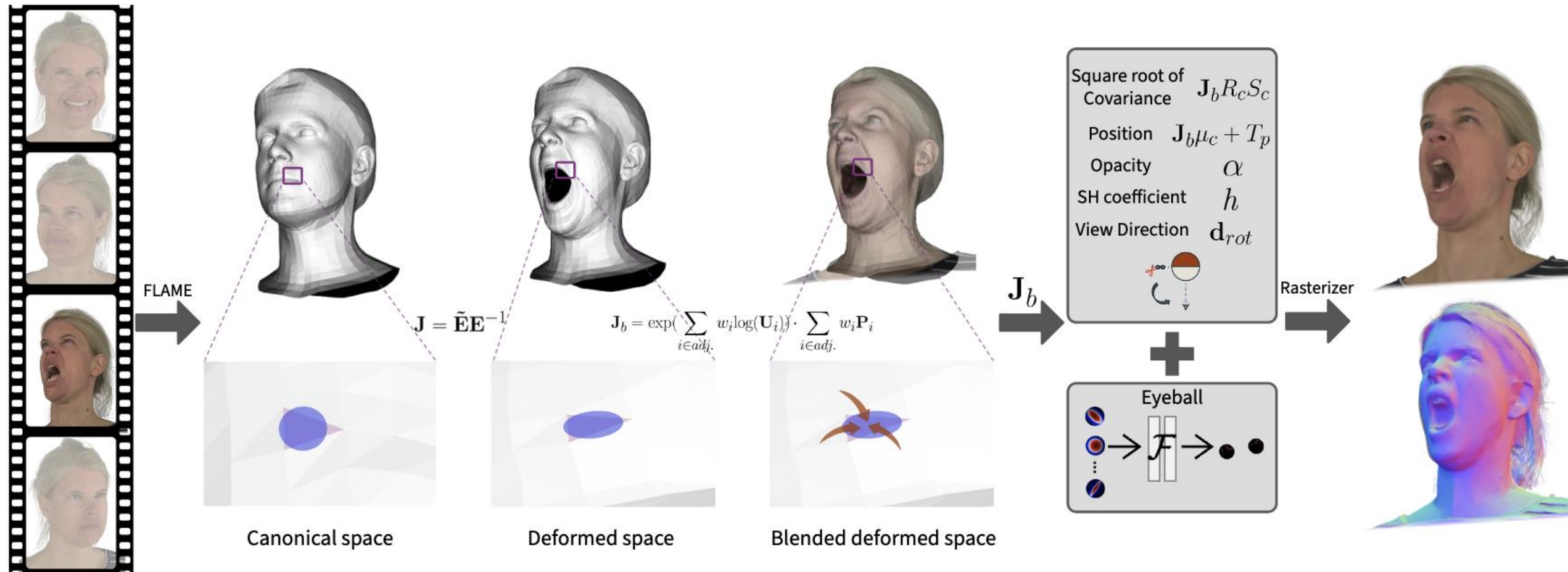
However,

Only using 2DGS and Jacobian Gradients lead to **floating artefacts** and **ambiguous normal**.



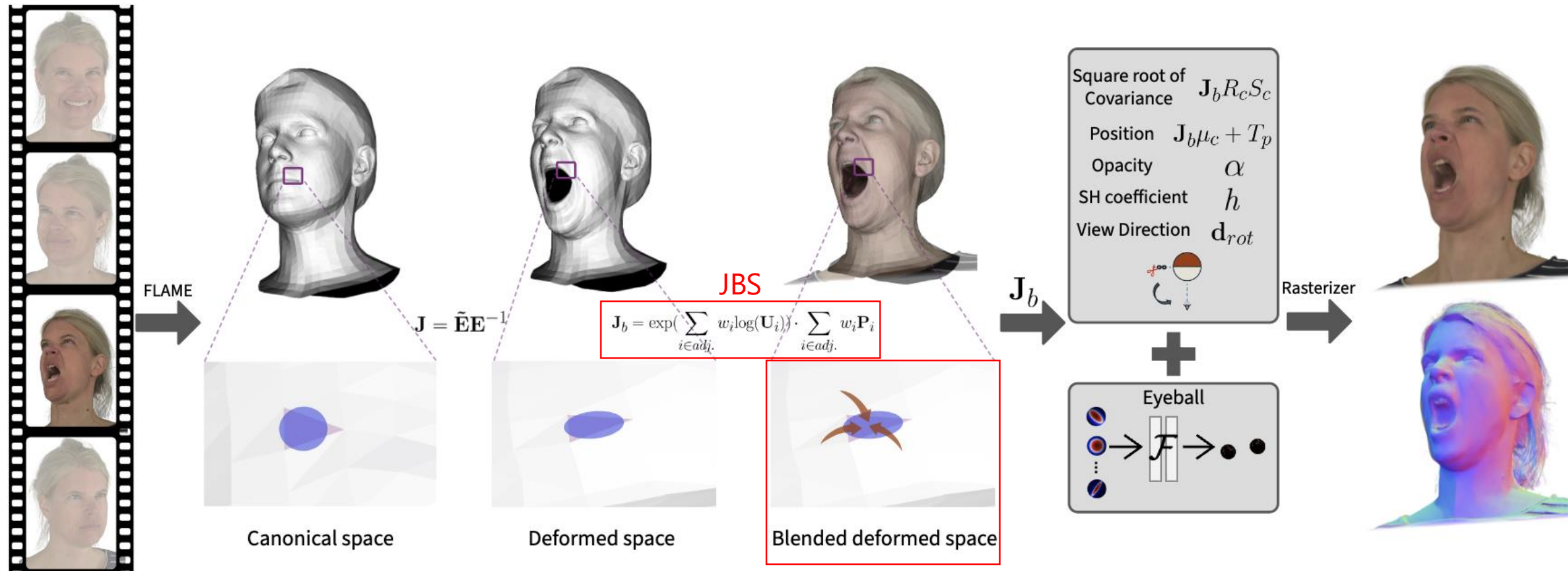
To handle these,

We propose Jacobian Blend Skinning (JBS).



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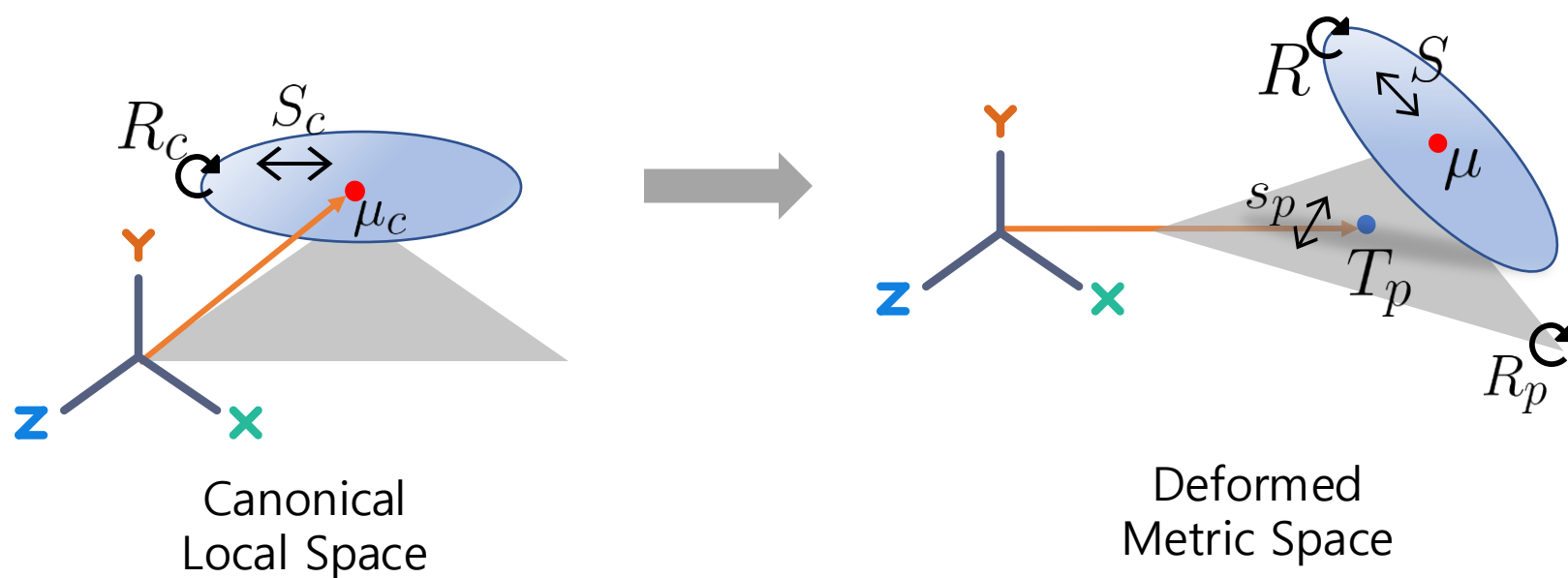
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Mesh Binding Inheritance

We follow 3DMFM binding inheritance strategy of GaussianAvatars (CVPR 2024, Qian et al.)

$$R = R_p R_c, \mu = s_p R_p \mu_c + T_p, S = s_p S_c$$



Similarity to Jacobian Transforms

Following DTF, we replace the GaussianAvatars' similarity transforms to Jacobian transforms.

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$$\Sigma^{1/2} = s_p R_p R_c S_c$$

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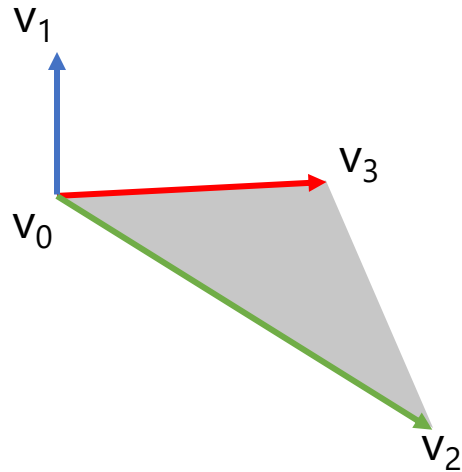
$$\Sigma^{1/2} = \boxed{s_p R_p} R_c S_c$$



$$\Sigma^{1/2} = \boxed{J} R_c S_c$$

Jacobian Deformation Transfer

Jacobian can represent stretch and shear transforms unlike similarity transforms.

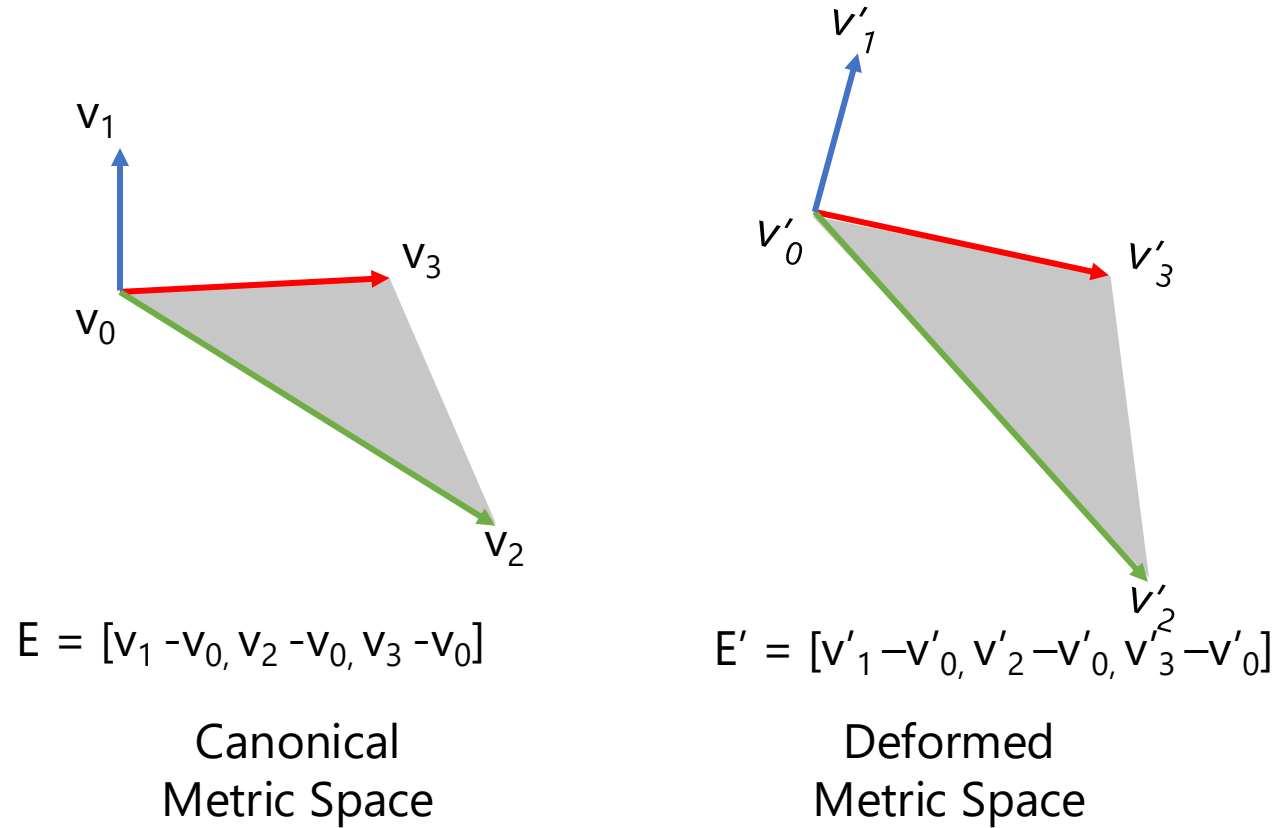


$$E = [v_1 - v_0, v_2 - v_0, v_3 - v_0]$$

Canonical
Metric Space

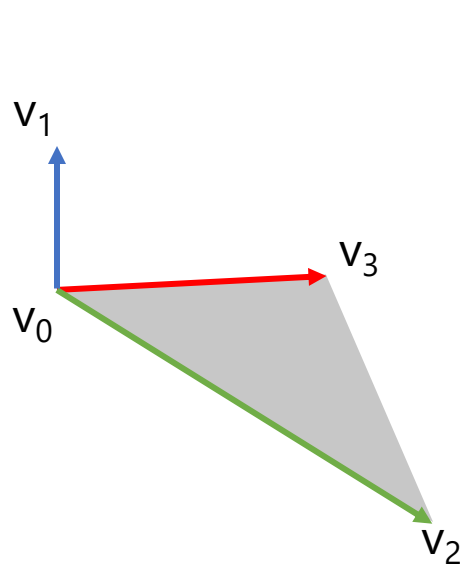
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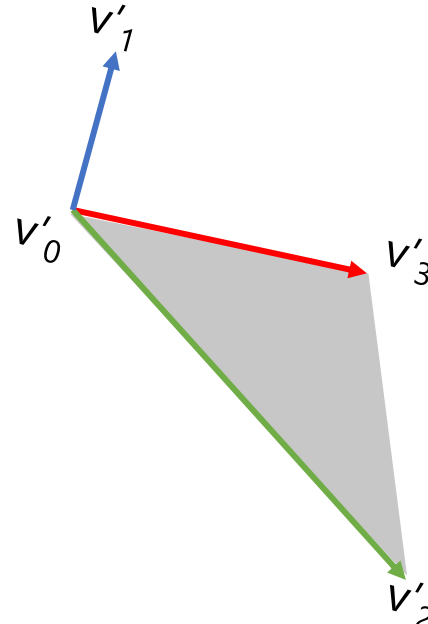
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$$E = [v_1 - v_0, v_2 - v_0, v_3 - v_0]$$

Canonical
Metric Space



$$E' = [v'_1 - v'_0, v'_2 - v'_0, v'_3 - v'_0]$$

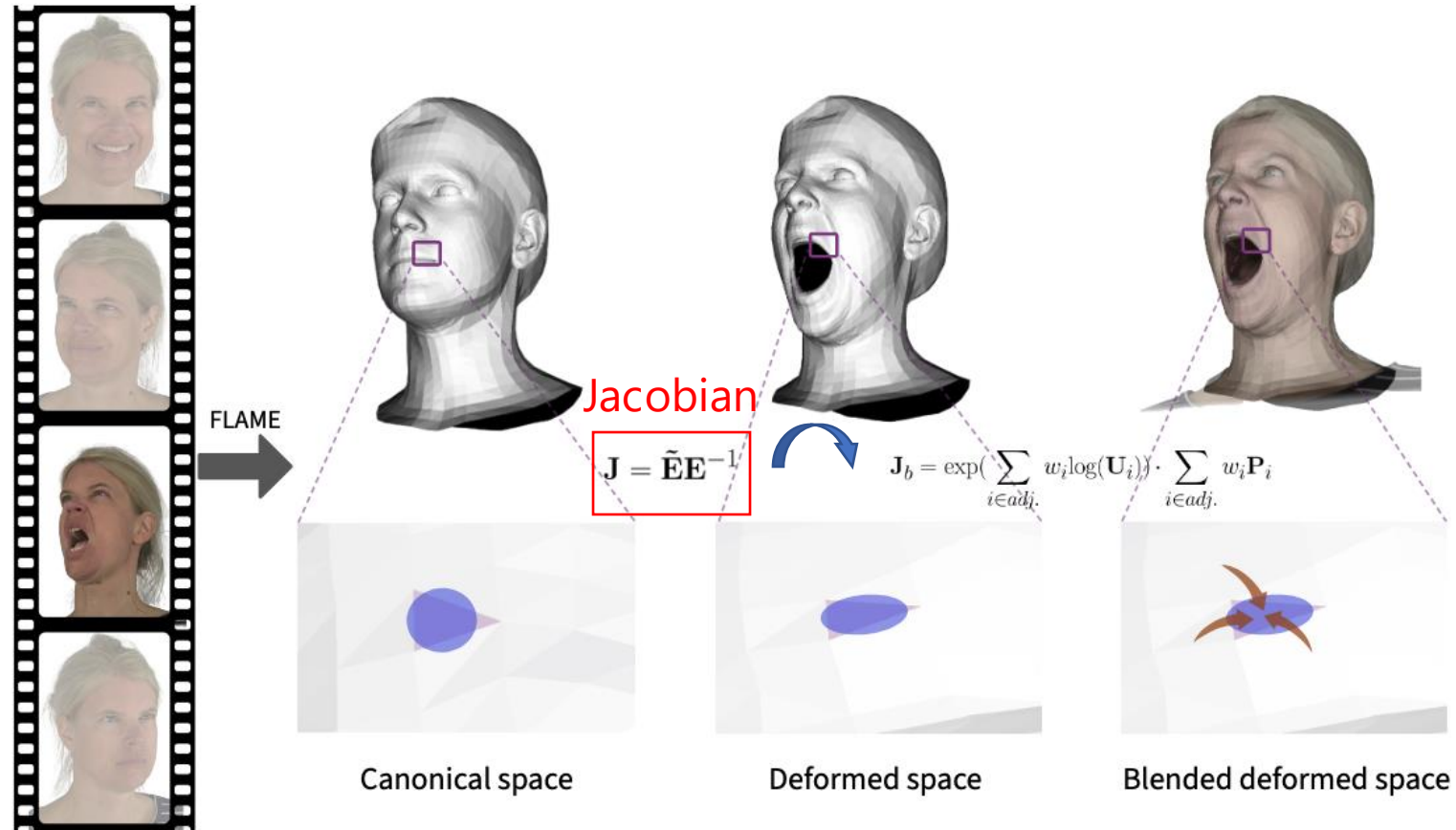
Deformed
Metric Space

$$J = E'E^{-1}$$

DTF (SIGGRAPH 2000, Sumner and Popovic.)

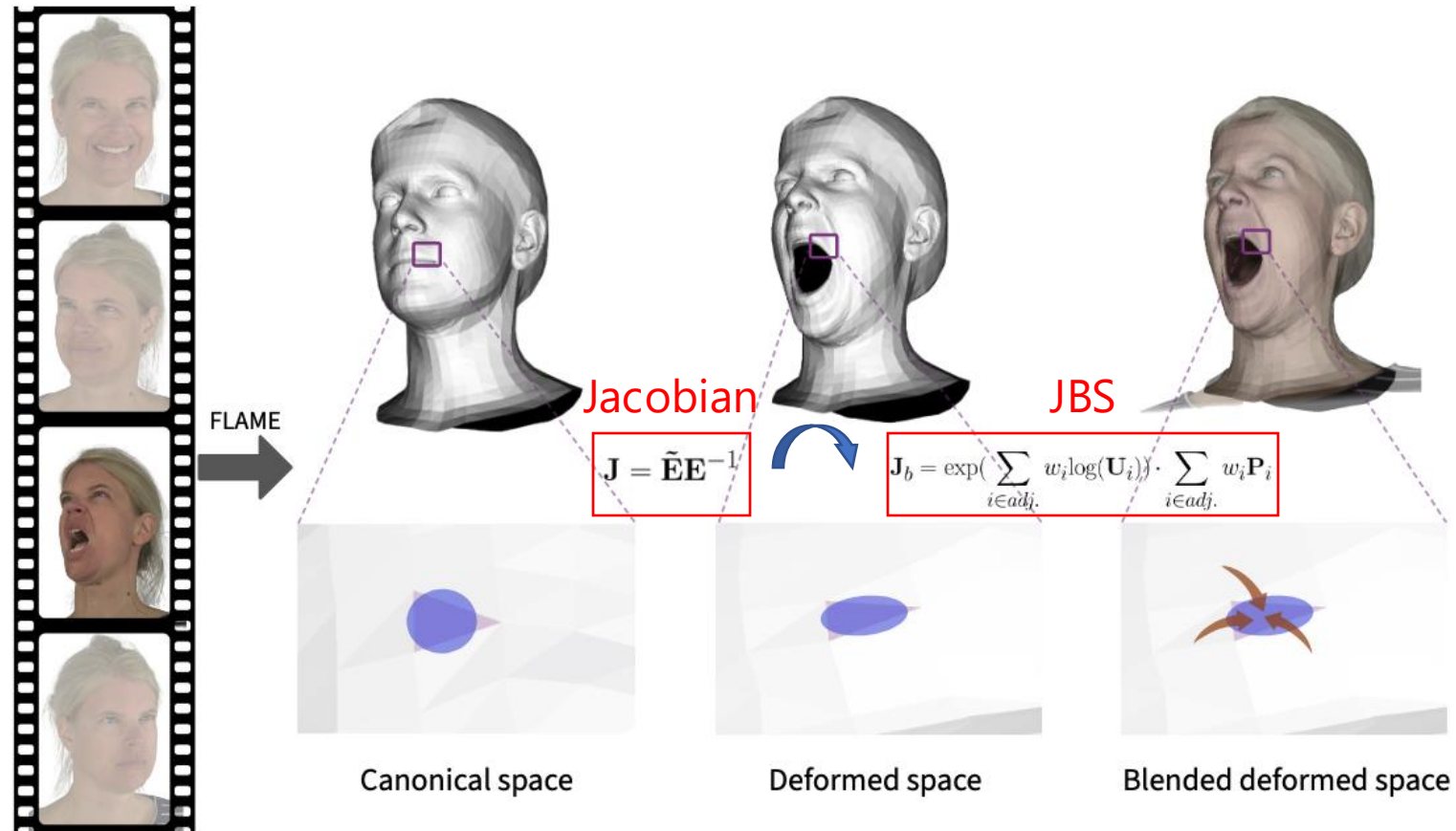
Jacobian to JBS

We replace again it with our JBS, blending Jacobians with adjacent transforms.



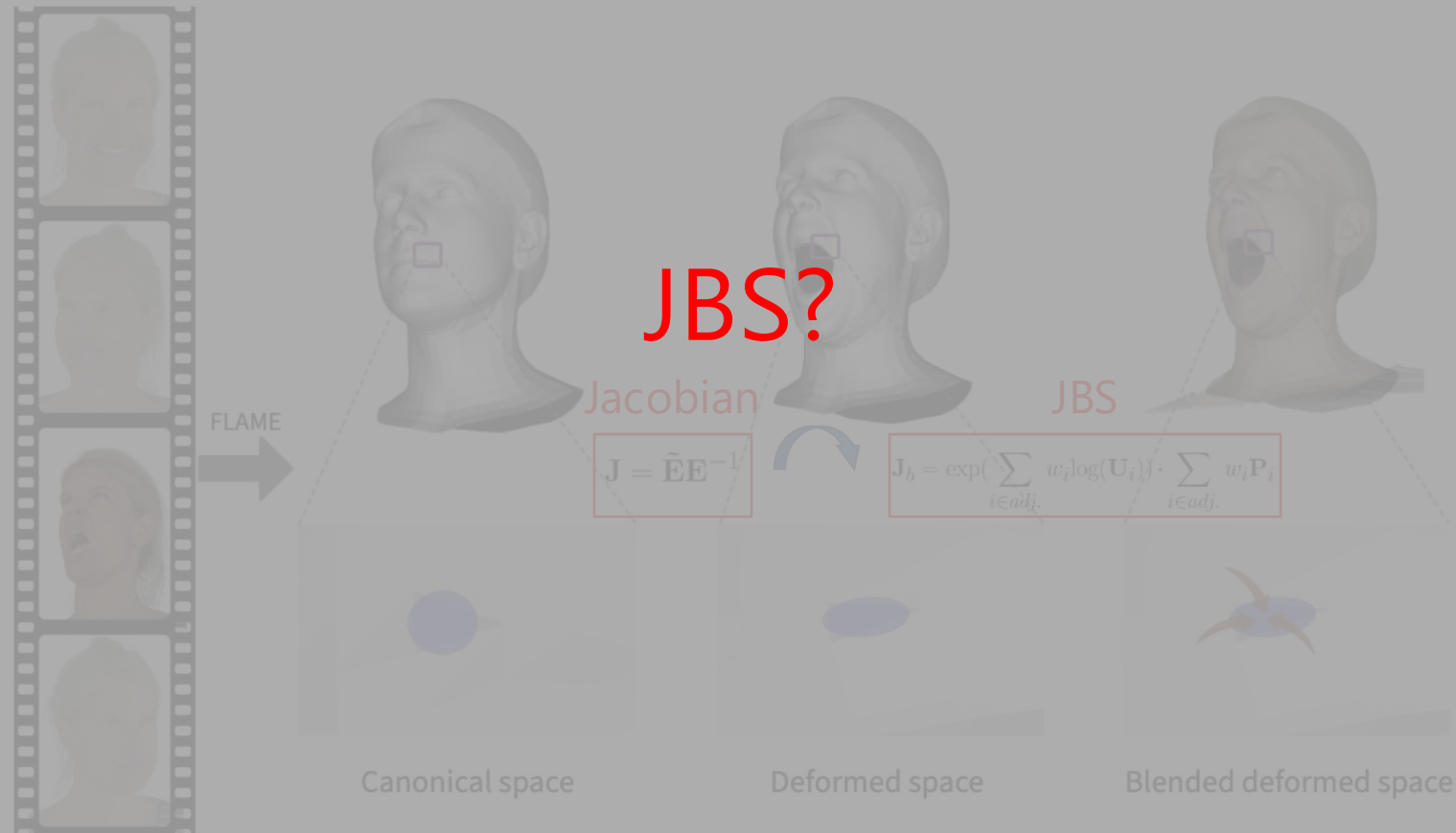
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Let's delve into JBS step-by-step

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$$\mathbf{J}_b := \text{JBS}(\mathbf{J}, w)$$

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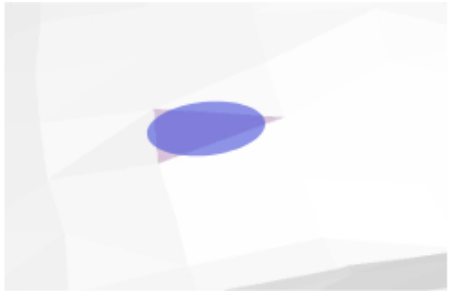
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$$\mathbf{J}_b := \text{JBS}(\mathbf{J}, w) = \exp(\underbrace{\sum_{i \in \text{adj.}} w_i \log(\mathbf{U}_i)}_{\mathbf{U}_b}) \cdot \underbrace{\sum_{i \in \text{adj.}} w_i \mathbf{P}_i}_{\mathbf{P}_b}$$

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Deformed space



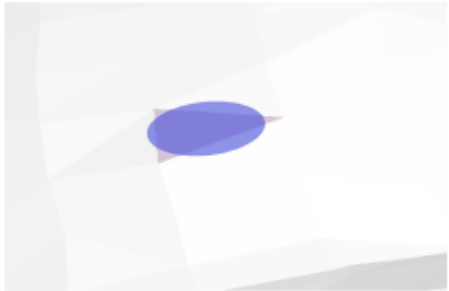
Blended deformed space

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In matrix-logarithm space



Deformed space



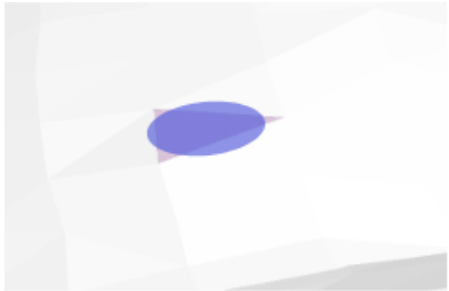
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In matrix-logarithm space In matrix-linear space



Deformed space



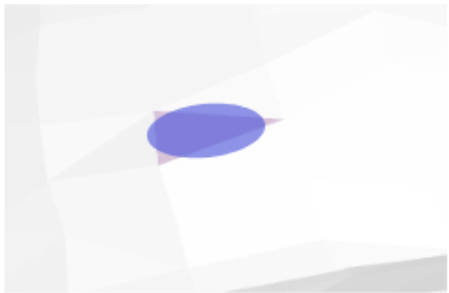
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Deformed space



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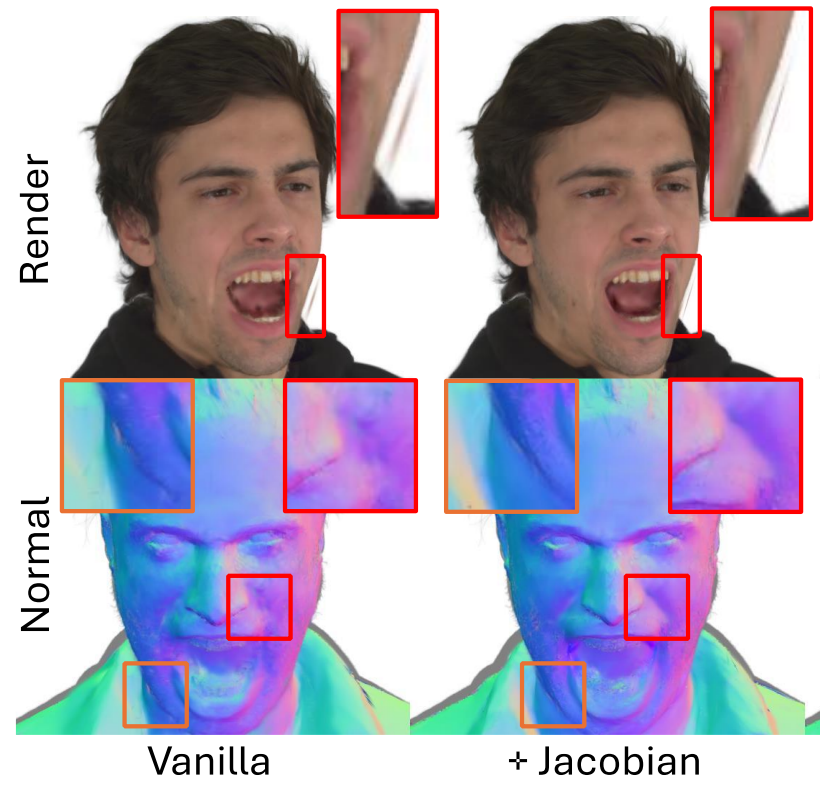
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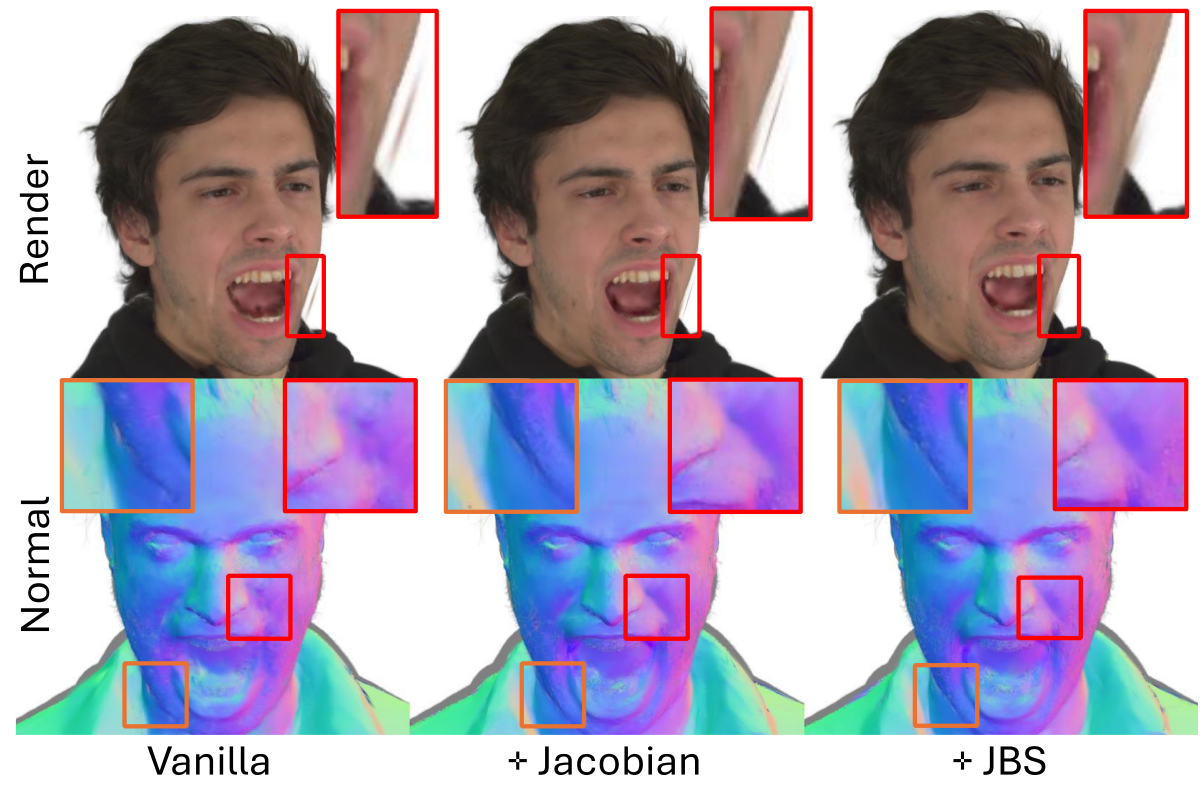


$$\Sigma^{1/2} = J_b R_c S_c \quad \mu = J_b \mu_c + T_p$$

Finally



Finally



Resolving hollow-illusion



Resolving hollow-illusion



Stop Gradient



Rotation R
Position μ

$$\mathcal{L}_{\text{eye}} = \sum_{i \in E} (1 - \alpha_i)^2$$

Resolving hollow-illusion

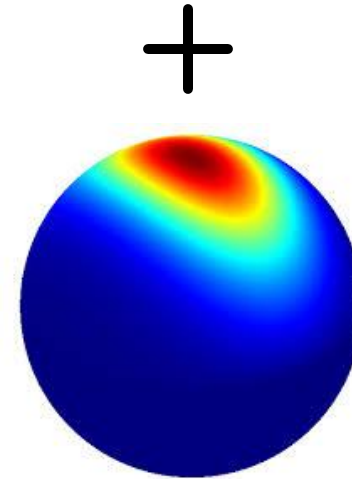


Stop Gradient



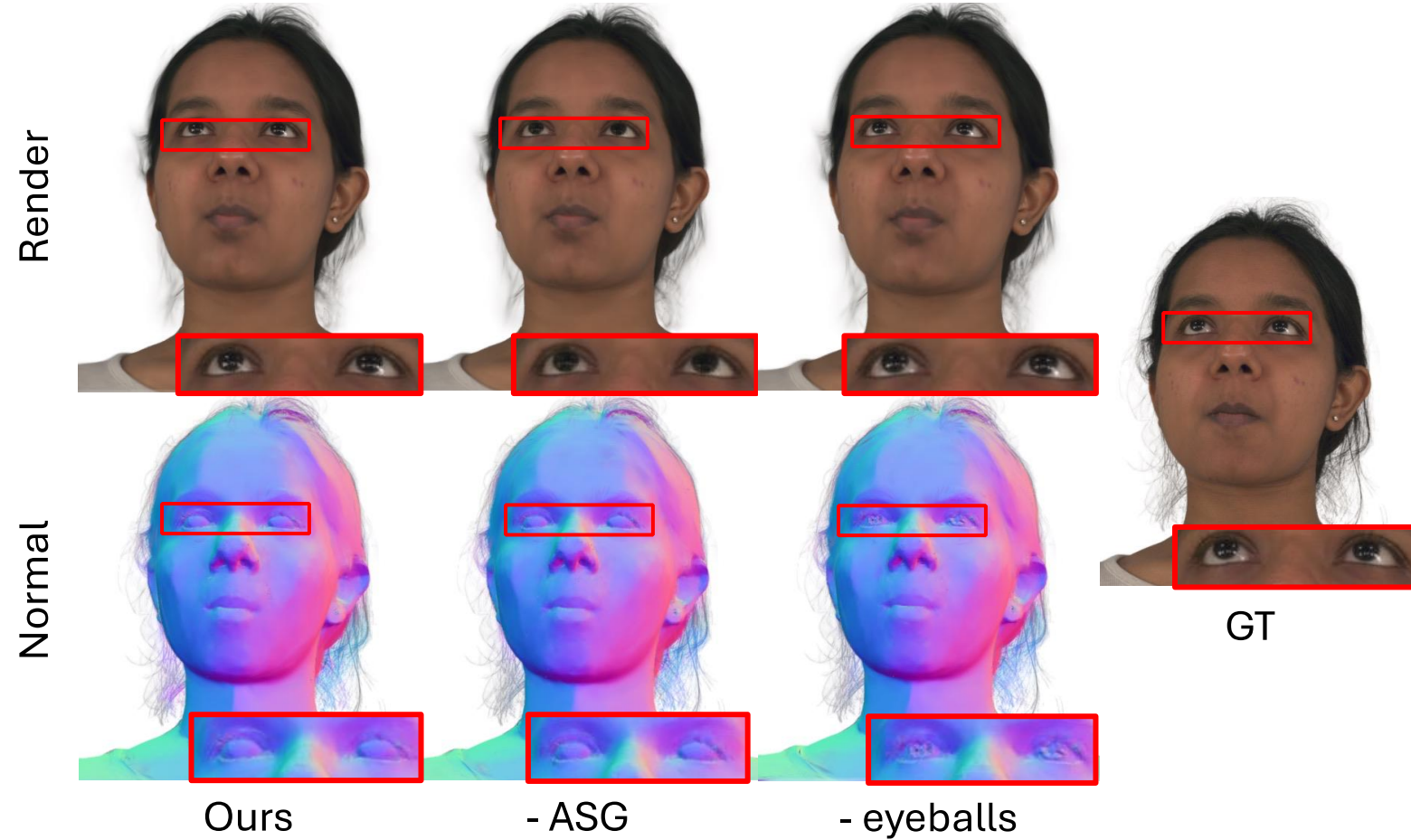
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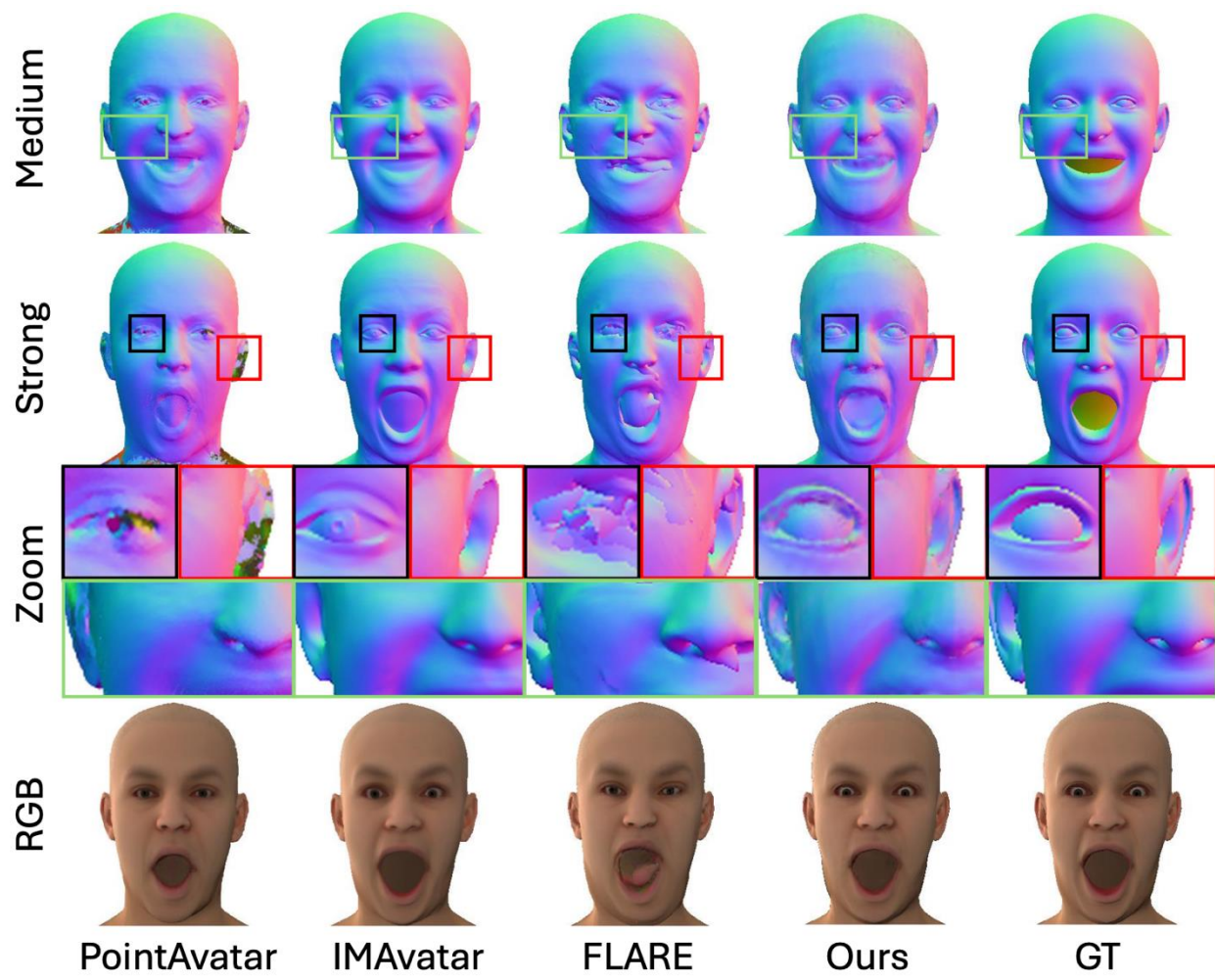


ASG (TOG 2013, Xu et al.)

Resolving hollow-illusion



Comparison on FaceTalk (CVPR 2023, Zheng et al.)



Comparison on NeRSemble (SIGGRAPH 2023, Kirschstein et al.)



GT

FLARE

PointAvatar

SA

GA

Ours



Source Actor

FLARE

PointAvatar

SA

GA

Ours

Application #1 **Relighting**

using GaussianShader (CVPR 2024, Jiang et al.)



Application #2 **Meshing** using Truncated Signed Distance Function (TSDF)



Thank you for watching!

For more details visit summertight.github.io/SurFhead

