

International Conference on Learning Representations (ICLR) 2026

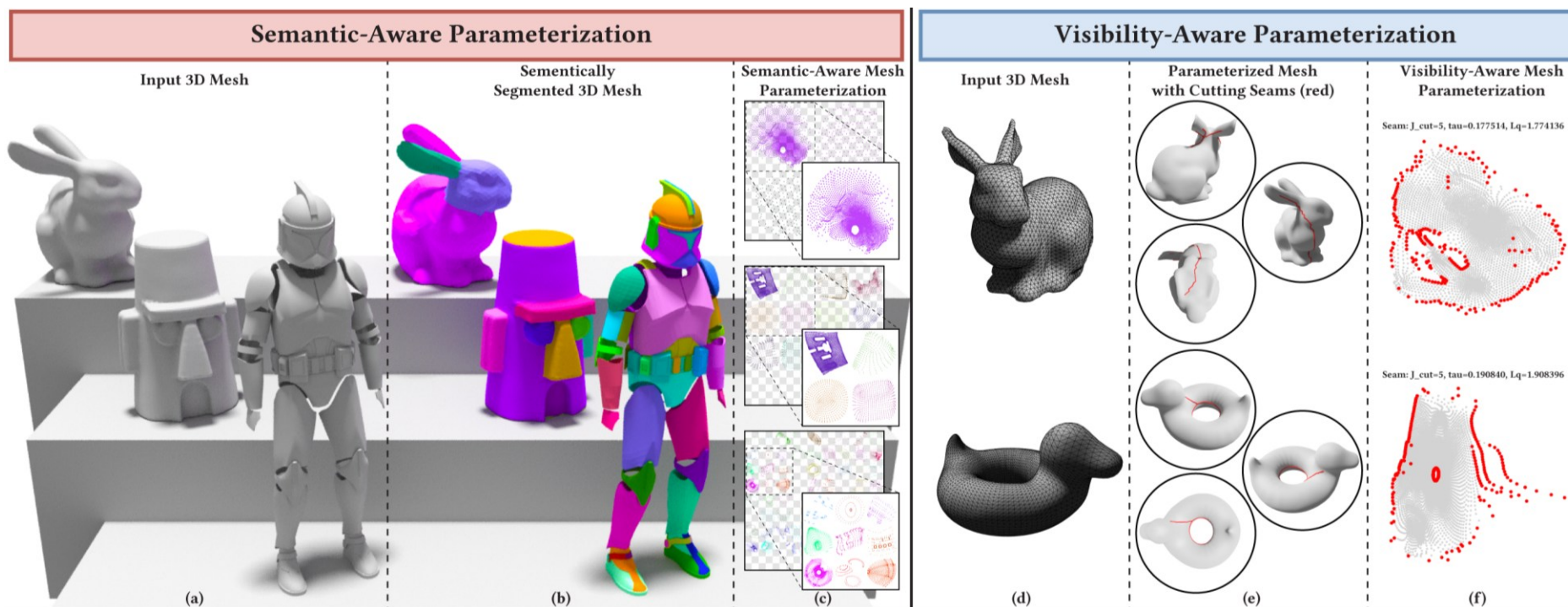
Unsupervised Representation Learning for 3D Mesh Parameterization with Semantic and Visibility Objectives

Amirhossein Zamani^{1,2,3}

Bruno Roy¹

Arianna Rampini¹

¹Autodesk Research ²Mila - Quebec AI Institute ³Concordia University



Behind the realism in games or films like Avatar is a complex process that has long challenged 3D artists!



The Last of Us Part II



Red Dead Redemption 2



Avatar

▪ Preliminary

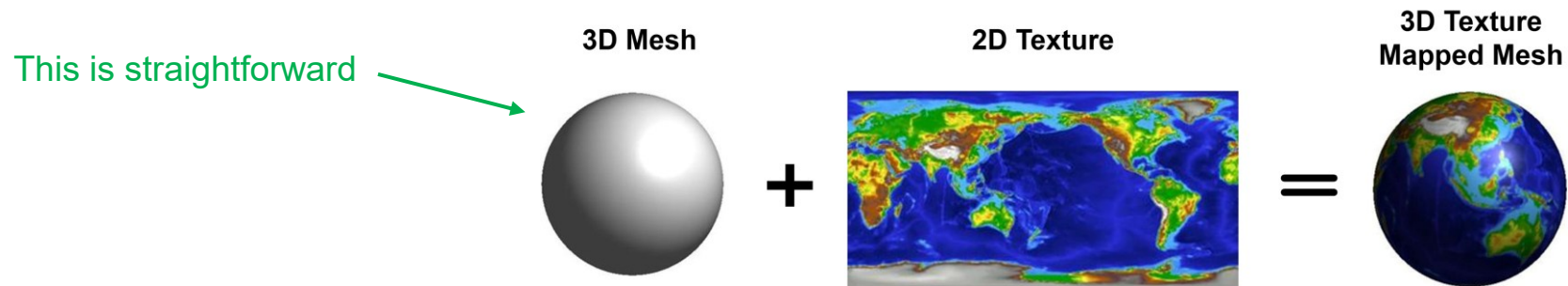
- Manual UV mapping by artist: mesh points → texture pixels

▪ Core Challenge

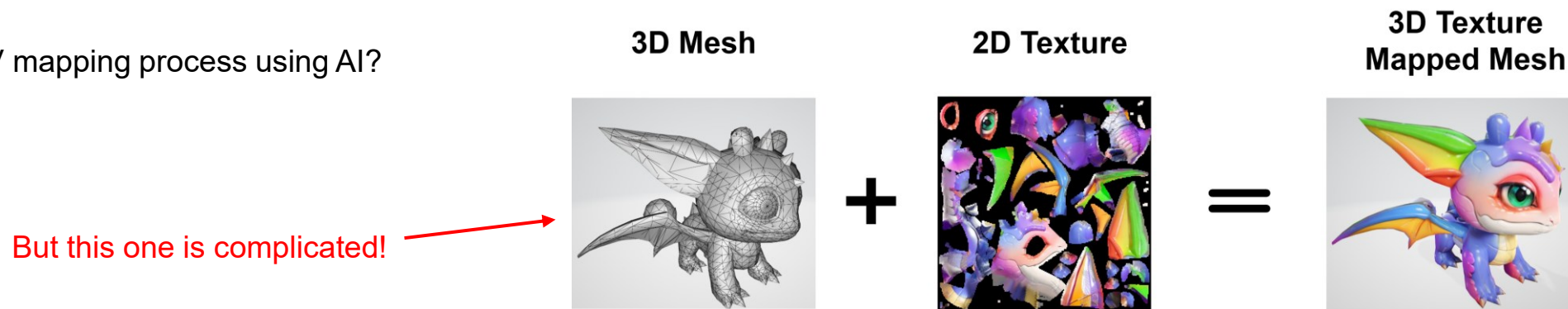
- UV mapping = ~60% of 3D content creation time

▪ Research Question

- Can we automate the UV mapping process using AI?

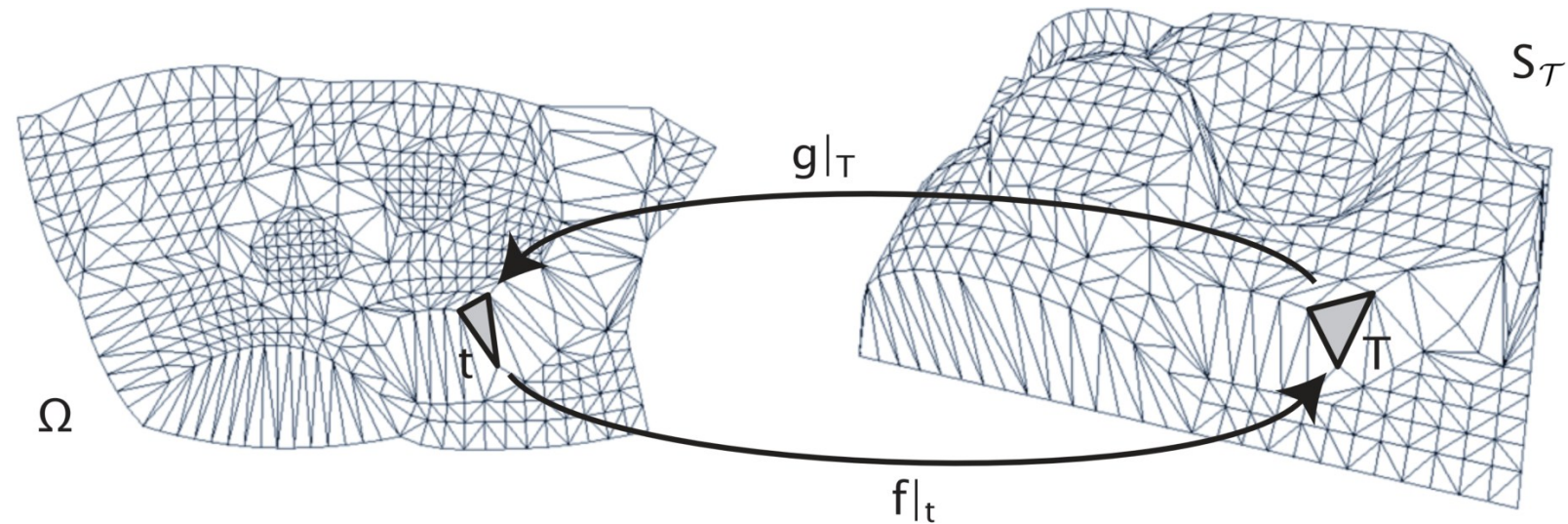


UV Mapping



- **Geometry-Preserving UV Parameterization:** Aim to flatten 3D surfaces to 2D while minimizing distortion and avoiding overlaps

- Conformality (angle-preserving)
- Equiareality (area-preserving)
- Stretch (length-preservation)
- Bijectivity



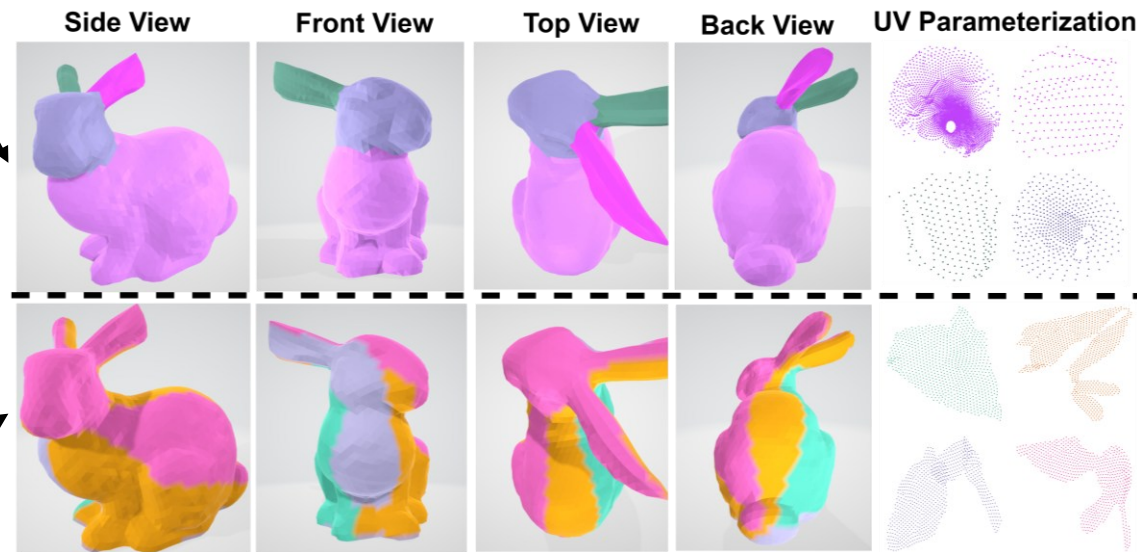
The Parameterization of a triangle mesh

(Image Credit: Kai Hormann, Bruno Lévy, Alla Sheffer [1])

- **Issue:** While these properties are necessary to achieve a high-quality parameterization, they are not sufficient for many downstream applications in content creation and texture synthesis

This is semantic-aware

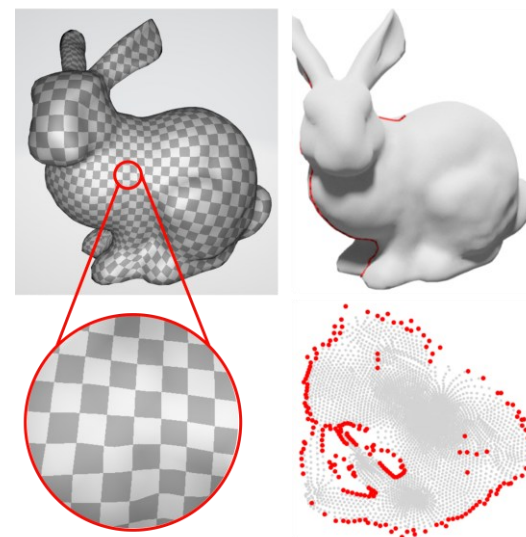
- Two important but missing properties we propose:
 - Semantic awareness:** 2D UV charts should align with semantically meaningful surface 3D parts so that textures designed for a semantic region remain coherent and transferable across 3D shapes



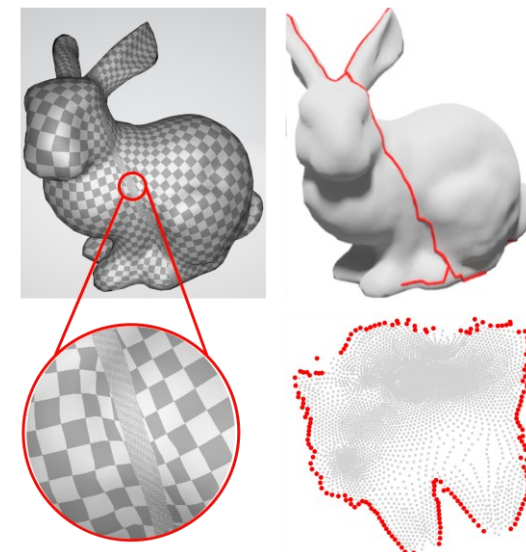
This is NOT semantic-aware

- Visibility awareness:** Seams should be placed where they are unlikely to be observed under typical viewpoints and lighting, so that seam artifacts are less perceptible after texturing and rendering

This is visibility-aware



This is NOT visibility-aware



Step 1: Semantic 3D Partitioning:

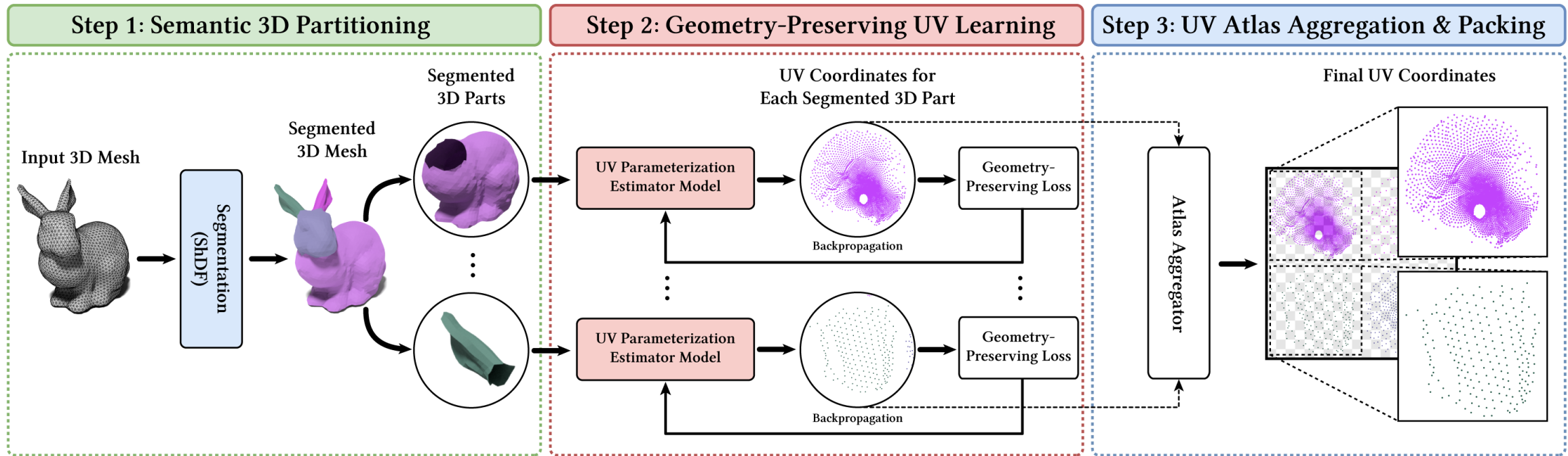
- Partition the input 3D mesh into semantic 3D parts by computing a per-vertex shape diameter function (ShDF) [2]

Step 2: Geometry-Preserving UV Learning:

- Learn a UV parameterization independently for each segmented 3D part using a multi-layer perceptron (MLP) architecture

Step 3: UV Atlas Aggregation and Packing:

- Aggregate and pack these generated UV islands into a unified UV atlas



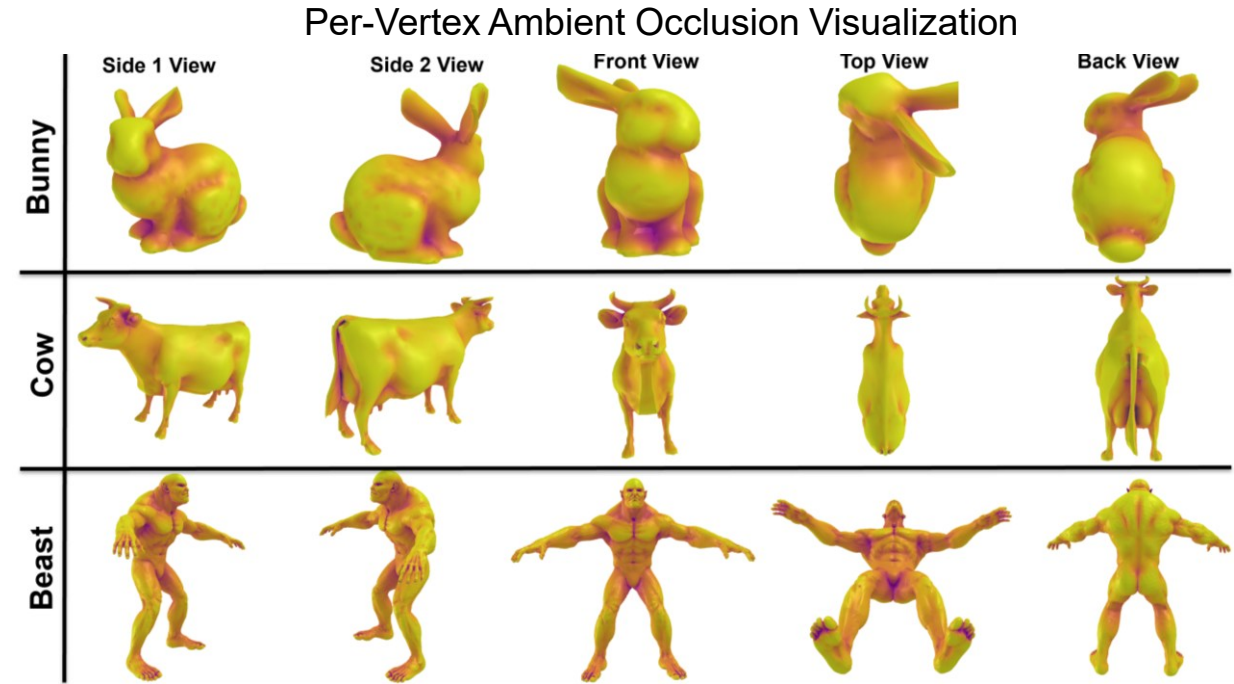
① Compute Per-Vertex Ambient Occlusion on the input mesh

$$V(p, \omega) = \begin{cases} 1, & \text{if the ray starting at } p \text{ in direction } \omega \text{ does not intersect the surface (unoccluded)} \\ 0, & \text{otherwise (occluded)} \end{cases}$$

The binary visibility function $V(p, \omega)$ measures whether direction ω is visible from a point p on the mesh surface with unit normal vector $n(p)$

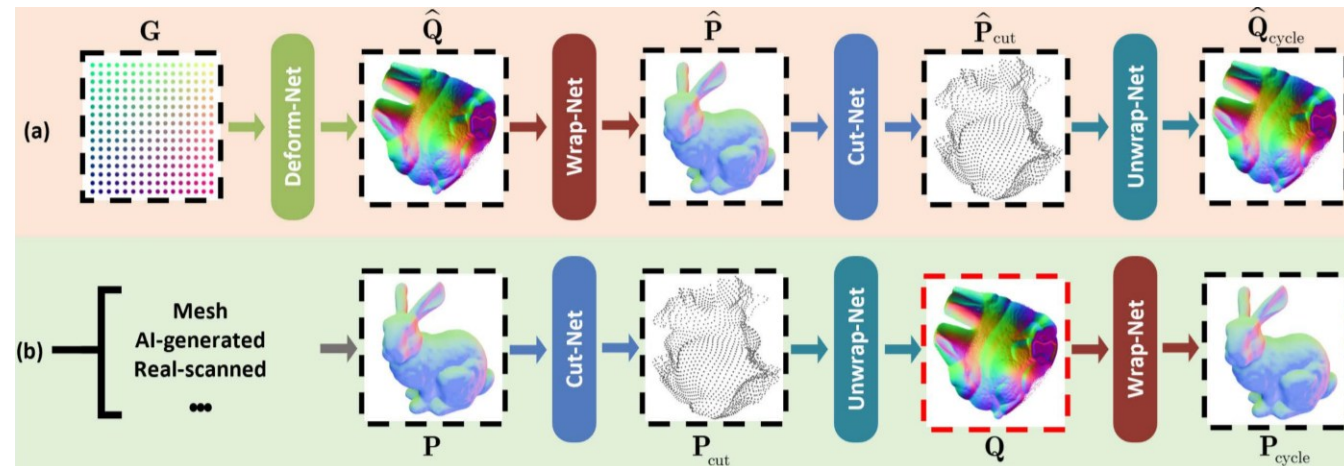
$$AO(p) = \frac{1}{\pi} \int_{\Omega^+(p)} V(p, \omega) (n(p) \cdot \omega) d\omega.$$

By this convention, $AO(p) = 1$ indicates a fully exposed point (no occlusion) and $AO(p) = 0$ indicates a fully occluded point.



② Apply a base UV-parameterization backbone to obtain candidate UV islands

- Intuitively, the **bi-directional cycle mapping** [3] enforces that projecting a 3D surface patch into 2D UV coordinates and re-projecting those UVs back onto the 3D mesh yields the original data.



[3] Zhao, Yuming, et al. "FlexPara: Flexible Neural Surface Parameterization." *IEEE Transactions on Pattern Analysis and Machine Intelligence* (2025).

③ Extract UV boundary points corresponding to 3D cutting seam points

$$\eta_i = \max_{j=1}^N (\|q_i - q_{i,j}\|_2) \longrightarrow s_i = \sigma(\beta(\eta_i - \tau))$$

Each point in UV space

N nearest neighbor UV coordinates

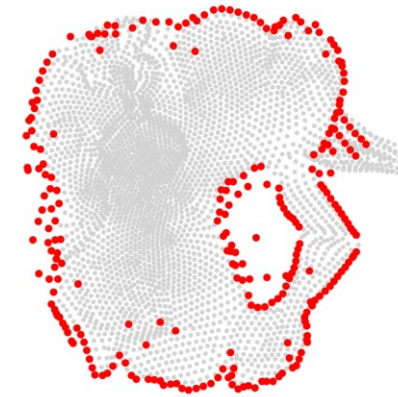
Per-vertex seam membership score

A threshold based on the UV domain size

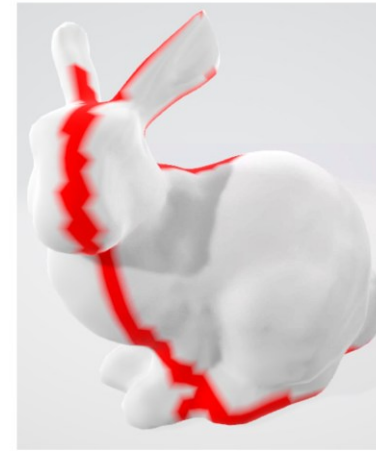
$$\tau = \tau_{\text{scale}} L(Q)$$

The side length of bounding square of the UV map

UV Space
Seam (J_cut=5, tau=0.090832, Lq=1.816645)



3D Space

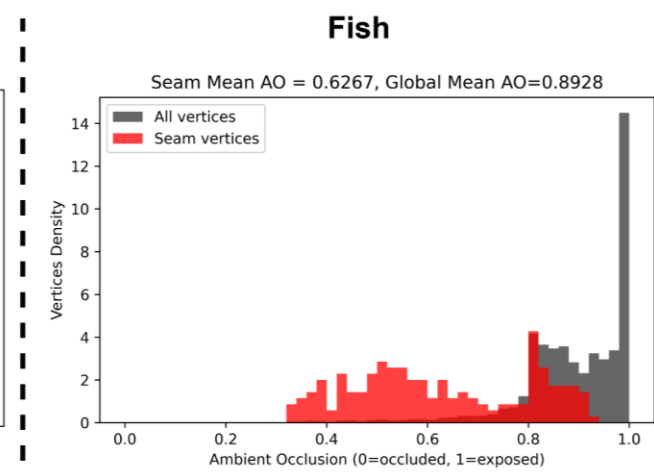
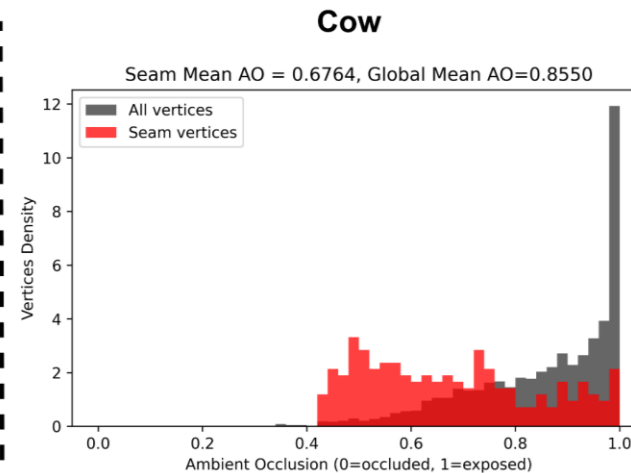
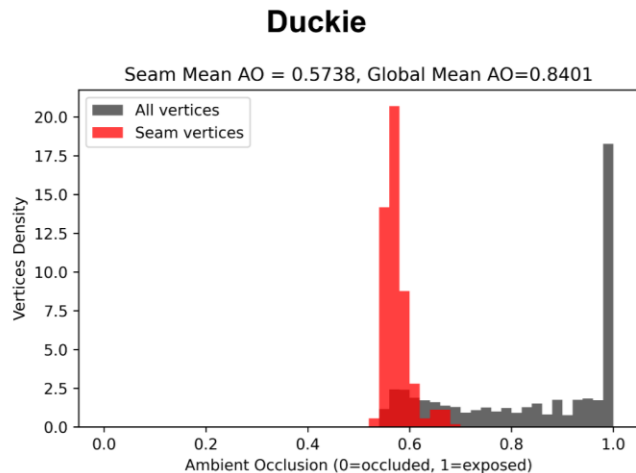


④ Minimize the AO-weighted average over the associated 3D vertices

$$\mathcal{L}_{\text{AO}} = \frac{\sum_i s_i \text{AO}_i}{\sum_i s_i + \epsilon} \longrightarrow \mathcal{L}_{\text{vis}}(\theta) = \mathcal{L}_{\text{wrap}} + \mathcal{L}_{\text{cycle}} + \mathcal{L}_{\text{repel}} + \mathcal{L}_{\text{dist}} + \lambda_{\text{vis}} \mathcal{L}_{\text{AO}}$$

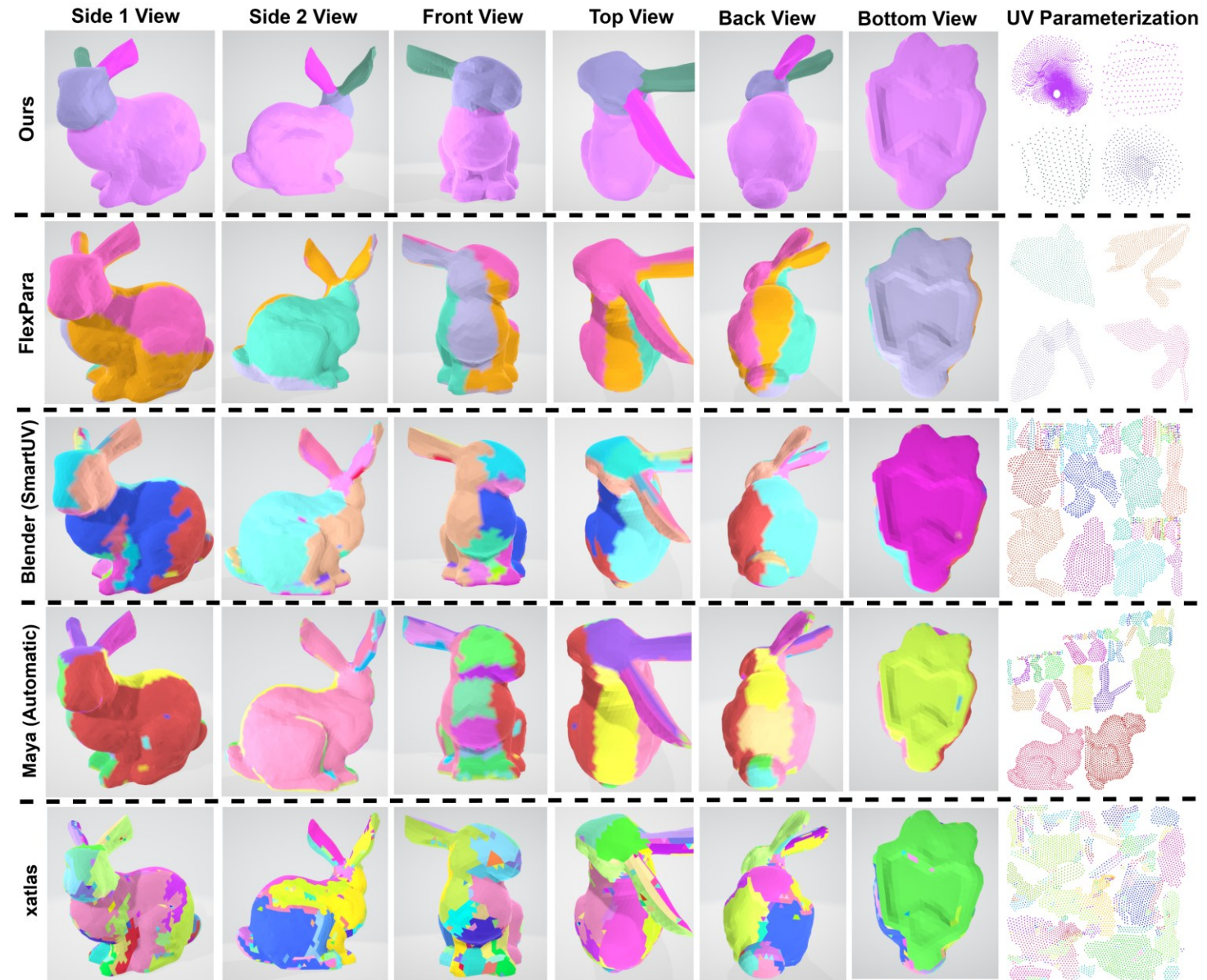
Per-vertex ambient occlusion

A small constant to avoid division by zero



Comparison of Semantic-Aware UV Parameterization Results

- Produces **UV charts that align closely with the mesh's semantic 3D parts** (e.g., clearly identifiable rabbit head, ears, body), unlike baseline methods.
- The main advantage: **Simplifies texture painting** and improves workflow efficiency for UV/layout artists.

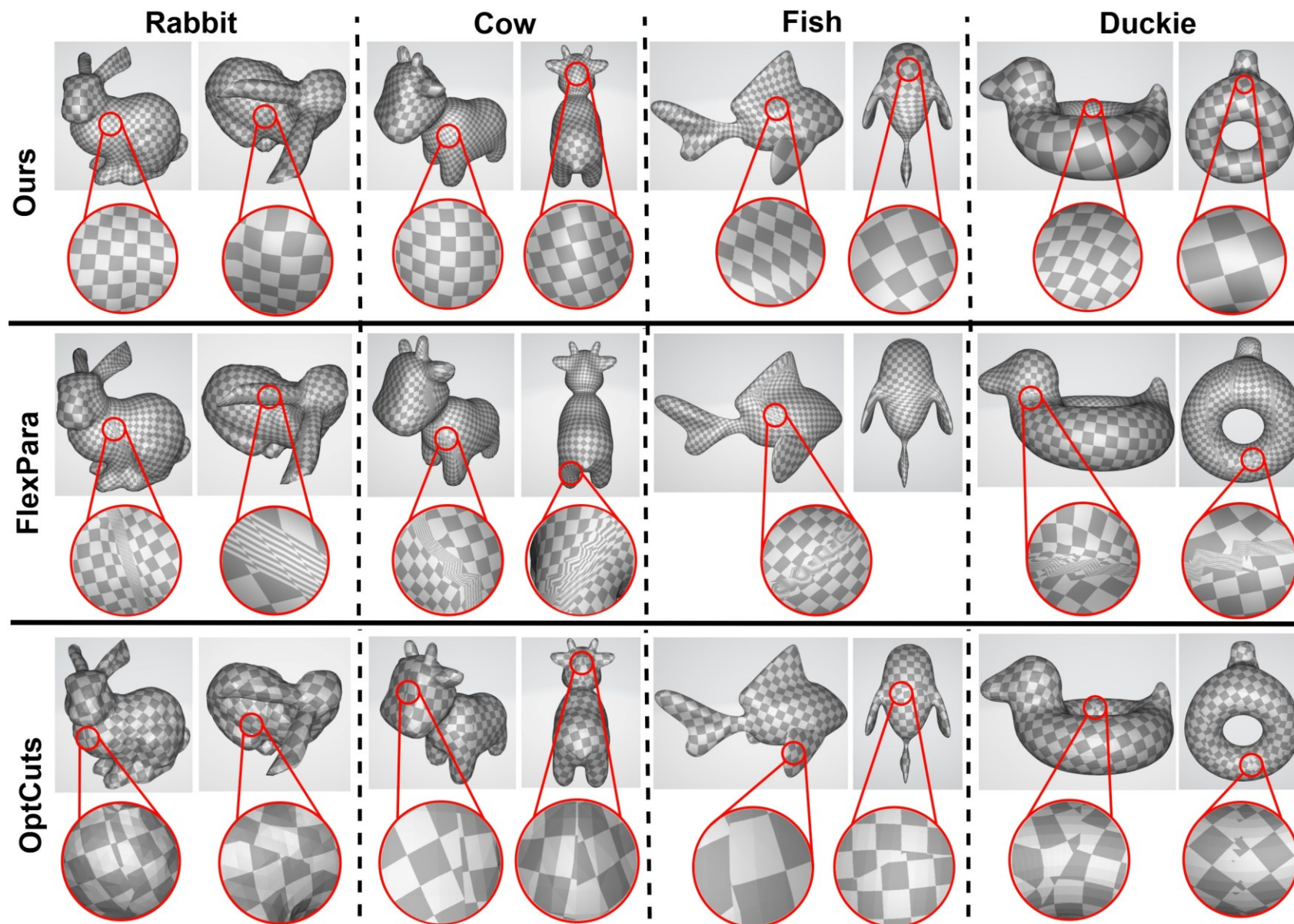


Comparison of Visibility-Aware UV Parameterization Results

Steer cutting seams toward occluded regions!

	Side View	Front View	Back View	Bottom View	Side View	Front View	Back View	Bottom View	Side View	Front View	Back View	Bottom View
Ambient Occlusion												
Ours												
FlexPara												
OptCuts												

Comparison of Checkerboard Texturing using Different UV Parameterization Methods



- Produces **improving texture continuity and consistency** from typical viewpoints.

User Preference Study

- 115 Participants = 70 general + 45 expert participants (UV/layout and modeling artists)

Table 1. User-study with 70 general participants

Method	General User Preference Percentage
Visibility-Awareness Evaluation	
OptCuts [1]	1.43 %
FlexPara (single-Chart) [2]	7.14 %
Ours - Visibility-Aware Param	91.42 %
Semantic-Awareness Evaluation	
xatlas [3]	4.28 %
Blender [4]	4.28 %
Autodesk Maya [5]	10 %
FlexPara (multi-Chart) [2]	7.14 %
Ours - Semantic-Aware Param	74.29 %

Table 2. User-study with 45 expert participants

Method	Expert User Preference Percentage
Visibility-Awareness Evaluation	
OptCuts [1]	1.48 %
FlexPara (single-Chart) [2]	4.81 %
Ours - Visibility-Aware Param	93.70 %
Semantic-Awareness Evaluation	
xatlas [3]	2.66 %
Blender [4]	5.33 %
Autodesk Maya [5]	8.44 %
FlexPara (multi-Chart) [2]	3.33 %
Ours - Semantic-Aware Param	80.22 %

Takeaway: The trade-off between semantic-/visibility-awareness and geometry-preserving properties can be controlled in our pipeline.

More importantly, the semantic and visibility properties are more often valuable in downstream tasks such as mass production, semantic editing, and 3D asset reuse, where consistent UV semantic and low seam visibility significantly reduce manual effort.

Quantitative Comparison

Table 3. Quantitative comparison of visibility-aware UV parameterization

Method	Visibility ↓	Conformality ↑	Equiareality ↑	Time (s) ↓
OptCuts [1]	0.7855	0.9341	0.8934	240
FlexPara (Single-Chart) [2]	0.8604	0.9097	0.6759	2
Ours - Visibility-Aware Param	0.6065	0.9175	0.6093	2
Ours - Semantic+Visibility-Aware Param	0.6534	0.9153	0.6369	15

Table 4. Quantitative comparison of semantic-aware UV parameterization

Method	Semantic Awareness		Conformality ↑	Equiareality ↑	Inference Time (sec) ↓
	Hamming Distance ↓	Rand Index ↑			
xatlas [3]	0.8896	0.7023	0.9792	0.9341	12
Blender [4]	0.8634	0.7173	0.9289	0.9023	< 1
Maya [5]	0.8615	0.7125	0.9272	0.8746	< 1
FlexPara (Multi-Chart) [2]	0.5980	0.6902	0.9592	0.7606	2
Ours - Semantic-Aware Param	0.3188	0.8151	0.9123	0.6707	15
Ours - Semantic+Visibility-Aware Param	0.3212	0.8087	0.9153	0.6369	15

[1] Li, Minchen, et al. "Optcuts: Joint optimization of surface cuts and parameterization." *ACM transactions on graphics (TOG)* 37.6 (2018): 1-13.

[2] Zhao, Yuming, et al. "FlexPara: Flexible Neural Surface Parameterization." *IEEE Transactions on Pattern Analysis and Machine Intelligence* (2025).

[3] jpcy. xatlas: Mesh parameterization / uv unwrapping library. <https://github.com/jpcy/xatlas>, 2025.

[4] Blender Foundation. *Blender - a 3D creation suite (Version 4.5)*. Blender Foundation, 2025. URL: <https://www.blender.org>.

[5] Autodesk, Inc. *Autodesk Maya (Version 2025.x)*. Autodesk, Inc., 2025. URL: <https://www.autodesk.com/products/maya/overview>.

Thank you for your attention!