



Flatten Anything: Unsupervised Neural Surface Parameterization

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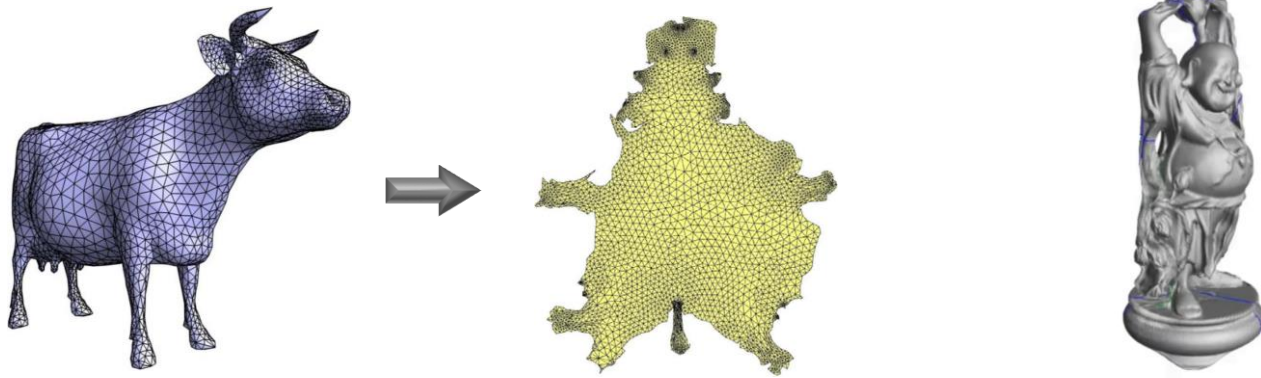
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Background

➤ Surface Parameterization

- 3D surface → *opened & unfolded & flattened* → 2D plane

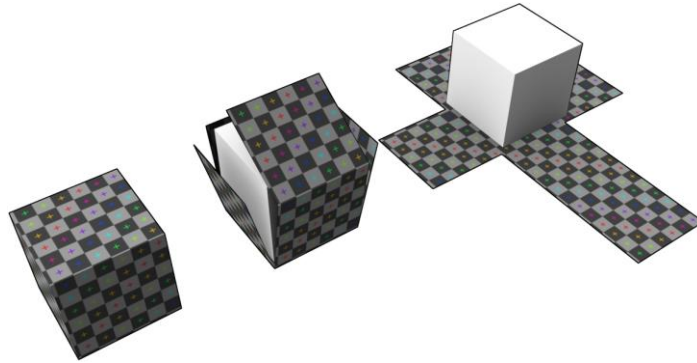


- Each 3D vertex (x, y, z) is mapped to a 2D UV coordinate (u, v) .
 - Satisfy certain *continuity* and *distortion* constraints.
-

Background

➤ Objectives

- Mimic the actual physical process of flattening a 3D surface onto a 2D plane.

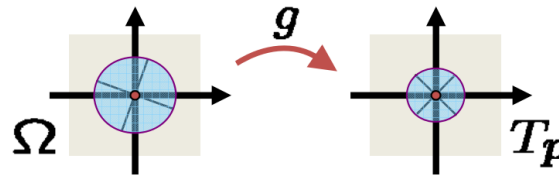


- High requirements and explicit constraints:

- Global Mapping (Instead of Local)**

- Free Boundary (Instead of Fixed)**

- Conformal (Angle-Preserving)**

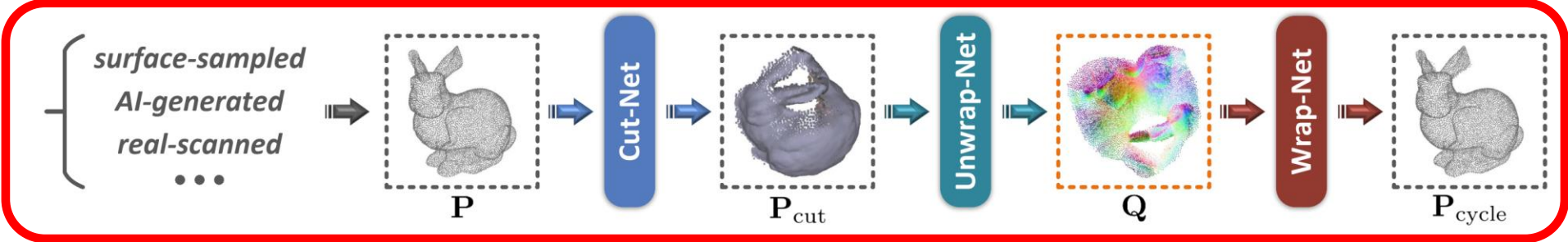
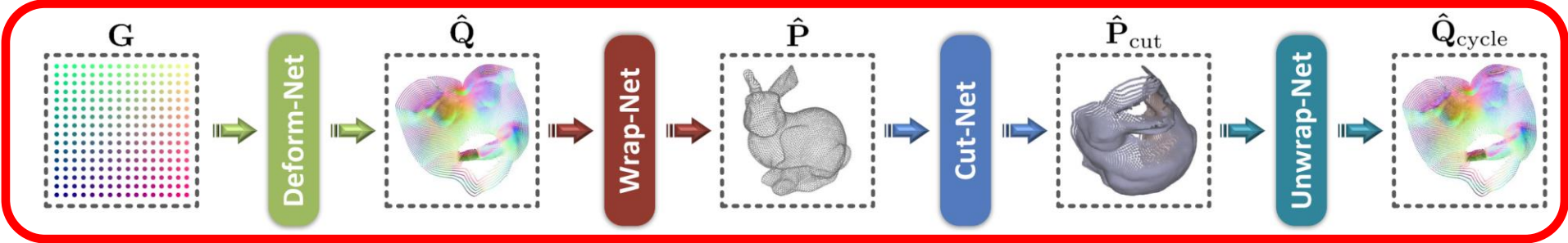


Flatten Anything Model (FAM)

- **Geometrically-Interpretable Network Components**
 - **Deforming Network (Deform-Net)**
 - ❑ Deform initial 2D grids to potentially-optimal UV coordinates.
 - **Wrapping Network (Wrap-Net)**
 - ❑ Fold from 2D to 3D.
 - **Surface Cutting Network (Cut-Net)**
 - ❑ Find cutting seams → highly-developable 3D surface manifold
 - **Unwrapping Network (Unwrap-Net)**
 - ❑ Unfold from 3D to 2D.
-

Flatten Anything Model (FAM)

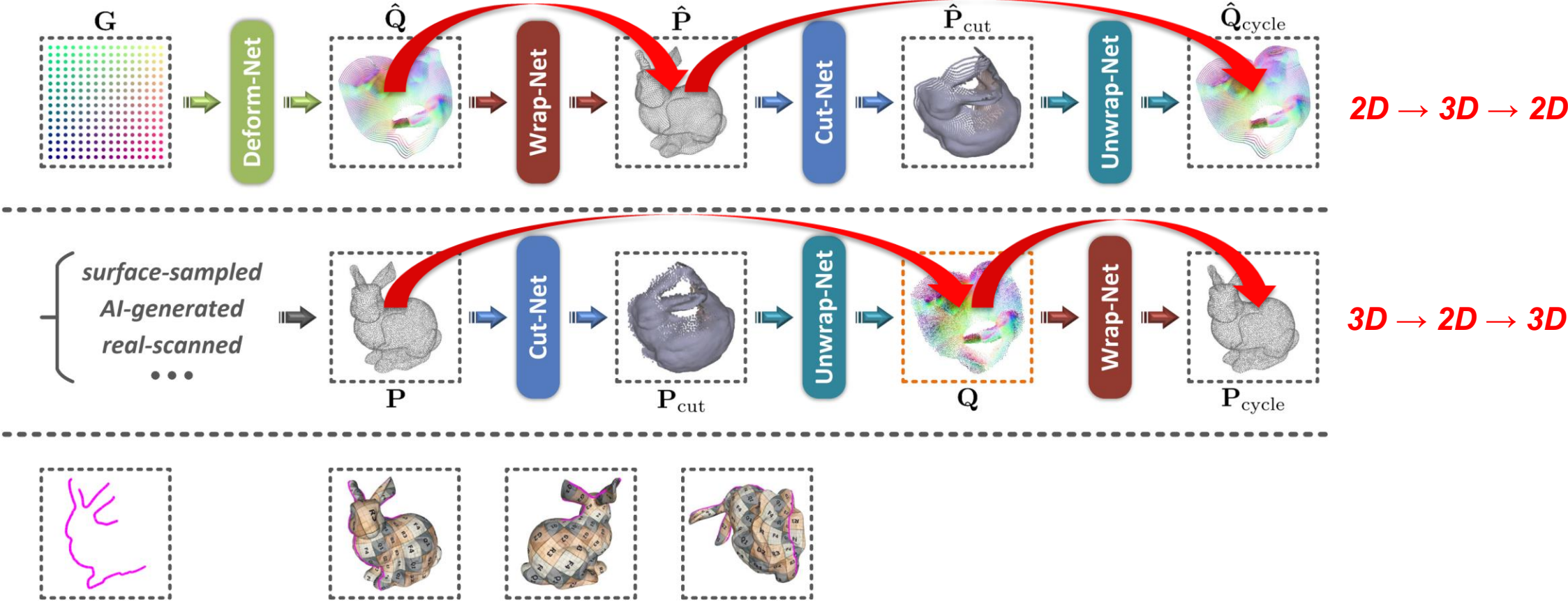
➤ Bi-Directional Cycle Mapping



Jointly Optimize Two Branches
Opposite Mapping Direction

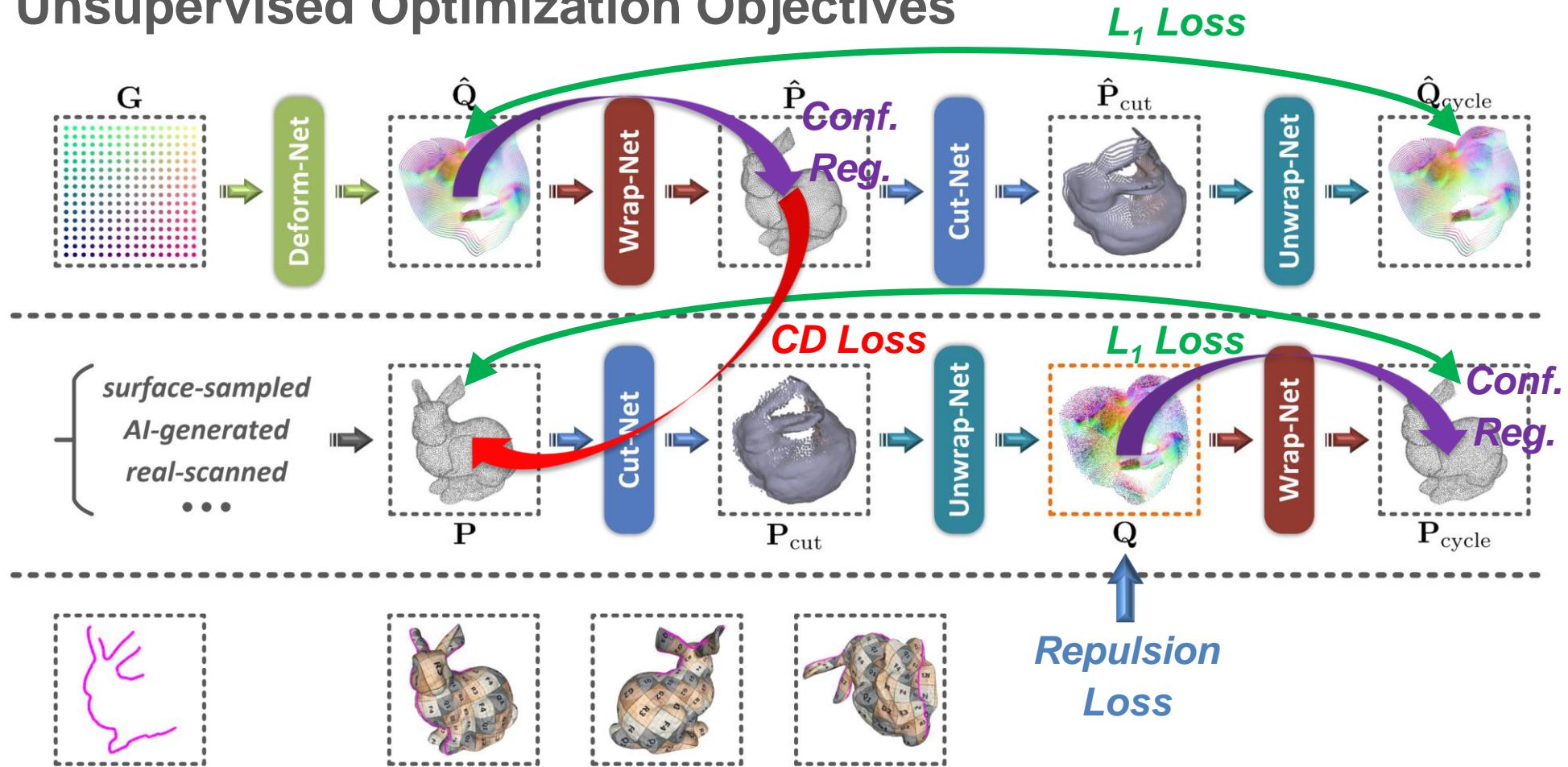
Flatten Anything Model (FAM)

➤ Bi-Directional **Cycle** Mapping



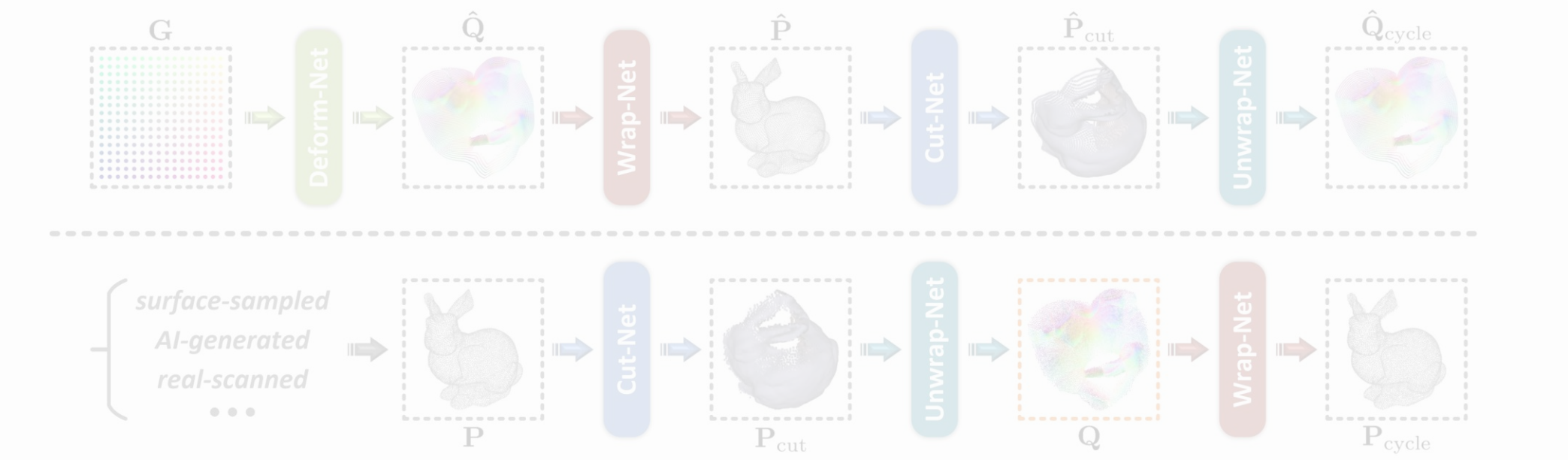
Flatten Anything Model (FAM)

➤ Unsupervised Optimization Objectives



Flatten Anything Model (FAM)

➤ Bi-Directional Cycle Mapping



Extraction of Cutting Seams

$$\eta_i = \max(\{\|\mathbf{q}_i - \mathbf{q}_i^{(k)}\|_2\}_{k=1}^{K_{\text{cut}}})$$

Experiments

➤ Mesh Parameterization: Comparison with SLIM^[R1]

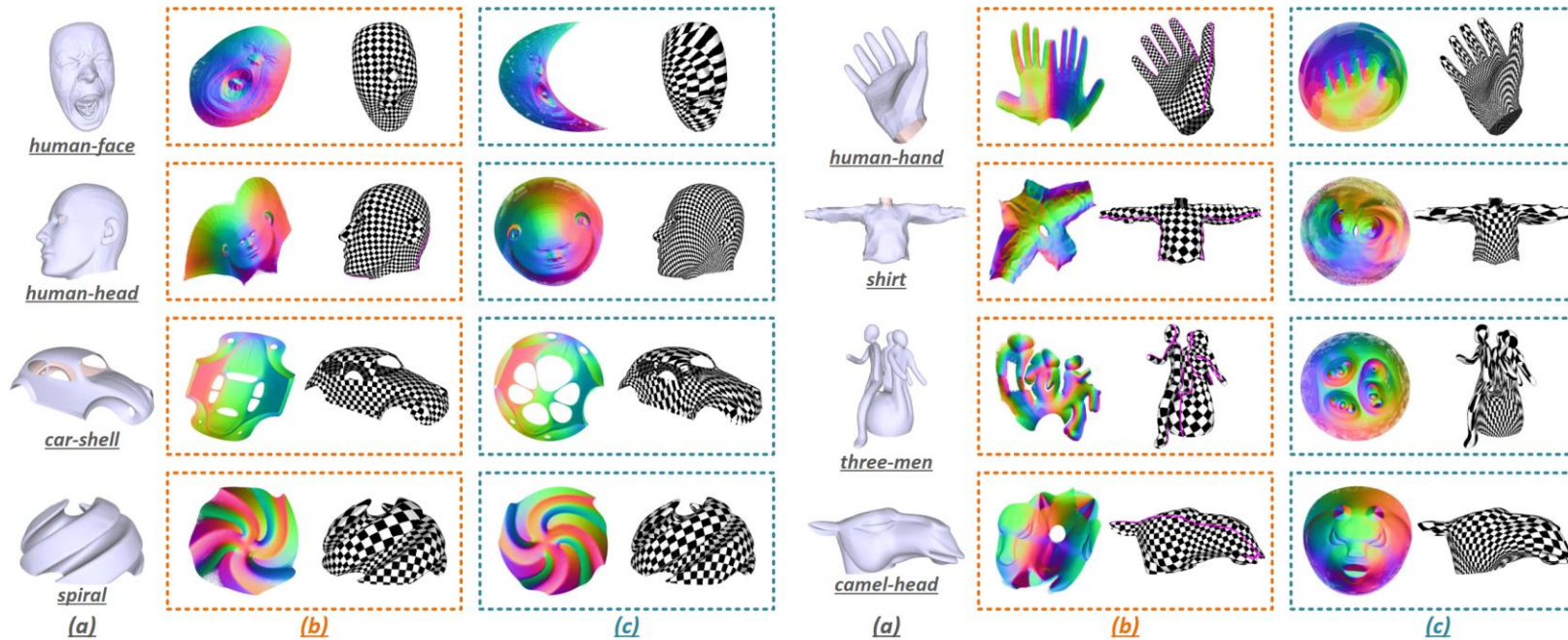


Table 1: Quantitative comparisons of our FAM and SLIM in terms of parameterization conformality.

Model	<i>human-face</i>	<i>human-head</i>	<i>car-shell</i>	<i>spiral</i>	<i>human-hand</i>	<i>shirt</i>	<i>three-men</i>	<i>camel-head</i>
SLIM	0.635	0.254	0.411	0.114	0.609	0.443	0.645	0.349
FAM	0.074	0.094	0.037	0.087	0.145	0.166	0.162	0.088

Testing Models	FAM	SLIM
Open-Surface Models (as in Figure 3)	0.156%	0.133%
Higher-Genus Models (as in Figure 4)	0.204%	N/A

Table 4: Quantitative self-intersection metrics of our parameterization results.

Experiments

➤ Point Cloud Parameterization: Comparison with FBCP-PC^[R2]

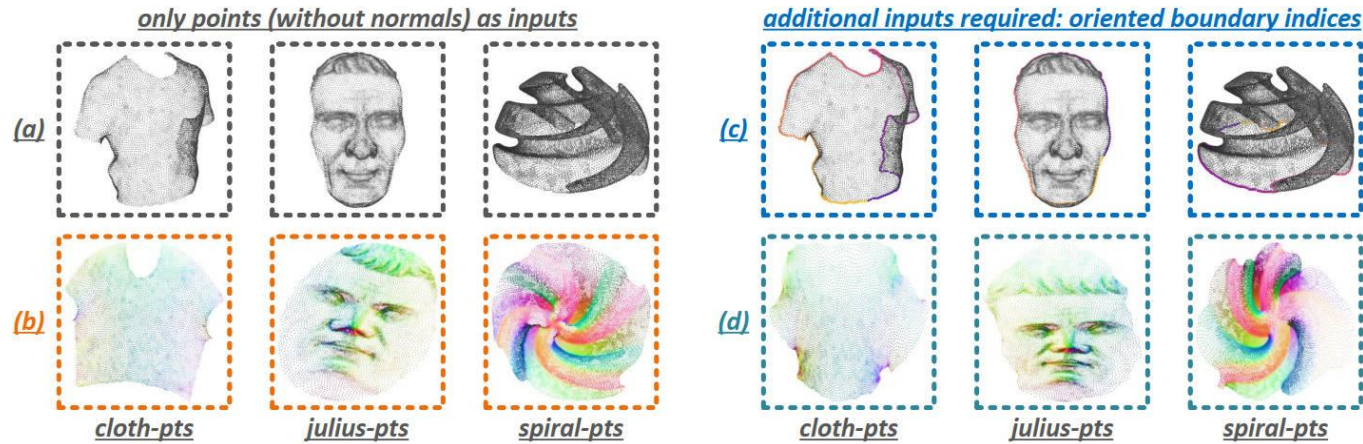


Figure 5: Point cloud parameterization achieved by our FAM (left) and FBCP-PC (right).

Table 3: Conformality metrics of our FAM and FBCP-PC for point cloud parameterization.

Model	<i>cloth-pts</i> (#Pts=7K)	<i>julius-pts</i> (#Pts=11K)	<i>spiral-pts</i> (#Pts=28K)
FBCP-PC	0.021	0.019	0.023
FAM	0.037	0.058	0.117

Experiments

➤ Robustness to Noise

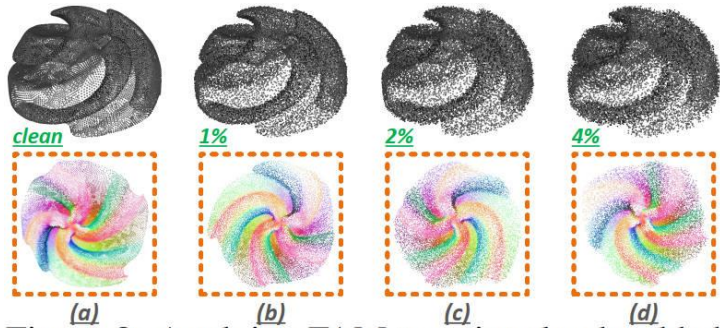
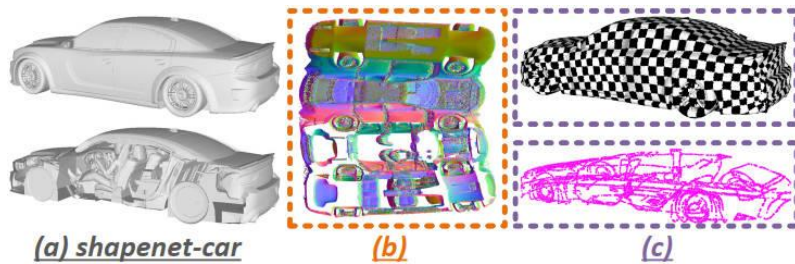
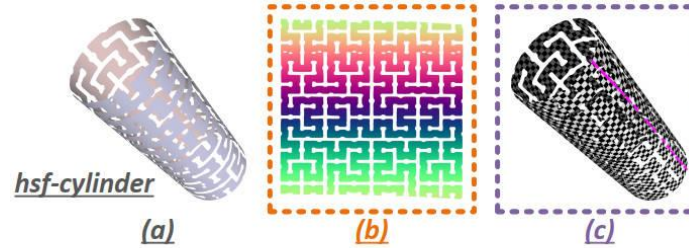


Figure 8: Applying FAM to point clouds added with different levels of Gaussian *noises*.

➤ Failure Case



➤ Stress Test



(a) Highly challenging stress tests of our FAM.

➤ Running Efficiency

Table 5: Optimization time costs (minutes) of our FAM and SLIM.

Model	<i>human-face</i>	<i>human-head</i>	<i>car-shell</i>	<i>spiral</i>	<i>human-hand</i>	<i>shirt</i>	<i>three-men</i>	<i>camel-head</i>
SLIM	39 min	38 min	19 min	25 min	28 min	31 min	17 min	40 min
FAM	around 18 min (basically unchanged for different models)							



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