

Improving RGB-D Point Cloud Registration by Learning Multi-scale Local Linear Transformation

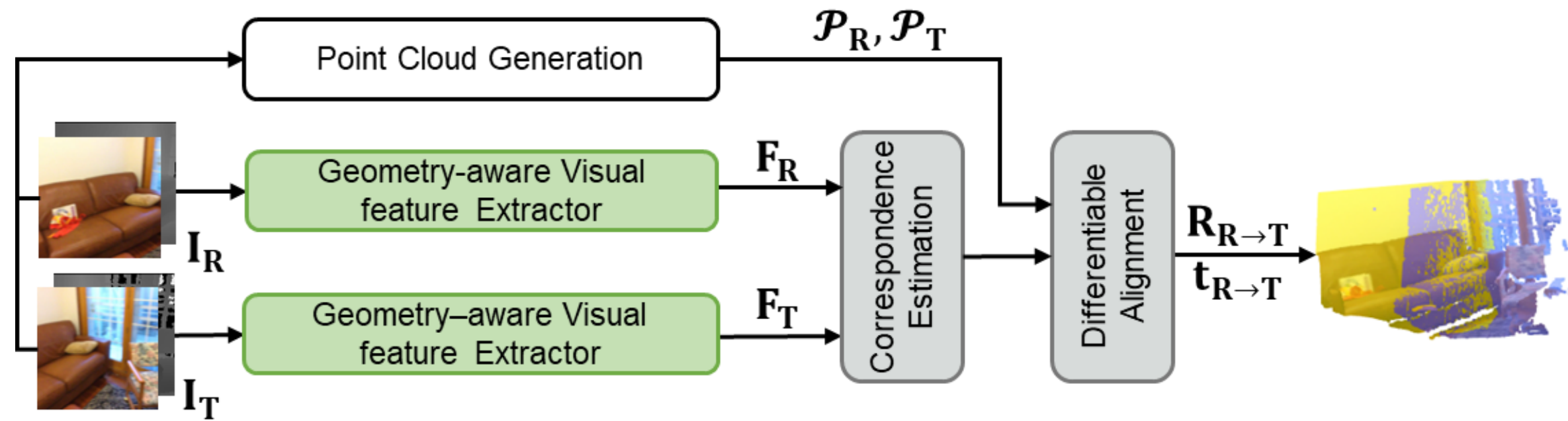
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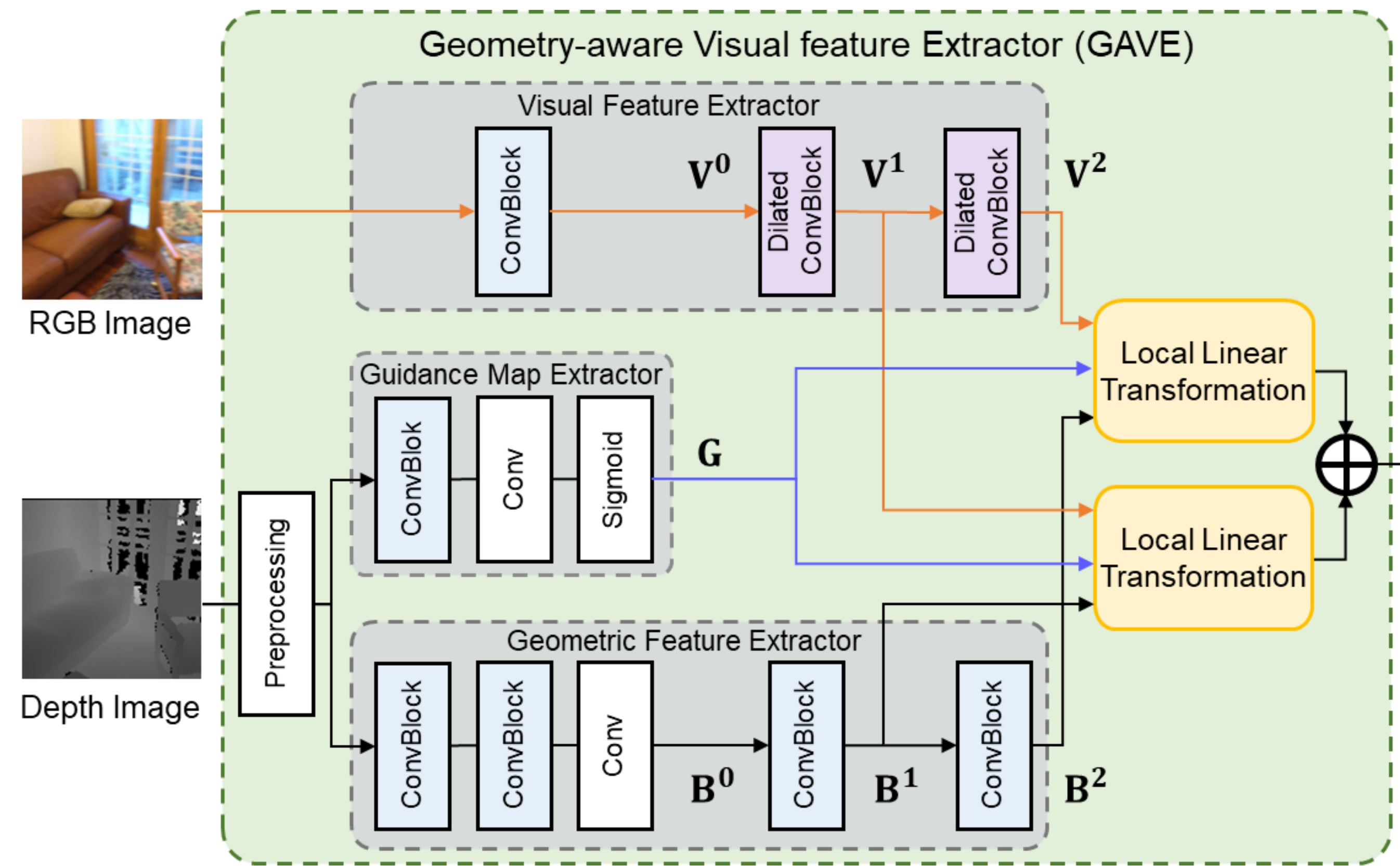


Introduction

- We introduce a **Geometry-Aware Visual Feature Extractor (GAVE)** to generate distinctive but comprehensive geometric-visual features for point cloud registration.
- We propose a **Local Linear Transformation (LLT)** module, which employs the Bilateral Grid and an edge-aware guidance map to generate our content-aware linear coefficient.
- By collaborate multi-scale LLT modules in GAVE, we observe a significant performance enhancement.

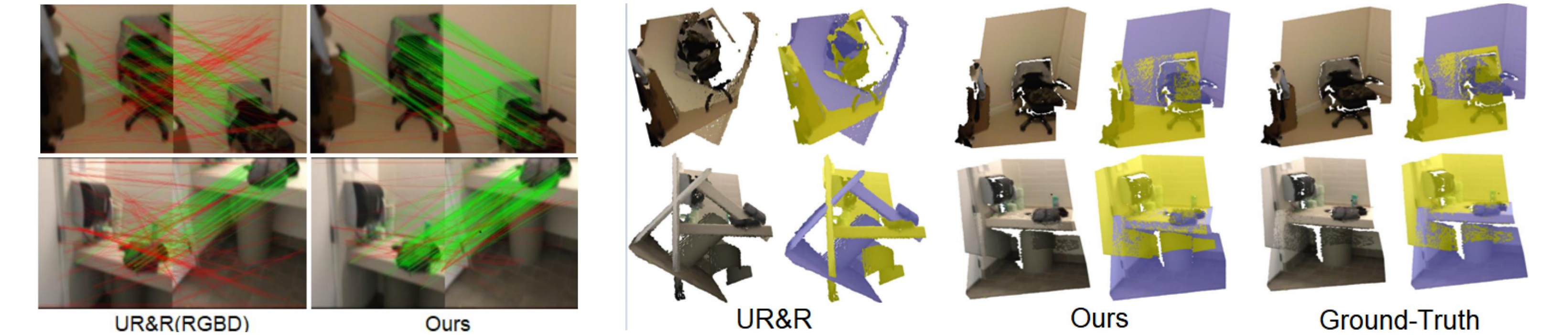


Network Architecture



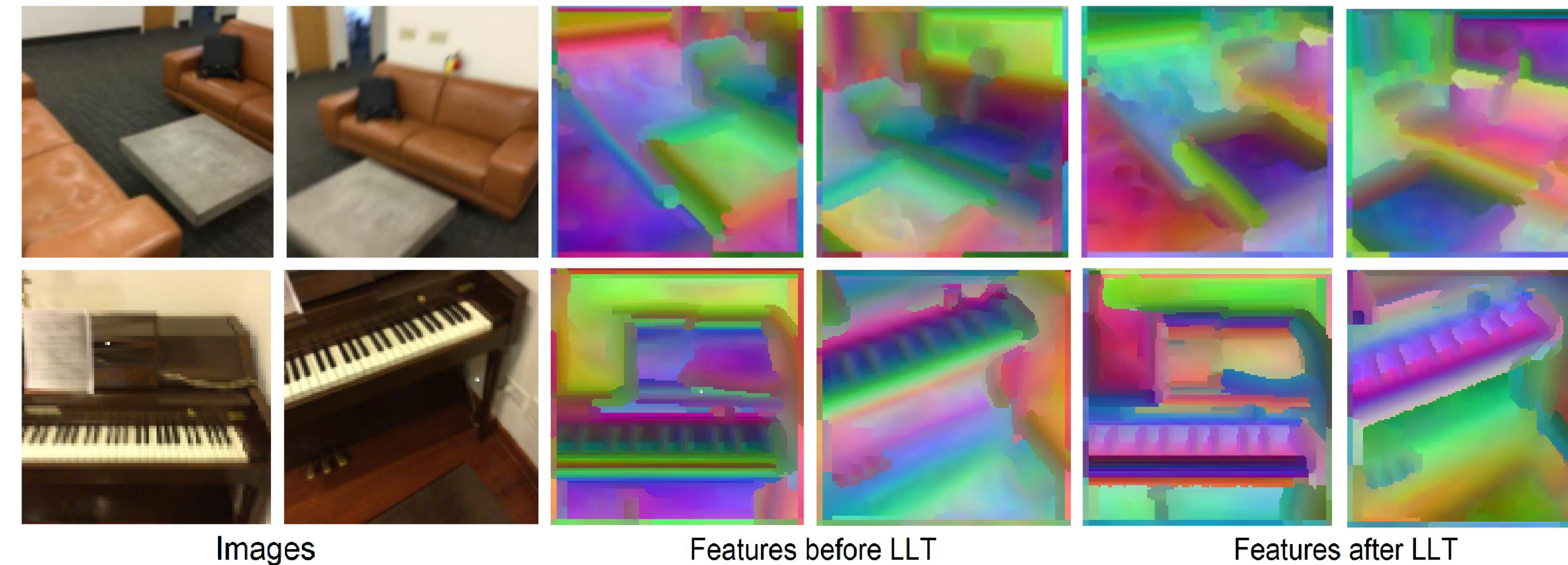
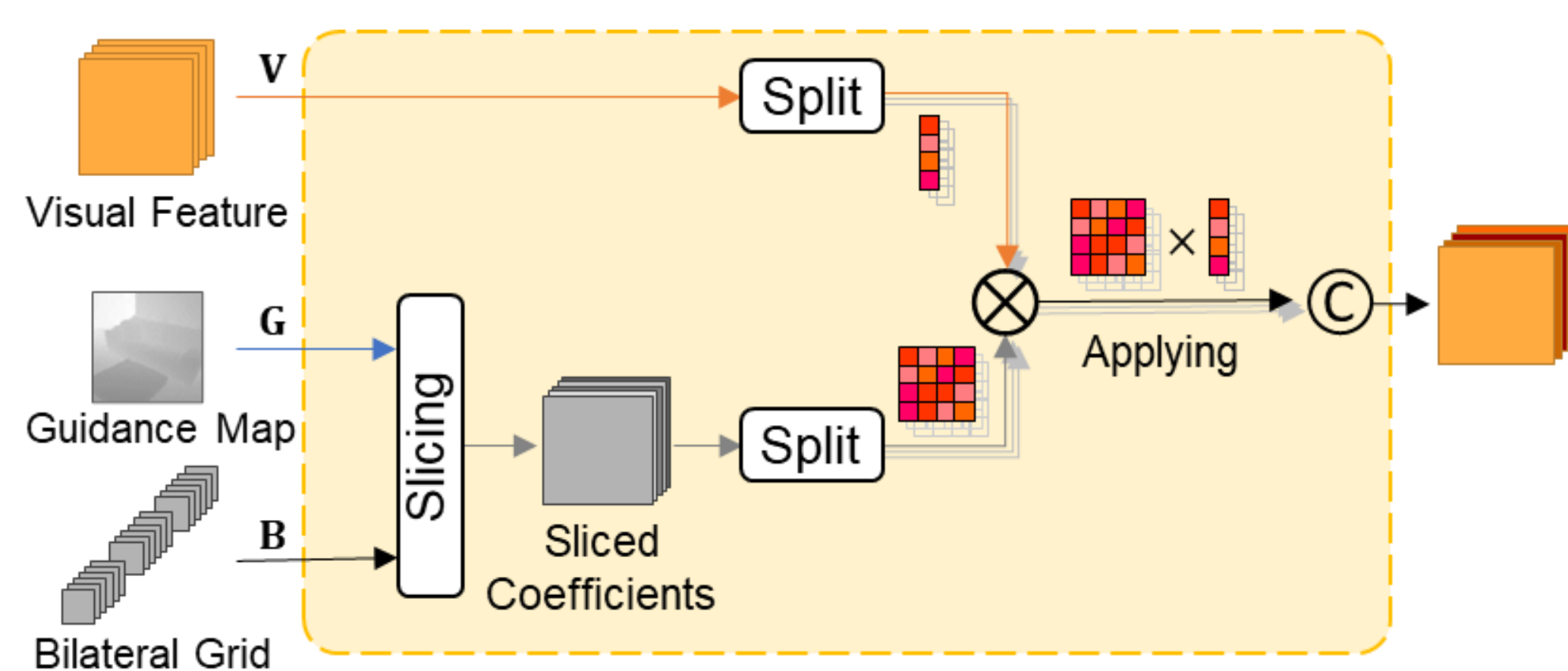
Experimental Results

Methods	Train Set	Sup	Features		Rotation			Translation			Chamfer		
			Visual	3D	5	10	45	5	10	25	1	5	10
SIFT [27]	N/A		✓		55.2	75.7	89.2	17.7	44.5	79.8	38.1	70.6	78.3
SuperPoint [14]	N/A		✓		65.5	86.9	96.6	21.2	51.7	88.0	45.7	81.1	88.2
FCGF [10]	N/A			✓	70.2	87.7	96.2	27.5	58.3	82.9	52.0	78.0	83.7
BYOC [16]	3D Match		✓	✓	66.5	85.2	97.8	30.7	57.6	88.9	54.1	82.8	89.5
DGR [5]	3D Match	✓		✓	81.1	89.3	94.8	54.5	76.2	88.7	70.5	85.5	89.0
3D MV Reg [19]	3D Match	✓		✓	87.7	93.2	97.0	69.0	83.1	91.8	78.9	89.2	91.8
UR&R [15]	3D Match		✓		87.6	93.1	98.3	69.2	84.0	93.8	79.7	91.3	94.0
UR&R (RGB-D) [15]	3D Match		✓	✓	87.6	93.7	98.8	67.5	83.8	94.6	78.6	91.7	94.6
Ours	3D Match		✓	✓	93.4	96.5	98.8	76.9	90.2	96.7	86.4	95.1	96.8



Local Linear Transformation (LLT)

- We progressively apply the local linear transformation to learn the visual-geometric features in a multi-scale manner.
- The slicing operation is performed between a **Bilateral Grid** and the **guidance map**.
- The final local linear transformation can be obtained by first using a point-wise transformation and then using a channel-wise concatenation



Ablation Study

- **Component effectiveness**
- **MS**: multi-scale strategy
- **DC**: dilated convolutions in the visual feature extractor

MS	DC	LLT	Rotation					Translation					Chamfer				
			Accuracy		Error		Mean	Accuracy		Error		Mean	Accuracy		Error		Mean
			5	10	45	Mean		5	10	25	Mean		1	5	10	Mean	
			88.4	94.2	98.6	3.8	1.1	67.3	83.8	94.5	8.5	3.0	78.9	91.7	94.6	6.5	0.2
✓			88.5	94.4	98.6	3.8	1.1	68.1	84.5	94.8	8.3	3.0	79.5	92.1	94.9	6.3	0.2
✓	✓		90.4	95.0	98.6	3.6	1.0	70.8	86.5	95.3	8.1	2.8	81.8	93.1	95.4	6.2	0.2
✓	✓	✓	93.4	96.5	98.8	3.0	0.9	76.9	90.2	96.7	6.4	2.4	86.4	95.1	96.8	5.3	0.1

- **Optimized in supervised manner**

Methods	Rotation					Translation					Chamfer				
	Accuracy		Error		Mean	Accuracy		Error		Mean	Accuracy		Error		Mean
	5	10	45	Mean		5	10	25	Mean		1	5	10	Mean	
UR&R (RGB-D)	92.3	95.3	98.2	3.8	0.8	77.6	89.4	95.5	7.8	2.3	86.1	94.0	95.6	6.7	0.1
Ours	96.5	97.8	98.8	2.7	0.8	83.8	93.8	97.6	5.8	2.0	91.2	96.7	97.6	4.8	0.1