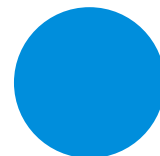
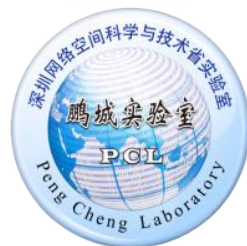
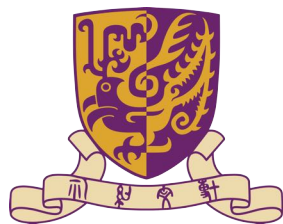
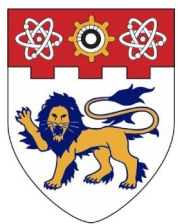
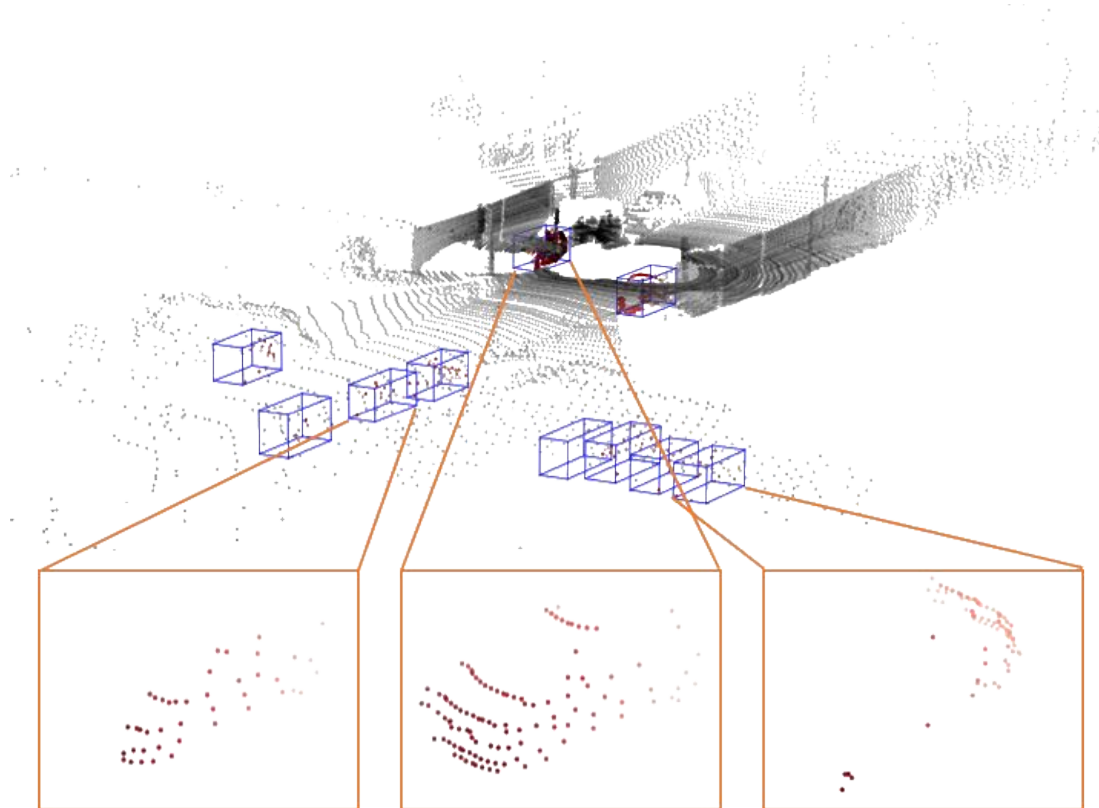


GRNet: Gridding Residual Network for Dense Point Cloud Completion

Haozhe Xie, Hongxun Yao, Shangchen Zhou,
Jiageng Mao, Shengping Zhang, Wenxiu Sun

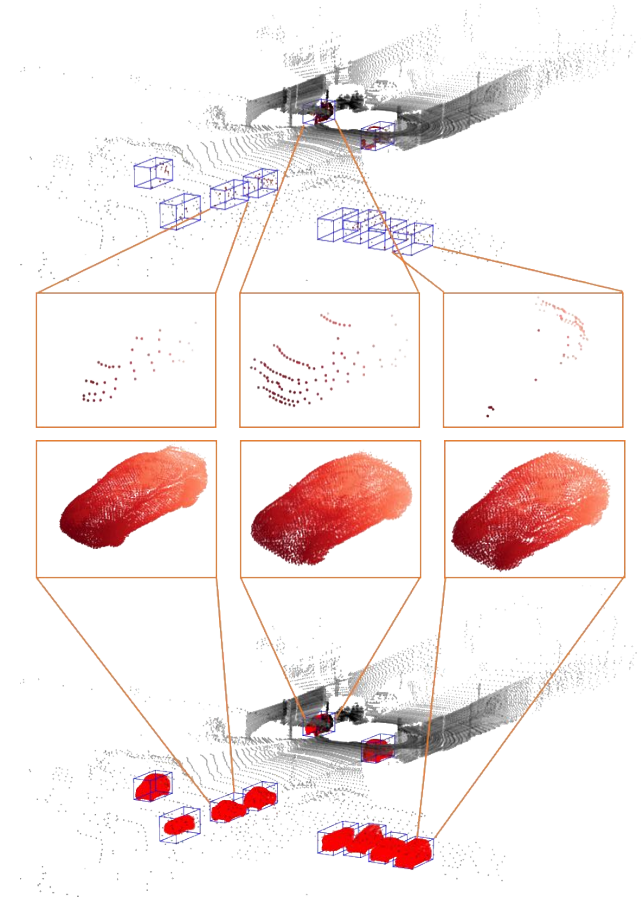
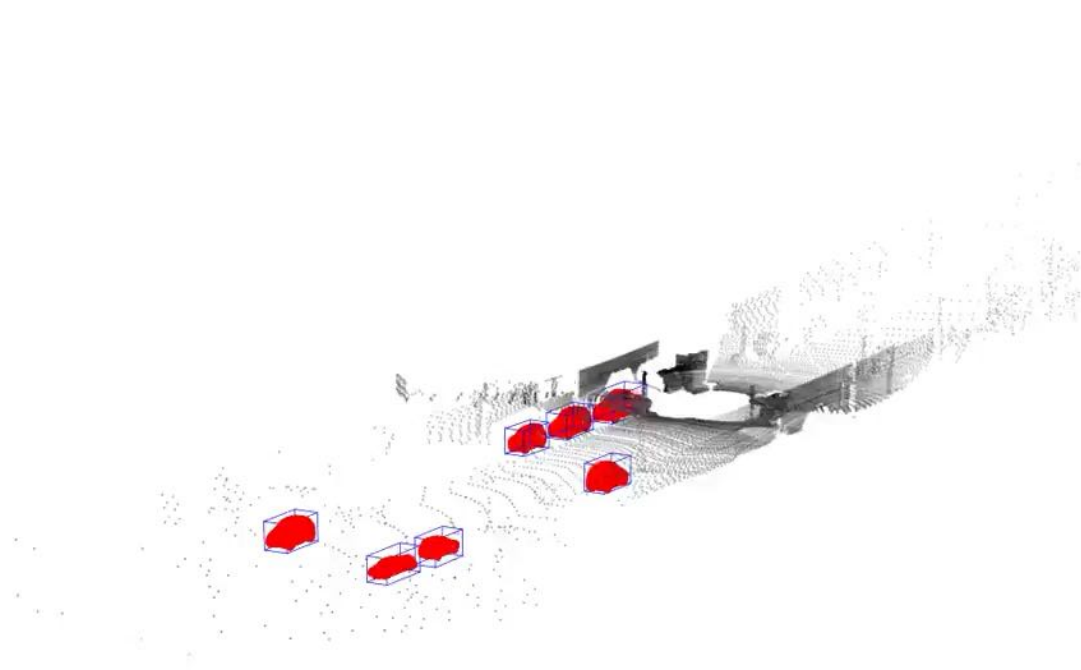


What is Point Cloud Completion?



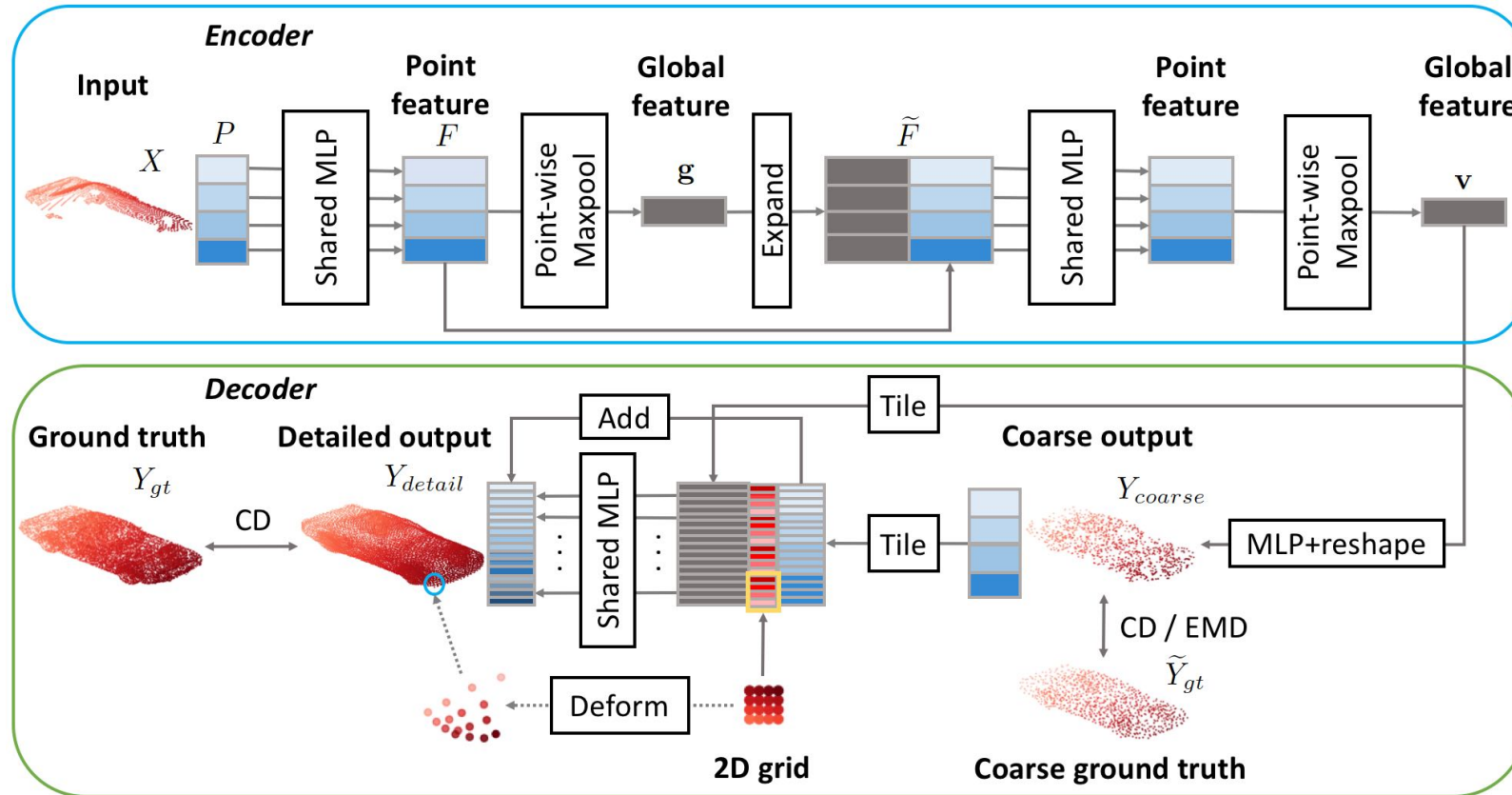
Source: PCN: Point Completion Network

What is Point Cloud Completion?



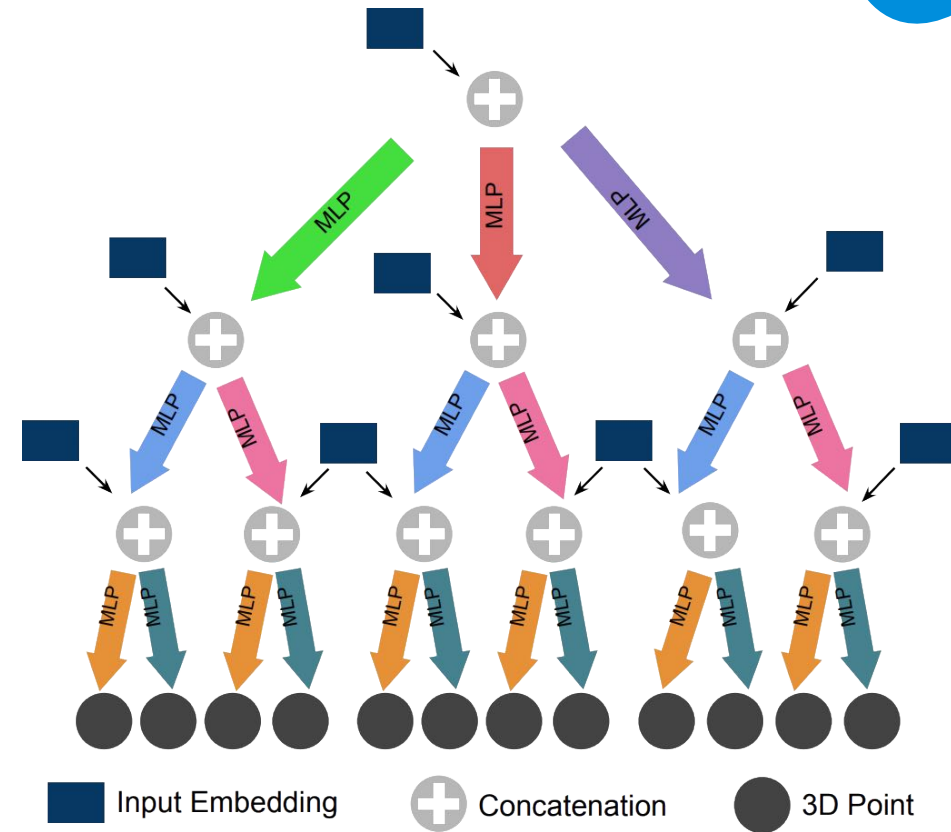
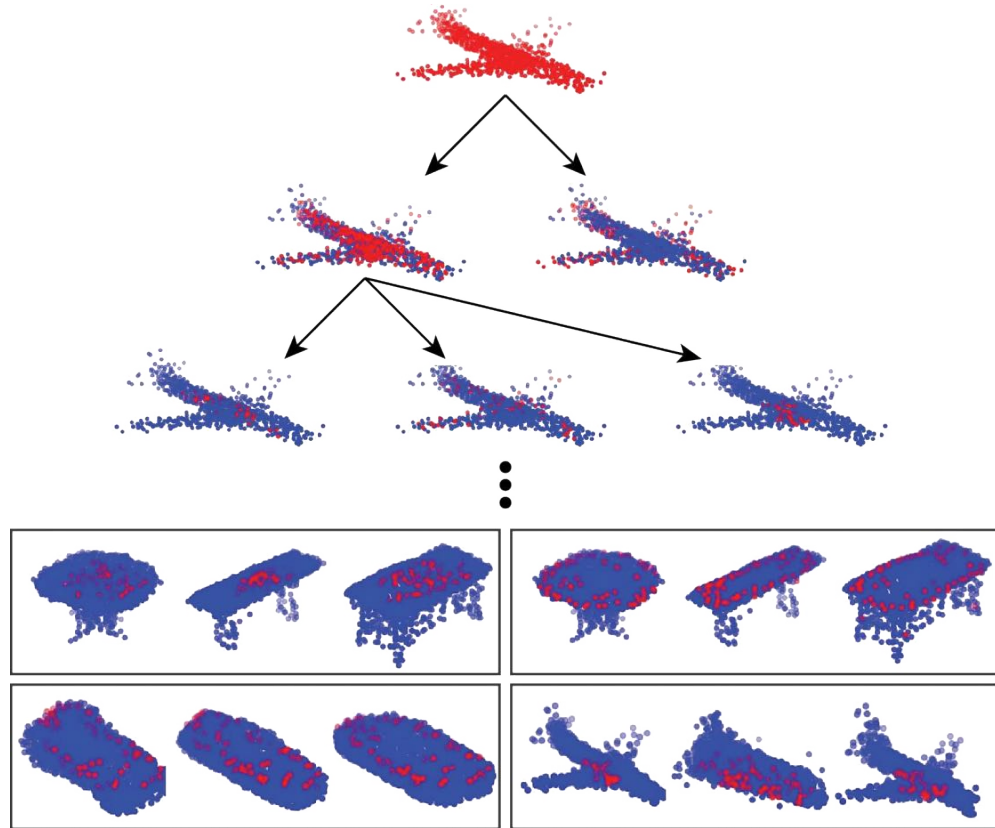
Source: PCN: Point Completion Network

State-of-the-arts



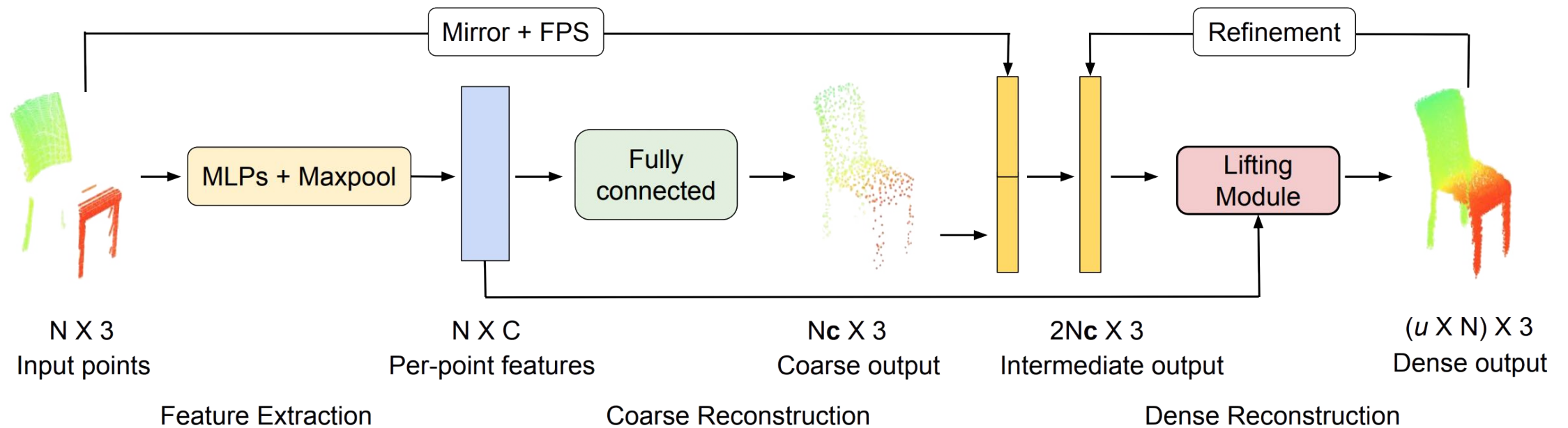
Yuan *et al.* PCN: Point Completion Network. 3DV 2018.

State-of-the-arts



Tchapmi *et al.* TopNet: Structural Point Cloud Decoder. CVPR 2019.

State-of-the-arts



Wang *et al.* Cascaded Refinement Network for Point Cloud Completion. CVPR 2020.

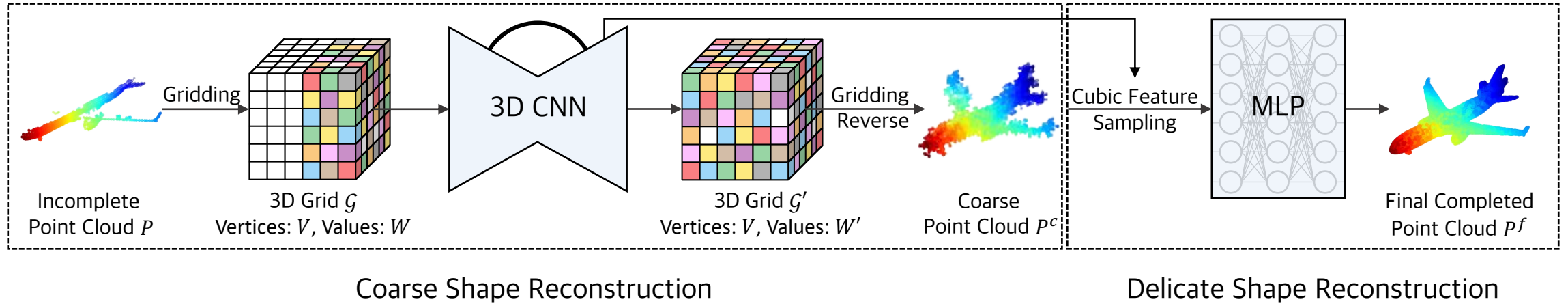
Drawbacks of MLP-based Methods



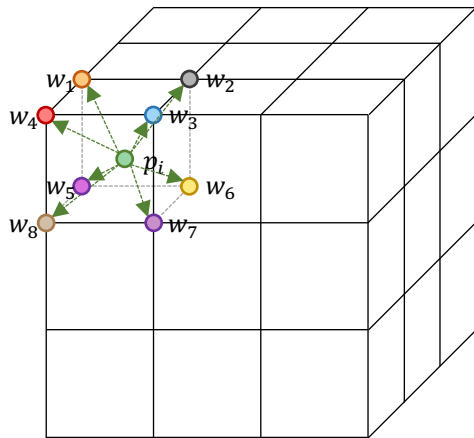
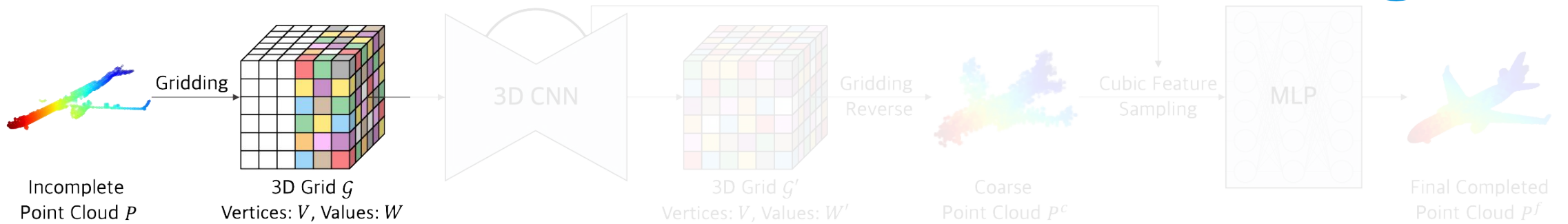
MLP-based methods ignore two important things:

- Geometric Structure
- Context of Neighboring Points

The Proposed Method: GRNet



Gridding

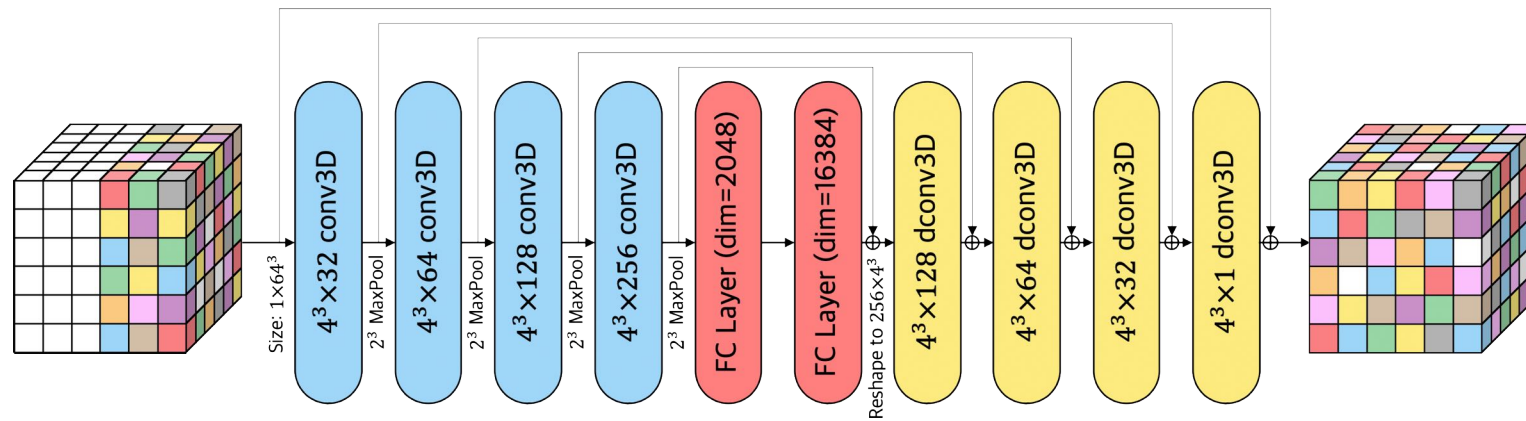
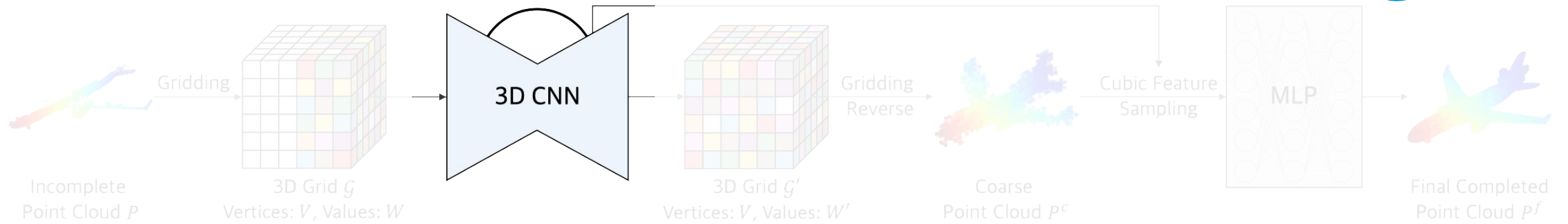


The value of the vertex w_i can be computed as

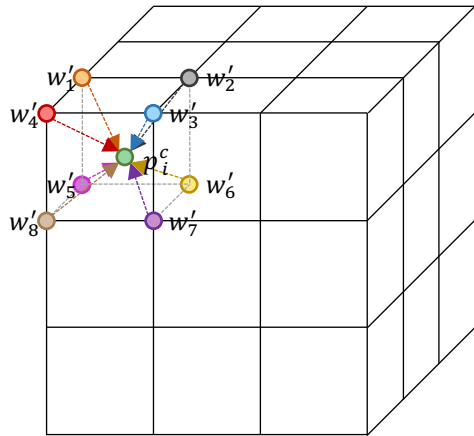
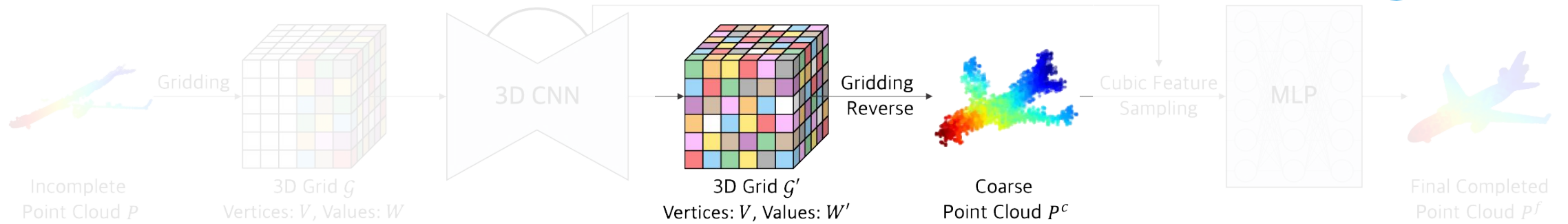
$$w_i = (1 - |x_i - x|)(1 - |y_i - y|)(1 - |z_i - z|)$$

where (x_i, y_i, z_i) and (x, y, z) are the coordinates of w_i and p_i , respectively.

3D-CNN



Gridding Reverse

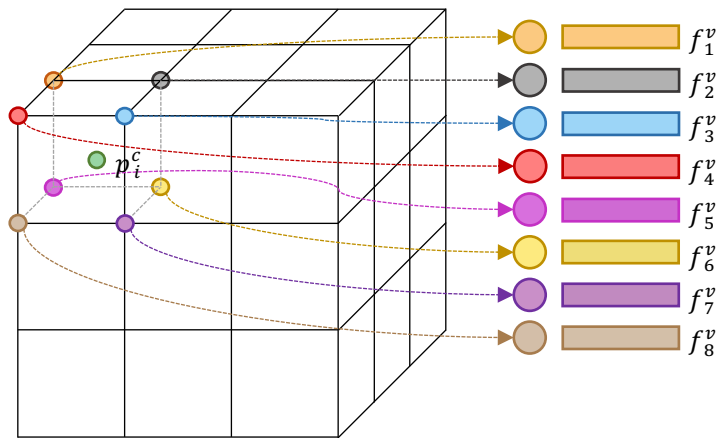
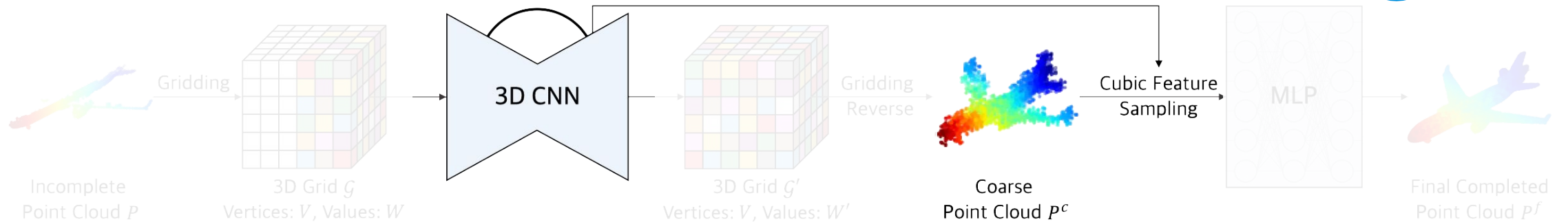


The coordinate of the generated point p_i^c can be computed as

$$p_i^c = \frac{\sum_i w'_i v_i}{\sum_i w'_i}$$

where the v_i and w'_i be the coordinate and value of the vertex i ($i = 1, \dots, 8$).

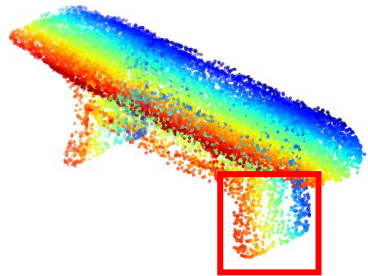
Cubic Feature Sampling



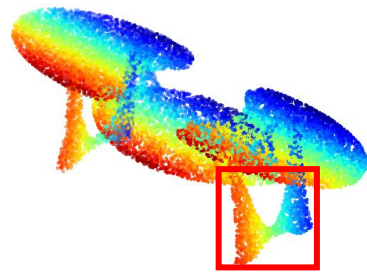
Therefore, the features f_i^c for point p_i^c can be computed as:

$$f_i^c = [f_1^v, f_2^v, \dots, f_8^v]$$

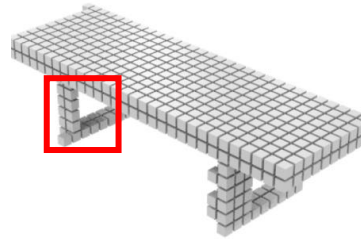
Gridding Loss



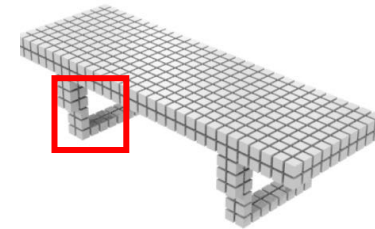
Reconstruction



Ground Truth



Reconstruction

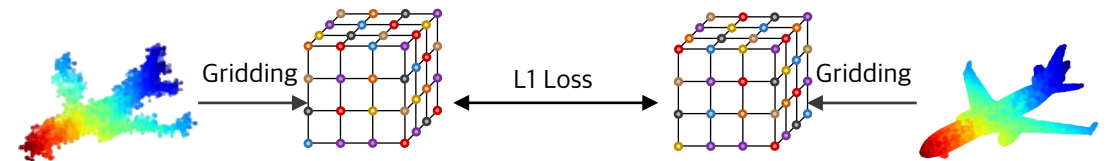


Ground Truth

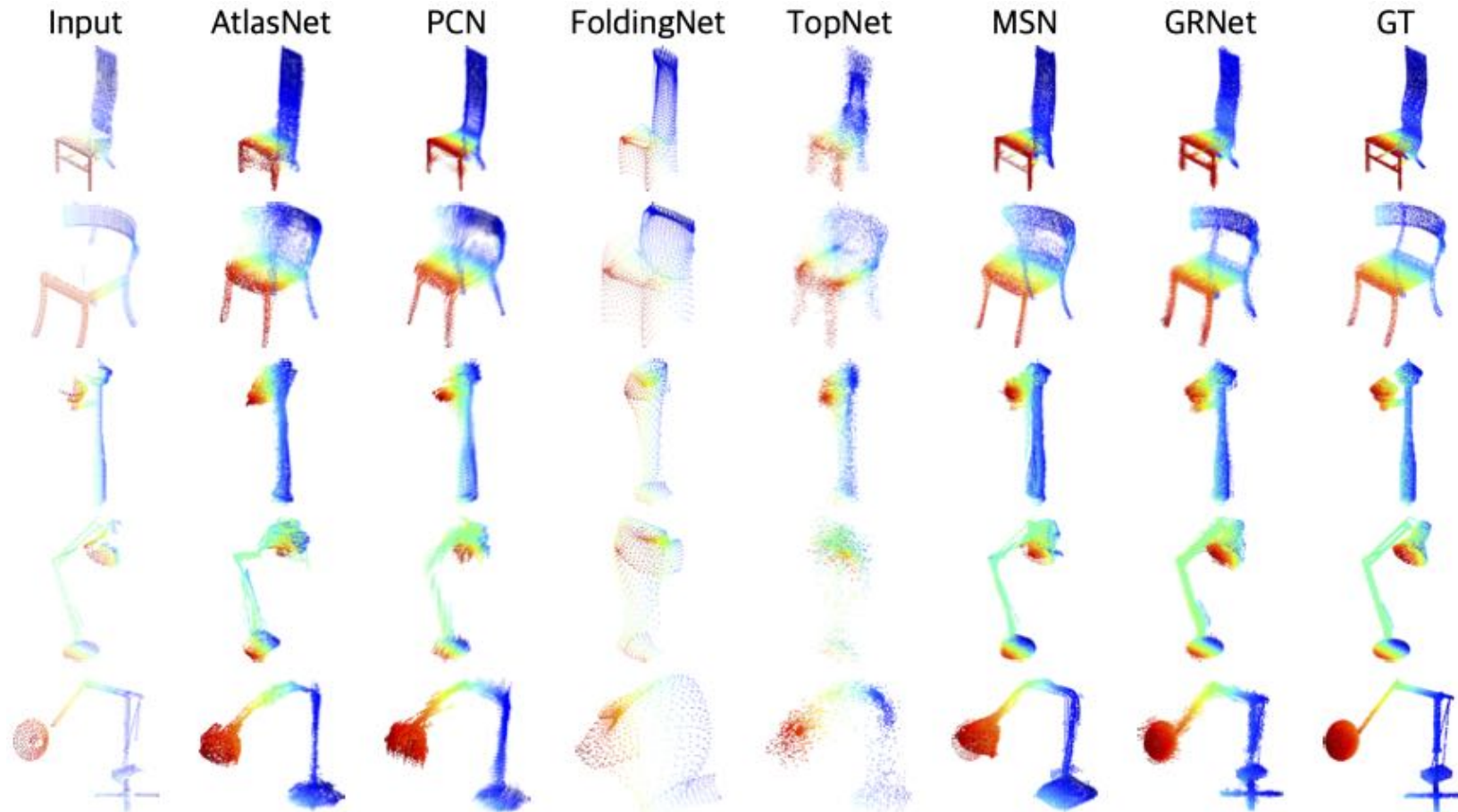
Chamfer Distance

$$\ell = \frac{1}{|P|} \sum_{p \in P} \min_{q \in Q} |p - q| + \frac{1}{|Q|} \sum_{q \in Q} \min_{p \in P} |p - q|$$

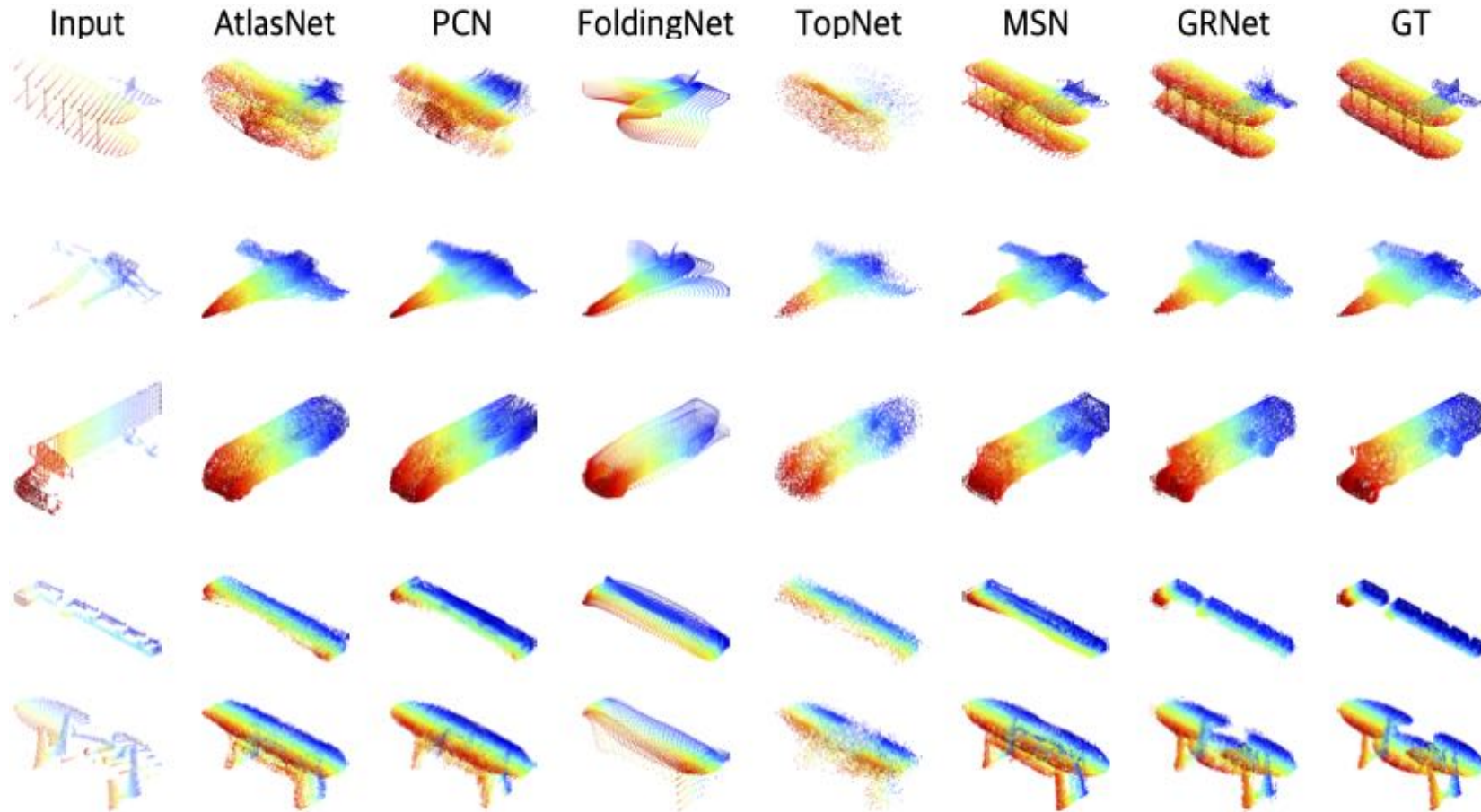
Gridding Loss



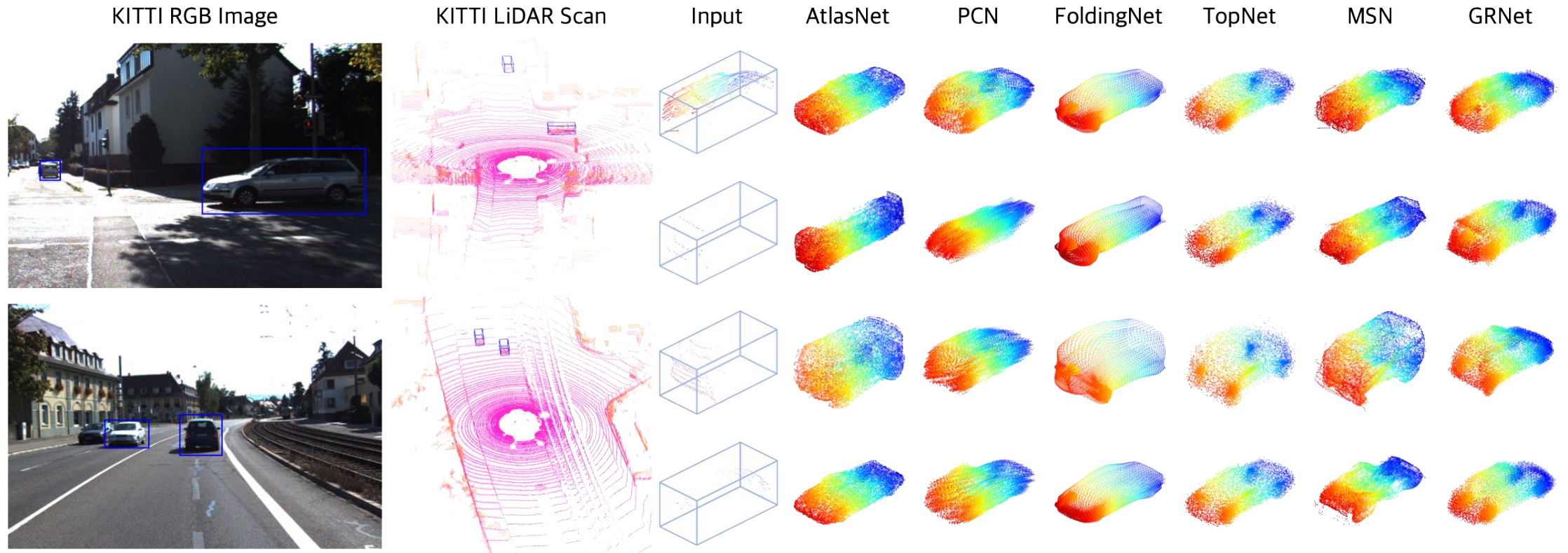
Qualitative Results on ShapeNet



Qualitative Results on ShapeNet



Qualitative Results on KITTI



Quantitative Results on ShapeNet

Chamfer Distance (L2)

Methods	Airplane	Cabinet	Car	Chair	Lamp	Sofa	Table	Watercraft	Mean
AtlasNet	1.753	5.101	3.237	5.226	6.342	5.990	4.359	4.177	4.523
PCN	1.400	4.450	2.445	4.838	6.238	5.129	3.569	4.062	4.016
FoldingNet	3.151	7.943	4.676	9.225	9.324	8.895	6.691	7.325	7.142
TopNet	2.152	5.623	3.513	6.346	7.502	6.949	4.784	4.359	5.154
MSN	1.543	7.249	4.711	4.539	6.479	5.894	3.797	3.853	4.758
GRNet	1.531	3.620	2.752	2.945	2.649	3.613	2.552	2.122	2.723

Quantitative Results on ShapeNet

F-Score@1%

Methods	Airplane	Cabinet	Car	Chair	Lamp	Sofa	Table	Watercraft	Mean
AtlasNet	0.845	0.552	0.630	0.552	0.565	0.500	0.660	0.624	0.616
PCN	0.881	0.651	0.725	0.625	0.638	0.581	0.765	0.697	0.695
FoldingNet	0.642	0.237	0.382	0.236	0.219	0.197	0.361	0.299	0.322
TopNet	0.771	0.404	0.544	0.413	0.408	0.350	0.572	0.560	0.503
MSN	0.885	0.644	0.665	0.657	0.699	0.604	0.782	0.708	0.705
GRNet	0.843	0.618	0.682	0.673	0.761	0.605	0.751	0.750	0.708

Leaderboard on Completion 3D

Chamfer Distance (L2)

Method	CD(10^{-4})	Airplane	Cabinet	Car	Chair	Lamp	Sofa	Table	Watercraft
GRNet	10.64	6.13	16.90	8.27	12.23	10.22	14.93	10.08	5.86
Haozhe Xie, Hongxun Yao, Shangchen Zhou, Jiageng Mao, Shengping Zhang, Wenxiu Sun: GRNet: Gridding Residual Network for Dense Point Cloud Completion. ECCV 2020									
TopNet	14.25	7.32	18.77	12.88	19.82	14.60	16.29	14.89	8.82
Lyne P. Tchapmi, Vineet Kosaraju, S. Hamid Rezaatofghi1, Ian Reid, Silvio Savarese: TopNet: Structural Point Cloud Decoder. CVPR19									
PointNetFCAE(Topnet-baseline)	16.88	10.30	19.06	11.82	24.68	20.30	20.09	17.57	10.50
N/A: TopNet-Baseline. N/A									
AtlasNet(Topnet-baseline)	17.77	10.36	23.40	13.40	24.16	20.24	20.82	17.52	11.62
N/A: TopNet-Baseline. N/A									
PCN(Topnet-baseline)	18.22	9.79	22.70	12.43	25.14	22.72	20.26	20.27	11.73
N/A: TopNet-Baseline. N/A									
Folding(Topnet-baseline)	19.07	12.83	23.01	14.88	25.69	21.79	21.31	20.71	11.51
N/A: TopNet-Baseline. N/A									

Reference: <https://completion3d.stanford.edu/results>



Thank you!

Project Page:

<https://haozhxie.com/project/grnet/>

