

# High Resolution UDF Meshing via Iterative Networks

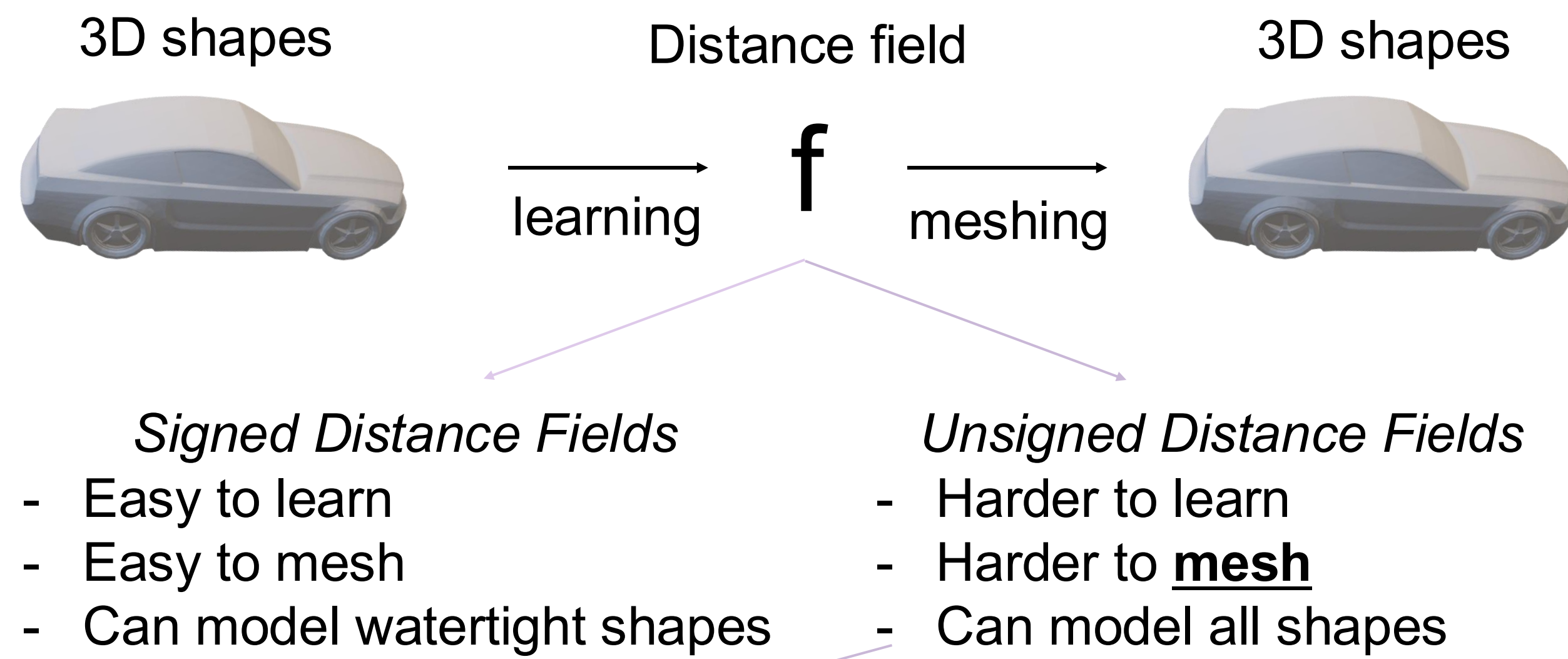
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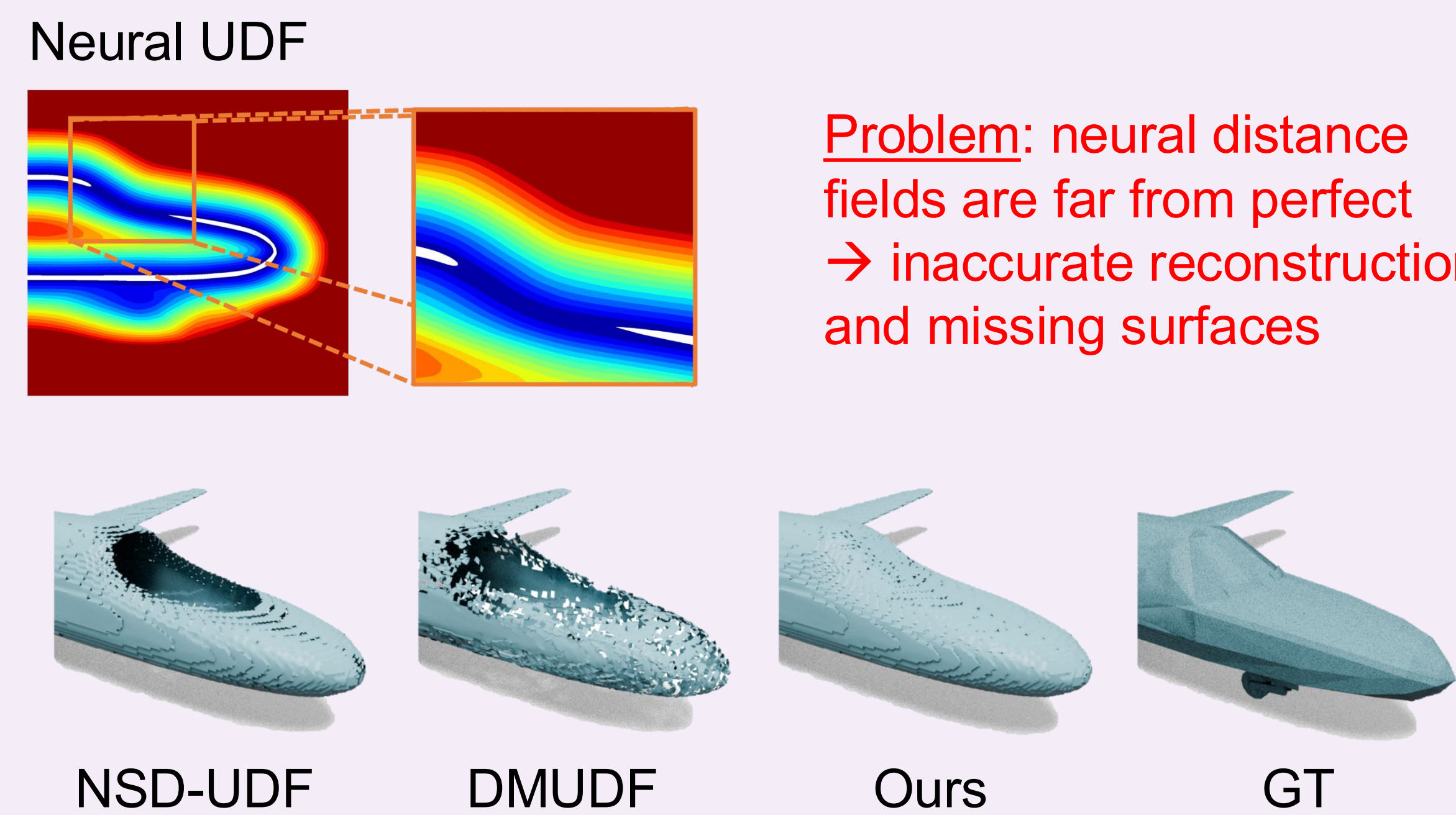
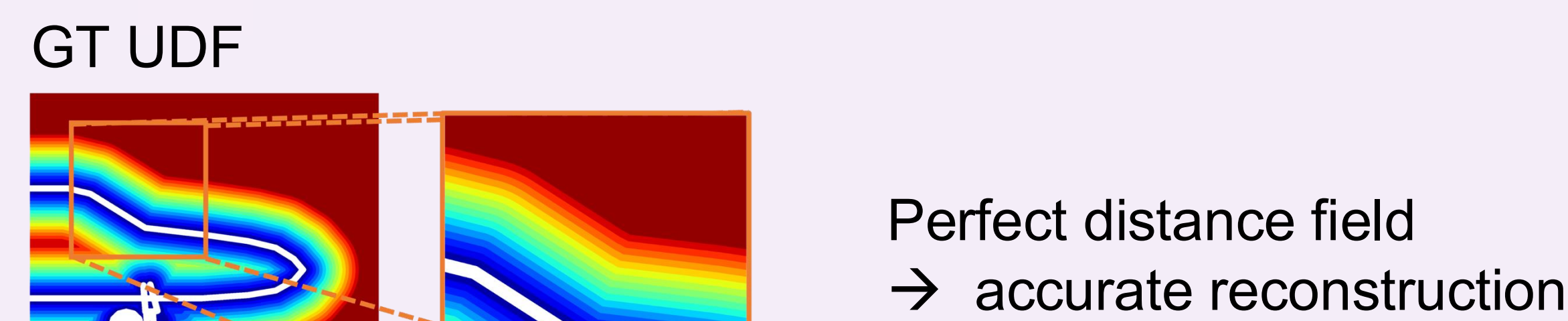
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NEURAL INFORMATION  
PROCESSING SYSTEMS

## 1. Meshing Unsigned Distance Fields

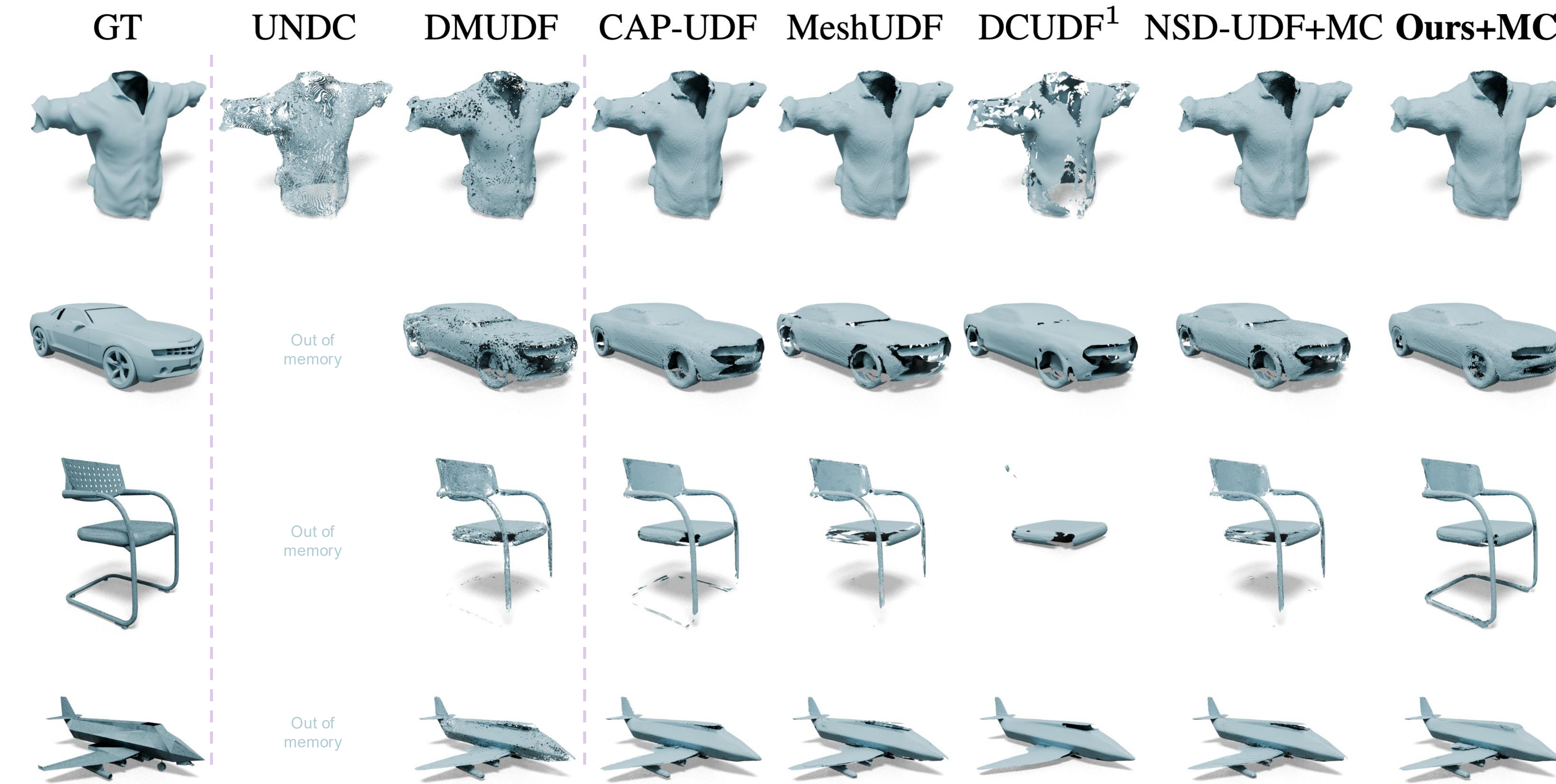


### Meshing Unsigned Distance Fields (our focus)

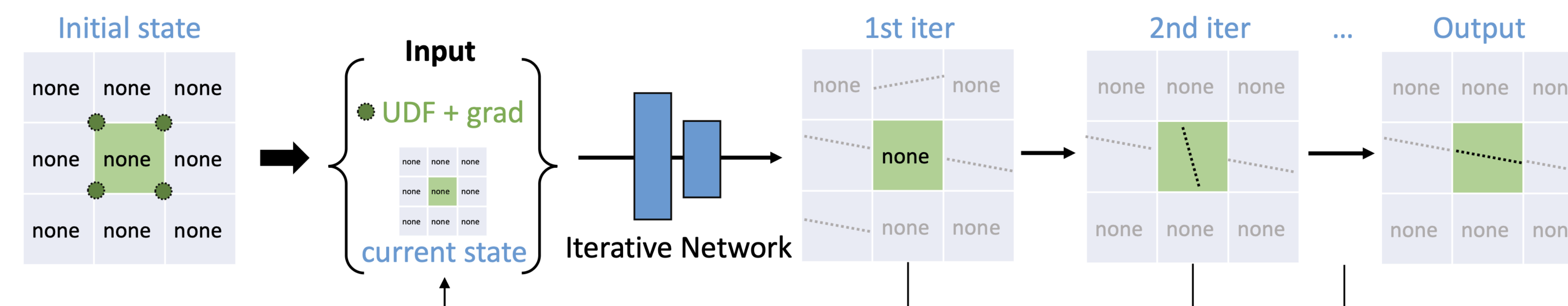


## 2. Increasing resolution worsens the problem

Counterintuitively, this problem manifests more at higher resolutions, or with noisier fields



## 3. Proposed Method

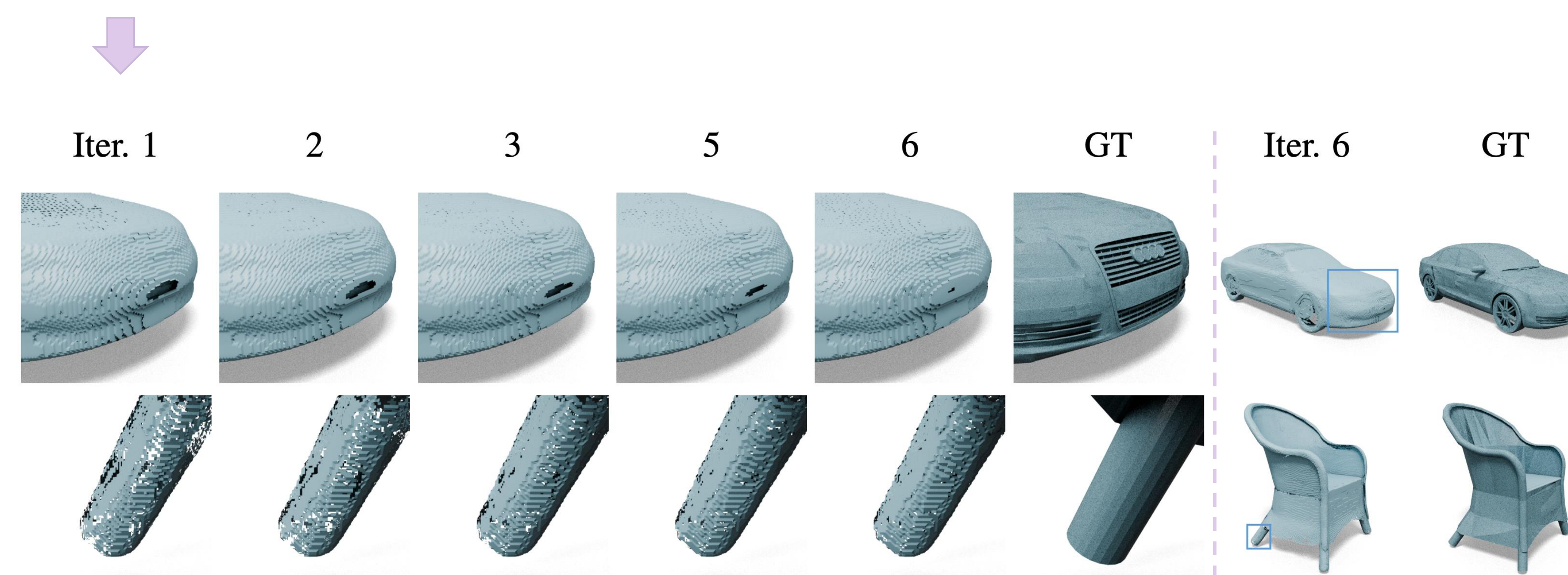


- We formulate high-resolution meshing as an iterative process: each iteration takes the previous output state as input and refines it
- The mesh is improved over multiple iterations, where each step integrates newly detected surfaces, distance values, and gradients from neighboring cells

$$y_{S,c}^{(i)} = f_{\theta} \left( U_S(c), \nabla U_S(c), \sigma(y_{S,N_c}^{(i-1)}) \right),$$

$$y_{S,N_c}^{(i-1)} = \left\| \begin{matrix} y_{S,c'}^{(i-1)} \\ c' \in N_c \end{matrix} \right\|,$$

$$y_{S,c}^{(0)} = [0, 0, \dots, 0],$$



## 4. Results

### a) MC-based methods

Res.	Method	MGN <sup>2</sup>			ShapeNet cars			ShapeNet chairs			ShapeNet planes		
		CD ↓	F1 ↑	IC ↑	CD ↓	F1 ↑	IC ↑	CD ↓	F1 ↑	IC ↑	CD ↓	F1 ↑	IC ↑
128	CAP-UDF	16.2	69.4	79.4	53.9	53.3	83.2	37.8	50.9	70.3	14.0	69.5	80.6
	MeshUDF	2.45	82.9	94.1	11.3	57.7	88.6	7.43	67.6	88.0	6.82	74.6	81.0
	DCUDF	13500	2.50	2.76	6580	7.74	14.7	17000	15.9	13.9	1880	27.6	24.5
	DCUDF-T	90.8	2.86	87.3	50.3	58.0	87.3	214	64.1	82.9	144	68.8	77.2
	DCUDF-T-acut	-	-	-	-	-	-	-	-	-	-	-	-
	NSD-UDF+MC	1.34	83.9	94.7	6.79	59.6	88.5	6.11	67.8	88.1	3.82	79.0	84.8
256	Ours + MC	2.04	81.9	94.1	5.64	59.2	88.9	3.68	66.7	88.4	3.00	78.5	84.8
	CAP-UDF	1.66	86.8	91.8	34.0	61.2	87.6	114	70.5	82.0	5.50	83.7	85.4
	MeshUDF	0.958	89.7	95.0	13.6	62.3	88.6	27.8	72.9	87.3	3.47	85.8	85.2
	DCUDF	14400	4.76	5.71	346	52.6	78.2	3530	49.9	54.3	27.9	84.7	82.0
	DCUDF-T	4.63	86.8	95.4	347	52.6	78.2	3560	50.0	54.3	47.3	80.4	78.7
	DCUDF-T-acut	-	-	-	-	-	-	-	-	-	-	-	-
512	NSD-UDF+MC	0.808	90.0	95.2	10.2	62.0	87.9	10.9	72.3	86.2	2.91	87.3	86.0
	Ours + MC	0.878	88.9	94.9	5.23	65.0	89.2	5.14	72.9	88.8	1.84	88.7	87.0
	CAP-UDF	0.872	90.6	94.6	31.8	61.7	87.5	63.9	71.7	82.0	5.94	87.5	86.2
	MeshUDF	0.798	90.6	94.8	82.7	57.0	81.7	378	61.5	65.7	12.6	88.1	84.6
	DCUDF	4.37	88.3	91.1	223	56.5	84.2	2950	55.0	70.1	48.7	85.5	82.2
	DCUDF-T	4.38	88.2	91.1	223	56.5	84.2	2000	55.0	70.1	63.0	85.4	81.3

### b) DC-based methods

Res.	Method	MGN <sup>2</sup>			ShapeNet cars			ShapeNet chairs			ShapeNet planes		
		CD ↓	F1 ↑	IC ↑	CD ↓	F1 ↑	IC ↑	CD ↓	F1 ↑	IC ↑	CD ↓	F1 ↑	IC ↑
128	UNDC	1.09	87.1	94.1	13.5	61.7	86.4	29.9	69.4	81.9	2.50	82.0	86.1
	DualMesh-UDF	216	68.1	68.4	952	34.4	45.5	7930	12.1	9.08	112	74.3	76.1
	DualMesh-UDF-T	0.806	89.9	95.4	5.56	63.5	89.5	5.34	75.1	89.1	1.96	84.3	87.5
	NSD-UDF+DualMesh-UDF	0.760	90.4	95.0	6.34	65.7	89.1	5.50	72.8	89.2	2.08	87.1	86.6
	Ours + DualMesh-UDF	0.787	90.5	94.9	4.80	66.2	89.7	3.39	72.8	89.8	1.56	87.7	88.2
	DCUDF	0.931	89.1	91.5	82.4	52.3	71.4	293	54.2	57.8	11.6	83.6	80.7
256	UNDC	176	66.3	66.4	846	34.0	45.1	8280	12.4	9.26	105	77.9	76.0
	DualMesh-UDF	0.722	91.2	95.1	10.6	64.3	87.0	22.8	72.2	84.1	2.43	88.2	87.0
	DualMesh-UDF-T	0.671	91.0	94.6	10.5	64.9	87.2	14.6	70.7	84.4	2.78	88.9	85.6
	NSD-UDF+DualMesh-UDF	0.662	91.2	94.7	5.48	65.7	88.8	4.97	71.9	86.4	1.87	90.0	87.5
	Ours + DualMesh-UDF	0.662	91.2	94.7	5.48	65.7	88.8	4.97	71.9	86.4	1.87	90.0	87.5
	DCUDF	0.722	91.2	94.7	5.48	65.7	88.8	4.97	71.9	86.4	1.87	90.0	87.5
512	UNDC	2.39	84.8	82.8	-	-	-	-	-	-	-	-	-
	DualMesh-UDF	167	63.7	63.9	870	32.5	43.0	8190	11.8	9.07	111	77.6	74.2
	DualMesh-UDF-T	0.827	90.2	93.2	37.8	58.2	79.3	72.7	63.7	72.5	4.77	89.0	84.0
	NSD-UDF+DualMesh-UDF	0.787	89.7	92.5	60.5	57.5	79.9	296	62.5	68.0	9.95	87.6	83.3
	Ours + DualMesh-UDF	0.726	89.7	92.8	9.65	63.0	85.6	10.1	70.4	81.9	2.51	90.1	85.8
	DCUDF	0.726	89.7	92.8	9.65	63.0	85.6	10.1	70.4	81.9	2.51	90.1	85.8

### c) Filtering strategies and speed

Res.	Filtering	CD ↓	IC ↑	Inference time ↓
256	Without filtering	5.26	89.2	7m
	Low confidence	5.24	89.2	1.5m
	UDF < 0.1 and low confidence	5.24	89.2	30s
	UDF < 0.1 and low confidence	5.24	89.2	30s
512	Without filtering	8.91	88.9	1h
	Low confidence	8.87	88.9	7m
	UDF < 0.1 and low confidence	8.87	88.9	1h
	UDF < 0.1 and low confidence	8.88	88.9	2.5m

Time	CAP	MeshUDF	DCUDF	NSD	UNDC	DMUDF	Ours
Query	90s	30s	/	30s	22s	19s	30s
Total	3.5m	35s	25m	35s	30s	20s	1m

<sup>1</sup>Run without the cutting step on cars, as recommended, producing a more complete but double-layered surface  
<sup>2</sup>Resolution is halved for MGN garments due to their simpler shapes

## References

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