



SDFoam: Signed-Distance Foam for explicit surface reconstruction

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Overview



Figure 1: Existing methods in literature reconstruct 3D scenes either by employing explicit or implicit geometry, each with their own advantages and drawbacks. Our method, **SDFoam**, jointly learns a signed distance field (SDF) and a 3D Voronoi Diagram (or foam), both of which are optimized during a ray-tracing process. Our method offers a good trade-off of rendering speed, visual fidelity and reconstruction accuracy.

Our contribution

- A **single representation** for view synthesis and geometry, producing cleaner, view-consistent surfaces with fewer artifacts.
- Novel **hybrid implicit-explicit** representation combines the Voronoi Diagram with SDF for better geometry alignment.
- Enables fast mesh extraction, up to **5× faster** than a naive density-based thresholding on Radiant Foam (RF).

Architecture

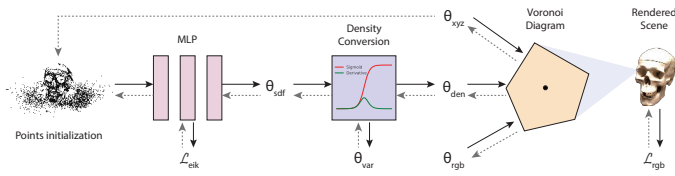


Figure 2: **SDFoam Architecture**. A point cloud is initialized and refined over time by learning an SDF from its points. Their SDF values are then converted to density, which are used jointly with color and position parameters to learn a ray-traced scene.

Results - Novel View Synthesis

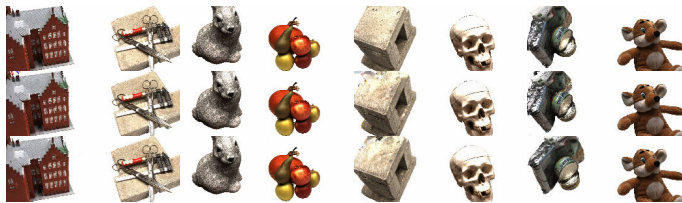


Figure 3: Novel View synthesis. Top: Ground truth. Center: RF. Bottom: SDFoam.

Scan ID	24	37	40	55	63	65	69	83	97	105	106	110	114	118	122	Mean
PSNR (IDF _{0.1})	24.36	22.39	26.34	26.77	28.38	24.81	22.96	28.33	21.33	27.03	23.72	24.21	24.78	25.93	28.84	25.34
PSNR (SDF _{Coarse})	25.59	24.96	26.55	26.27	28.81	25.78	23.77	28.90	26.12	28.30	24.08	25.80	27.13	27.58	29.29	26.60
SSIM (IDF _{0.1})	0.796	0.770	0.823	0.825	0.942	0.901	0.880	0.952	0.839	0.917	0.892	0.907	0.866	0.927	0.938	0.885
SSIM (SDF _{Coarse})	0.803	0.828	0.814	0.900	0.937	0.911	0.870	0.949	0.904	0.914	0.886	0.904	0.884	0.922	0.939	0.891

Figure 4: Our joint Voronoi-SDF formulation achieves better results and acts as an additional regularization to boost visual appearance, removing occasional floaters in the scene, as shown in Fig. 3.

Qualitative Results

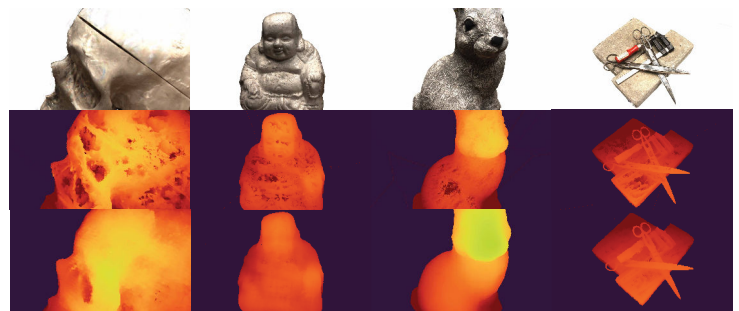


Figure 5: Mesh reconstruction qualitative results. Top to bottom: ground truth, Radiant Foam, SDFoam. Modelling the voronoi cells as local SDFs improves the consistency of the extracted surface, thus filling the typical *holes* derived from the ray-tracing procedure in Radiant Foam.

Results - 3D Reconstruction

ScanID	Chamfer Distance (w/ mask) ↓					PSNR ↑		SSIM ↑	
	IDR	NeRF	NeuS	RF*	SDFoam	RF	SDFoam	RF	SDFoam
scan27	1.63	1.83	0.83	6.13	1.86	31.28	29.80	0.877	0.848
scan34	1.87	2.39	0.98	3.53	2.87	31.58	30.42	0.921	0.899
scan40	0.63	1.79	0.56	6.02	1.80	32.12	30.68	0.886	0.851
scan55	0.48	0.66	0.37	1.31	0.86	33.04	32.01	0.957	0.946
scan63	1.04	1.79	1.13	7.10	2.23	35.97	35.26	0.969	0.962
scan65	0.79	1.44	0.59	2.10	1.52	32.96	32.43	0.958	0.952
scan69	0.77	1.50	0.60	2.84	1.30	29.18	28.17	0.937	0.907
scan83	1.33	1.20	1.45	7.54	1.27	33.10	32.82	0.974	0.970
scan97	1.16	1.96	0.95	5.41	1.53	31.25	30.12	0.974	0.931
scan105	0.76	1.27	0.78	8.16	2.16	32.37	32.15	0.953	0.945
scan106	0.67	1.44	0.52	6.79	1.67	29.08	28.41	0.948	0.931
scan110	0.90	2.61	1.43	2.54	3.03	29.86	29.37	0.954	0.941
scan114	0.42	1.04	0.36	1.78	1.15	32.28	31.01	0.946	0.930
scan118	0.51	1.13	0.45	1.96	1.52	31.73	31.01	0.968	0.951
scan122	0.53	0.99	0.45	1.82	1.34	34.82	34.12	0.976	0.969
mean	0.90	1.54	0.77	4.33	1.74	32.04	31.18	0.947	0.929

Figure 6: Quantitative evaluation on the DTU dataset. *RF → large Chamfer distances are due to floaters being difficult to filter with a naive density-based thresholding. Our method offers a good trade-off between mesh reconstruction and visual fidelity.

