

Flexible Isosurface Extraction for Gradient-Based Mesh Optimization

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ISOSURFACING

Gradient



GENERATING MESH VIA GRADIENT-BASED OPTIMIZATION

Why not directly optimize the implicit field?



Evaluating objectives on meshes:

Efficient, benefit from mesh representation.

 \checkmark Ready to use in downstream applications.



GENERATING MESH VIA GRADIENT-BASED OPTIMIZATION

Why not directly optimizing the vertex positions of a mesh?





- Using implicit field to parametrize surface:
 - Support Arbitrary topology.
 - \checkmark Easy for machine learning algorithms.

DESIRED ISOSURFACING METHOD

Accuracy + mesh quality.

Well-defined gradient differentiation.



M = (V, F)

PREVIOUS METHODS FAIL TO SATISFY BOTH PROPERTIES







Reference

WE INTRODUCE FLEXICUBES



Marching Cubes 15k tris









DMTет 15к tris







Reference 91k tris



Reference

- ✓ High-fidelity
- ✓ Good Quality
- ✓ Convergence →



FlexiCubes

APPLICATIONS W/ FLEXICUBES



3D Reconstruction from Images



Generative 3D Modeling





Animated 3D Reconstruction





Mesh Regularizations

... and more!

Physics Simulation

WHY IS PRIOR WORK NOT SUFFICIENT?



Ground Truth









DEEP MARCHING TETRAHEDRA (DMTET)

SDF vertex position
zero crossing:
$$u_e = \frac{s(x_i)x_j - s(x_j)x_i}{s(x_i) - s(x_j)}$$

ExtractedSurfaceGT Points





MARCHING CUBES



optimization step.

Especially for neural implicits s.t.

$$s = f_{\theta}(x)$$

Increase grid resolution

Not ideal for optimization settings.



XExcessive number of triangles -> bad for downstream applications.

X Requires more SDF samples at every



DUAL CONTOURING $v_d = \operatorname{argmin}_{v_d} \sum \nabla s(u_e) \cdot (v_d - u_e)$ v_d $u_e \in \mathcal{Z}_e$ u_e +surface feature. v_d ╋ + **Dual Contouring**

- \checkmark Vertex placement is more adaptive to
- **X** Grad. issue due to singularity in QEF.



DUAL CONTOURING



$$v_d = \underset{v_d}{\operatorname{argmin}} \sum_{u_e \in \mathcal{Z}_e} \nabla s$$

- ✓ Vertex placement is more adaptive to surface feature.
- X Grad. issue due to singularity in QEF.
- X Produce non-manifold results



 $s(u_e) \cdot (v_d - u_e)$

OUR FORMULATIONS

MAIN IDEA

Improve fitting with more elements?



MAIN IDEA

Improve fitting with more elements? additional flexibility! Recall prior dual methods diverge during optimization.



Marching Cubes

Dual Representation

OUR FORMULATIONS

We introduce 3 types of parameters into Dual Marching Cubes:

- Interpolation weights to position dual vertices in space.
- Splitting weights to control how to split quadrilaterals into triangles.
- Deformation vector for spatial alignment.



Appended as additional output channel in neural implicit.

new parameters

OUR FORMULATION

Interpolation weights:

• α per-cell adjusting interpolation along each edge.



Linear interpolation: $u_{e} = \frac{\alpha_{i}s(x_{i})x_{j} - \alpha_{j}s(x_{j})x_{i}}{\alpha_{i}s(x_{i}) - \alpha_{j}s(x_{j})}$

OUR FORMULATION

Interpolation weights:

- α per-cell adjusting interpolation along each edge.
- β per-cell adjusting vertex position within each dual face





INTERPOLATION WEIGHTS

- α per-cell adjusting **interpolation** along each edge.
- β per-cell adjusting vertex position within each dual face





Movable regions of the dual vertex.

INTERPOLATION WEIGHTS

- α per-cell adjusting interpolation along each edge.
- β per-cell adjusting vertex position within each dual face

 $\alpha, \beta \in \mathbb{R}^+ \longrightarrow \checkmark$ Preserves grad.





 When a cell emits multiple dual vertices, they lies in non-overlapping convex hulls.

EXTRACTING FACES





SPLITTING WEIGHTS

Raw output from Dual Contouring are quadrilateral faces.

Quad-split weights γ controlling how quads get split to tris.



During **optimization**, γ interpolates the surface between two possible splits.



At **inference**, we split along the diagonal with dominant γ

ABLATING PARAMETERS



Baseline

+ interp. weights

+ grid deform

+ split weights

EXTENSIONS

- Tetrahedral mesh extraction.
- Hierarchically adaptive meshing





Uniform 32³ 2.1k tris, CD:4.3





Uniform 32³ 2.1k tris, CD:5.7

Adaptive 64³ 4.5k tris, CD:2.5

Adaptive 128³ GT 7.6k tris, CD: 2.4 10k tris



Adaptive 64³ 5.7k tris, CD:4.7

Adaptive 128³ 22k tris, CD: 4.4

GT 104k tris

VALIDATIONS

VISUAL COMPARISON





FlexiCubes

Reference

FlexiCubes @ 64 resolution

Marching Cubes (MC_SDF)



Baselines

QUANTITATIVE COMPARISONS OF RECONSTRUCTION ERROR

Evaluated on dataset collected by Myles et al. [2014] which contains 79 highly-detailed, diverse 3D shapes.

	64 ³	$CD(10^{-5})\downarrow$	F1 ↑	ECD $(10^{-2})\downarrow$	EF1 ↑
ſ	MC_{SDF}	6.84	0.55	2.55	0.14
Extracting mesh from ground truth SDF	DC _{hermite}	5.90	0.61	3.80	0.23
	NDC _{SDF}	6.16	0.57	1.22	0.26
ſ	MC	6.33	0.66	1.25	0.25
Deconstruct mash via antimization	DMTet(64)	7.50	0.66	3.77	0.28
Reconstruct mesh via optimization	DMTet(80)	5.17	0.66	3.59	0.29
	FIEVICURES	4 87	0.70	0 71	0 4 3
	I LEXICOBES	1.07	0.70	0.71	0.15
	128 ³	$CD(10^{-5})\downarrow$	F1 ↑	ECD $(10^{-2})\downarrow$	EF1 ↑
	$\frac{128^3}{MC_{SDF}}$	$CD(10^{-5}) \downarrow$ 4.72	F1 ↑ 0.68	$ECD (10^{-2}) \downarrow$ 1.13	EF1 ↑ 0.33
	$\frac{128^{3}}{MC_{SDF}}$ $DC_{hermite}$	$CD(10^{-5}) \downarrow$ 4.72 4.59	F1 ↑ 0.68 0.69	$ECD (10^{-2}) \downarrow$ 1.13 3.82	EF1 ↑ 0.33 0.40
	$\frac{128^{3}}{MC_{SDF}}$ $DC_{hermite}$ NDC_{SDF}	$CD(10^{-5}) \downarrow$ 4.72 4.59 5.04	F1 ↑ 0.68 0.69 0.65	$ \begin{array}{c} \text{ECD } (10^{-2}) \downarrow \\ 1.13 \\ 3.82 \\ 0.79 \end{array} $	EF1 ↑ 0.33 0.40 0.43
	$\frac{128^{3}}{MC_{SDF}}$ $\frac{DC_{hermite}}{NDC_{SDF}}$ MC	$ \begin{array}{r} 1.07 \\ \hline CD(10^{-5}) \downarrow \\ 4.72 \\ 4.59 \\ 5.04 \\ 4.51 \\ \end{array} $	F1 ↑ 0.68 0.69 0.65 0.72	$ \begin{array}{c} \text{ECD} (10^{-2}) \downarrow \\ 1.13 \\ 3.82 \\ 0.79 \\ 1.32 \end{array} $	EF1 ↑ 0.33 0.40 0.43 0.44
	$\frac{128^{3}}{MC_{SDF}}$ $\frac{DC_{hermite}}{NDC_{SDF}}$ MC $DMTet(128)$	$ \begin{array}{c} \text{CD}(10^{-5}) \downarrow \\ 4.72 \\ 4.59 \\ 5.04 \\ 4.51 \\ 4.98 \\ \end{array} $	F1 ↑ 0.68 0.69 0.65 0.72 0.74	$ECD (10^{-2}) \downarrow$ 1.13 3.82 0.79 1.32 1.50	EF1 ↑ 0.33 0.40 0.43 0.44 0.39

BETTER TRIANGLE QUALITY

Min angle of extracted triangles.







FlexiCubes

MESH OPTIMIZATION WITH REGULARIZATIONS

We can further improve triangle quality by adding additional regularizers.

Equilateral Edge Length



MC

FlexiCubes

APPLICATIONS

PHOTOGRAMMETRY THROUGH DIFFERENTIABLE RENDERING

Nvdiffrec jointly optimizes shape, materials, and lighting from images.



PHOTOGRAMMETRY THROUGH DIFFERENTIABLE RENDERING

FlexiCubes improves geometric fidelity and mesh quality.



Nvdiffrec w/ DMTet

Nvdiffrec w/ FlexiCubes

and and

Reference

Extracting Triangular 3D Models, Materials, and Lighting From Images Munkberg et. al. CVPR 2022

PHOTOGRAMMETRY THROUGH DIFFERENTIABLE RENDERING

The outputs are compatible with standard graphics workflow.



Multiview inputs



Simulating reconstructed asset with physics in Omniverse

MESH SIMPLIFICATION OF ANIMATED OBJECTS

End-to-end optimization w/ FlexiCubes avoids mesh stretching.





End-to-end optimization

Reference



T-pose optimization

Appearance-Driven Automatic 3D Model Simplification Hasselgren et. al. Eurographics Symposium on Rendering. 2021

DIFFERENTIABLE PHYSICS SIMULATION WITH TET MESH



(f) Existing Methods (State Supervision)

(g) gradSim (Image Supervision)

gradSim: Differentiable simulation for system identification and visuomotor control Jatavallabhula and Macklin et. al. ICLR 2021

DIFFERENTIABLE PHYSICS SIMULATION WITH TET MESH



Initialization

Optimized Results w/ FlexiCubes

Reference

gradSim: Differentiable simulation for system identification and visuomotor control Jatavallabhula and Macklin et. al. ICLR 2021

3D GENERATIVE MODELING FOR MESHES W/ GET3D



GET3D: A Generative Model of High Quality 3D Textured Shapes Learned from Images Gao, Shen, Wang, Chen, Yin, Li, Litany, Gojcic, Fidler, NeurIPS 2022

3D GENERATIVE MODELING FOR MESHES W/ GET3D

GET3D w/ FlexiCubes generates meshes with better details and tessellation.



Motorbike

Chair





Car

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3D GENERATIVE MODELING FOR MESHES W/ GET3D

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Car

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ADDING MESH REGULARIZATIONS

FlexiCubes supports optimizing regularizations defined on meshes.



Marching Cubes

FlexiCubes

Developability of triangle meshes Stein et. al. SIGGRAPH 2018

PERFORMANCE

Performance of Isosurfacing operations

128 ³	Forward Time (ms)	Backward Time (ms)	Memory (MB)
MC	5.08	0.58	72.85
DMTet	6.94	1.39	168.27
<i>DMC_{centroid}</i> [Nielson 2004]	7.34	1.74	150.75
FlexiCubes	14.06	9.53	816.17

Performance of various application (1 iteration)								
Applications (96^3)	NVDIFFRECMC		GET3D					
Isosurface	DMTet	FlexiCubes	DMTet	FlexiCubes				
Time per iter. (ms)	307	315	510	610				
Memory(GiB)	13.1	15.3	11.6	11.1				

LIMITATIONS AND FUTURE WORK

Limitations:

- Self-intersections
- Weaker guarantees of topological correctness in adaptive and tetrahedral meshing. Future work:
- Integrate volumetric rendering with mesh-based representation.
- Extend to 4D spatiotemporal meshing.
- Integrate adaptive hierarchical meshing into generative modeling pipelines.

THANK YOU FOR LISTENING!

Takeaways:

- FlexiCubes is designed for gradient-based mesh optimization.
- Incorporate additional DoFs into mesh extraction.
- Drop-in replacement for better mesh quality and geometric fidelity!



Populate your 3D world with assets generated w/ FlexiCubes!

Visit our project page to explore additional results and learn more details!



