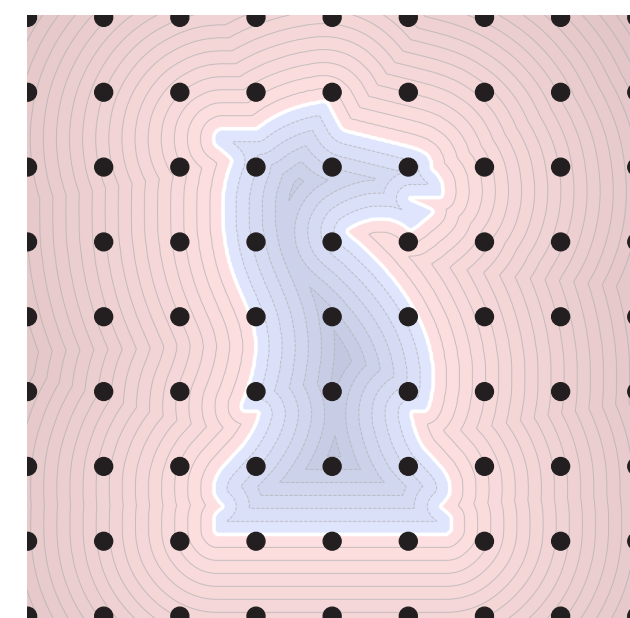
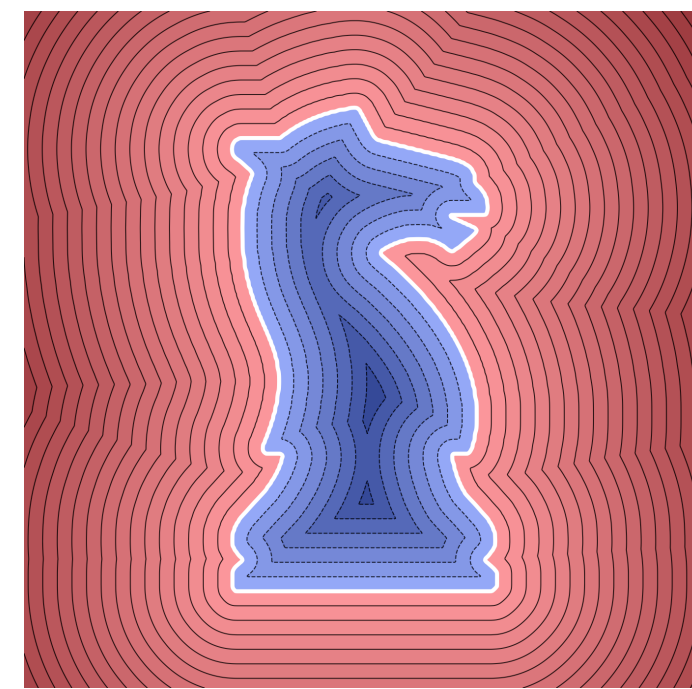


# DUAL CONTOURING OF SIGNED DISTANCE DATA

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## PROBLEM

A Signed Distance Function (SDF) stores, at each point in space, the distance to a surface (negative inside of it and positive outside).



Surface reconstruction



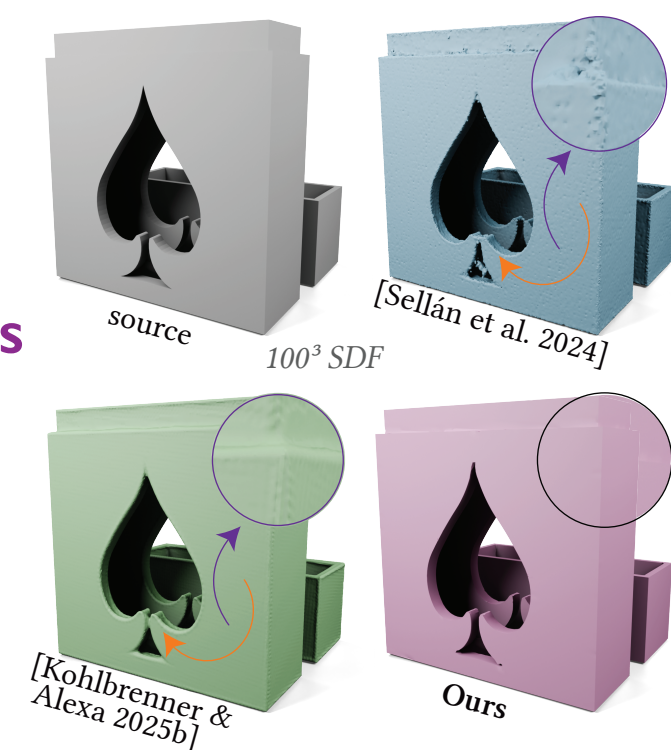
In some applications, we only have access to samples on a grid, not the full function...

... which complicates recovering the true surface represented by the SDF.

“Marching” (cubes/tetrahedra) methods cannot recover sharp features, unless these are aligned with their volumetric discretization.



Point cloud methods smooth edges and merge gaps, unlike ours.

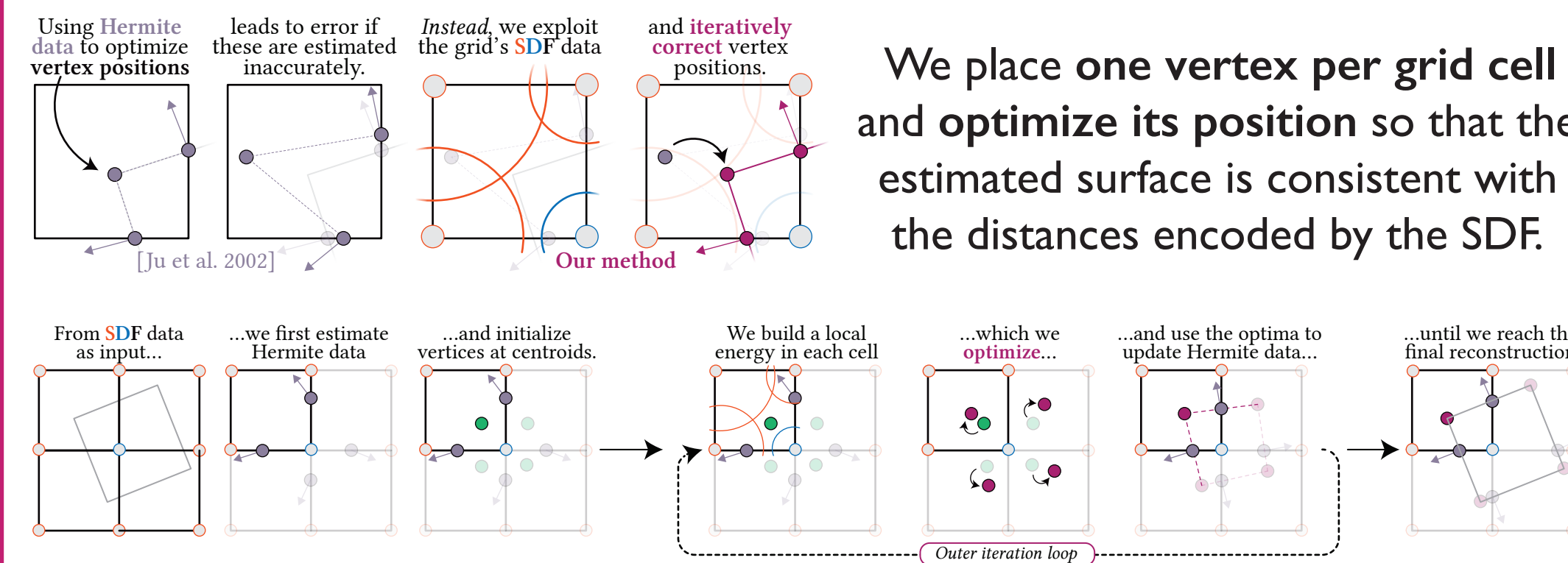


And Dual Contouring relies on gradient information, which we do not require.



## METHOD

Goal: reconstruct a surface from grid-sampled distances (no gradients, no learning, no additional queries).



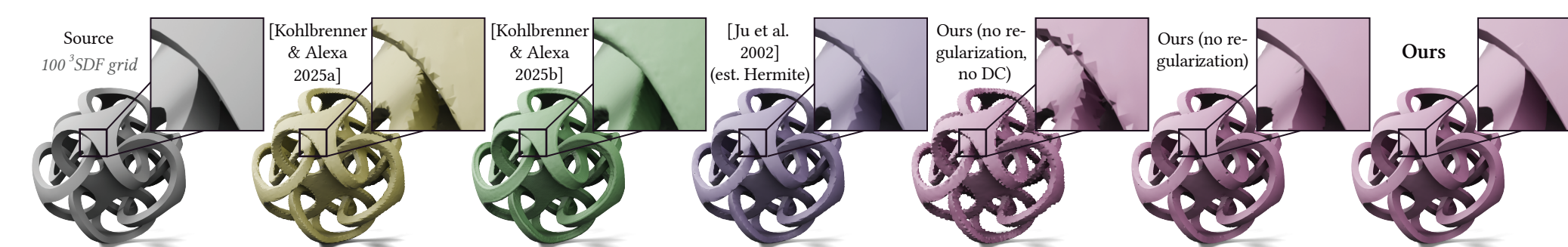
We place one vertex per grid cell and optimize its position so that the estimated surface is consistent with the distances encoded by the SDF.

We alternate between locally optimizing vertex positions and globally updating geometric estimates (Hermite data: intersections and normals).

$$\mathbf{x}^{k,r+1} = \arg \min_{\mathbf{x}} \tilde{E}_d^{k,r}(\mathbf{x}) + w_H E_H^k(\mathbf{x}) + \mu \|\mathbf{x} - \mathbf{x}^{k,r}\|^2$$

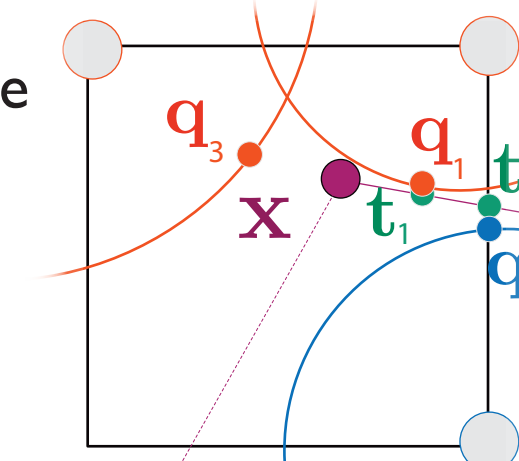
The local optimization minimizes a quadratic energy combining a linearized distance term, the original Dual Contouring energy, and a regularization term.

$$E_H^{k,i}(\mathbf{x}) = \sum_{e_j \in c_i} ((\mathbf{x} - \mathbf{h}_j^k) \cdot \mathbf{n}_j^k)^2$$

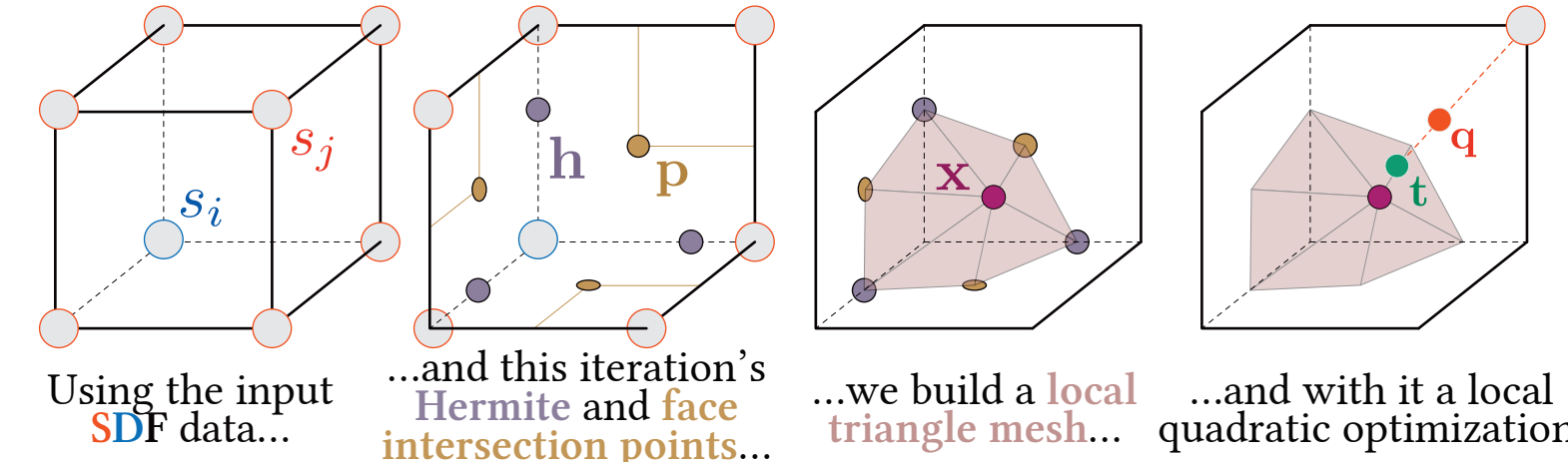


An SDF sample at a given point can be interpreted as a sphere that must be tangent to the true surface.

The linearized distance term measures how far the mesh is from satisfying the sphere constraints.

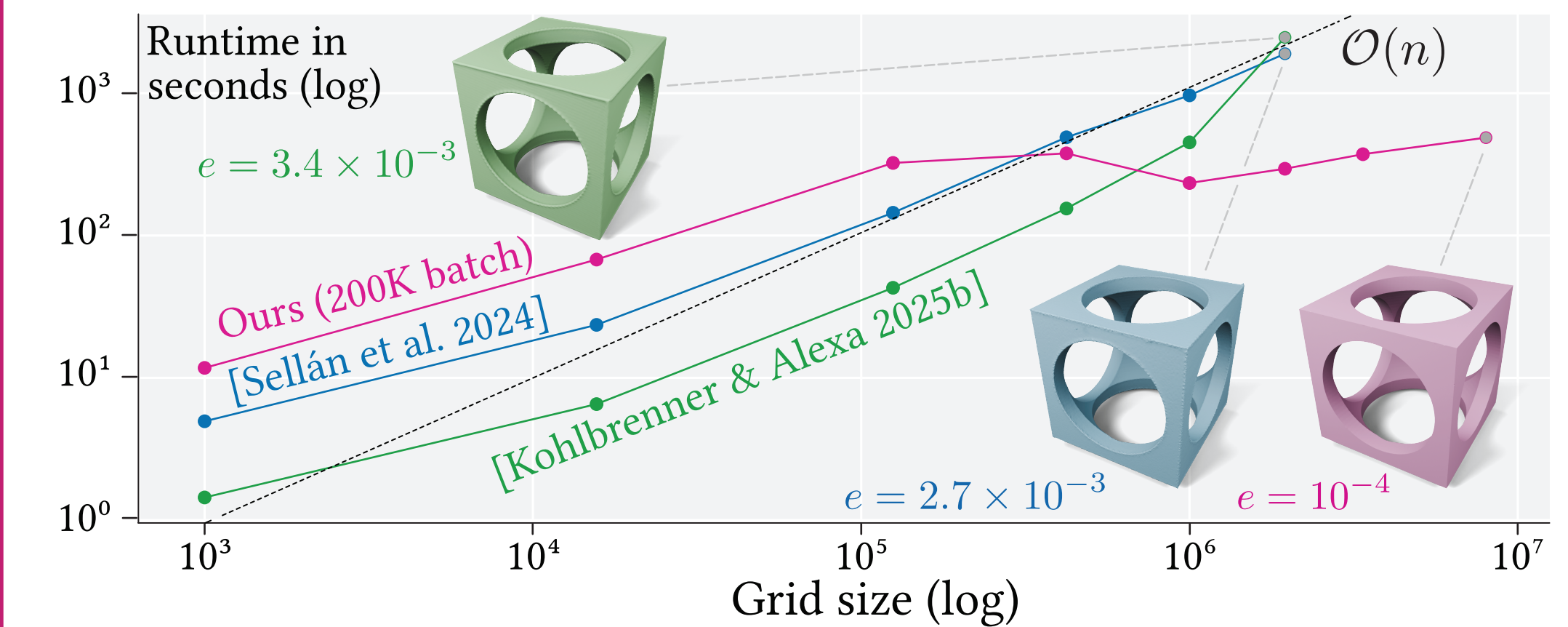


$$(|s_j| - d(\mathbf{u}_j, \mathcal{M}_i))^2 \approx \|t_j - \mathbf{q}_j\|^2 \quad \xrightarrow{\text{Removing tangential motion penalization}} \quad (|s_j| - d(\mathbf{u}_j, \mathcal{M}_i))^2 \approx ((t_j - \mathbf{q}_j) \cdot \mathbf{d}_j)^2$$



## RESULTS

We obtain more faithful reconstructions with a better performance than existing methods.



Our method demonstrates improved accuracy in applications such as motion geometry reconstruction...



... and is successful even when restricting the amount of information available to a narrow band around the input.

