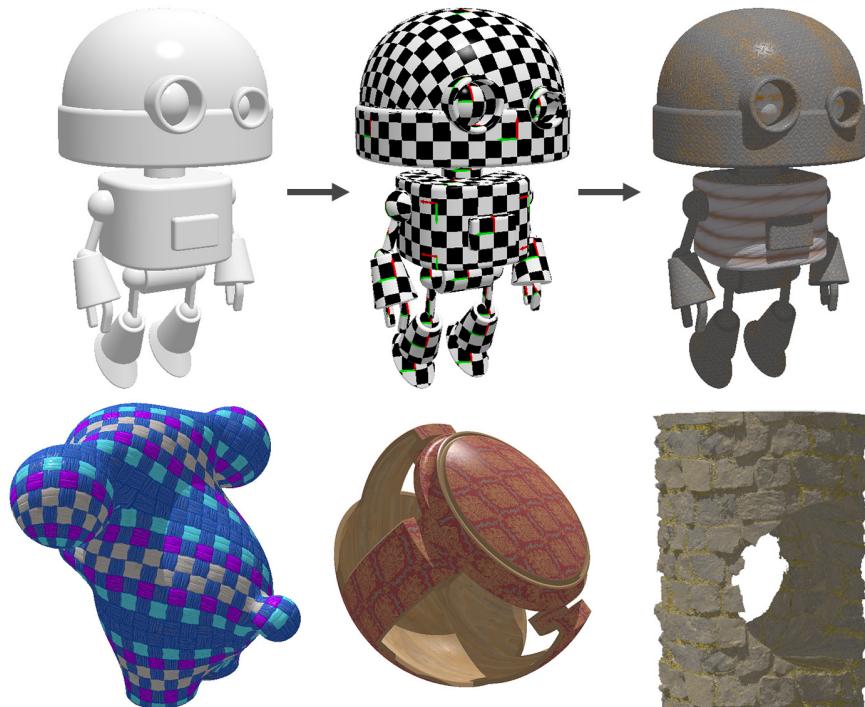


# Implicit UVs: Real-time semi-global parameterization of implicit surfaces

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**Figure 1:** Given an implicit surface (top left), we generate parameterizations with multiple user-placed seeds (top middle, seeds marked with red and green frames) and use them to apply rendering maps (top right). We can merge adjacent fields interactively, hence extending local uv-patches into semi-global parameterizations on smooth regions (bottom left). Our method implicitly segments the surface by sharp features and follows the geometry of the shapes, enabling the texturing of complex objects (bottom middle). Finally, since we never mesh the surface, we can still use the modeling advantages of implicit shapes, for example, to perform topological changes in real time (bottom right, a vertical cylinder textured using our method then carved by a horizontal cylinder).

## Abstract

Implicit representations of shapes are broadly used in computer graphics since they offer many valuable properties in design, modeling, and animation. However, their implicit and volumetric nature makes applying 2D textures fundamentally challenging. We propose a method to compute point-wise and parallelizable semi-global parameterizations of implicit surfaces for texturing, rendering, and modeling purposes. Our method not only defines local patches of parameterization, but also enables the merging of multiple adjacent patches into large and spatially coherent ones that conform to the geometry. Implemented in shaders into a sphere-tracing pipeline, our method allows users to edit the uv-fields with real-time visualization. We demonstrate how to add rendering details (texture, normal, displacement, etc.) using our parameterization, as well extending modeling tools with implicit shell maps. Furthermore, the textured objects remain implicit and can still be used in a modeling pipeline.

## 1 Introduction

Implicit surfaces are one of the most ubiquitous ways to represent shapes on a computer. Contrary to *explicit* representations, like a triangular mesh, which describes the  $\mathbb{R}^3$ -embedding of a collection of connected triangles, *implicit* representation only encodes a subset  $M \subset \mathbb{R}^3$ , defined as the set of zeros of a scalar field  $f : \mathbb{R}^3 \rightarrow \mathbb{R}$ ,

$$M = \left\{ x \in \mathbb{R}^3 \mid f(x) = 0 \right\}. \quad (1)$$

Note that while it is possible that  $M$  contains volumetric regions (3D), curves (1D), or isolated points (0D), if we consider slight perturbations to  $f$ 's value,  $M$  is usually a surface (2D). This scalar field  $f$  gives us a way to *query* the location of  $M$ . If, additionally,  $f$  satisfies the *Eikonal equation*,

$$\|\nabla f\| \equiv 1, \quad (2)$$

then  $f$  becomes a *Signed Distance Field (SDF)*, meaning that there exists a volumetric region  $\Omega$  where  $\partial\Omega = f^{-1}(0)$  and

$$f(x) = \begin{cases} -d(x, \partial\Omega), & \text{if } x \in \Omega, \\ d(x, \partial\Omega), & \text{else,} \end{cases} \quad (3)$$

When a surface  $M$  is encoded with an SDF  $f$ , an important technique, *sphere tracing* [Har96], computes the intersection of a ray and the surface by iterating

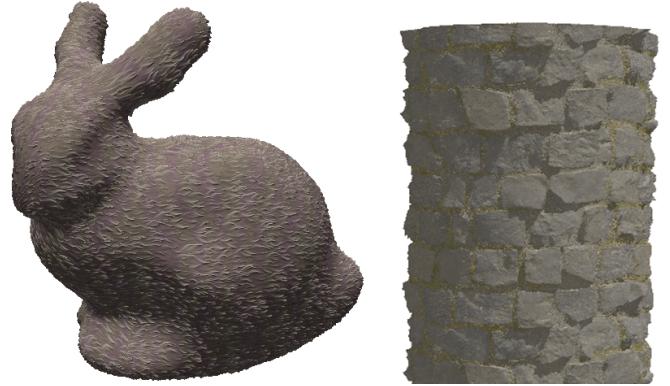
$$x_{k+1} = x_k + f(x_k)\ell, \quad (4)$$

where  $\ell$  is the direction of the ray and  $f(x_k) = d(x_k, M)$  is the largest “safe” (not intersecting with  $M$ ) step size. If  $f$  can be evaluated in parallel, a GPU implementation in shader language of Eq. 4 [JQ14] can significantly speed up ray-tracing to achieve real-time rendering. This technique became the default rendering method for implicit surfaces because it enables 3D geometry editing in a fully implicit setting with real-time visualization.

We can model a scalar field  $f$  on a computer by storing either its values on a voxelized grid or its analytic expression. For example, we can store the values  $f(x_{ijk}) = \|x_{ijk} - q\| - r$  of an SDF of a sphere  $B_q(r)$  on the voxels  $x_{ijk}$  of a 3D grid, or we can store the sphere center  $q \in \mathbb{R}^3$  and its radius  $r > 0$ . These two methods each have their trade-offs. While voxelized functions are more expressive and can handle freestyle edits, analytic functions enjoy a minimal memory footprint and infinite resolution. In this paper, though our method is agnostic to the type of functions (SDFs or non-SDFs) and the storage method (voxel or analytical expression), we focus on real-time geometry modeling and use analytic SDFs in most of our examples.

Many well-known design tools build on implicit representations. For example, in *constructive solid geometry (CSG)*, one models the intersection of two shapes  $A$  and  $B$  by computing  $f_{A \cap B} = \max(f_A, f_B)$ , where  $f_\bullet$  denotes the SDF of  $\bullet$ . In *inverse rendering*, researchers often use implicit representation to model arbitrary geometry [VSJ22] because changing the surface topology can be made differentiable by changing  $f$ 's value.

Implicit representation enjoys many appealing features, but what are some of its drawbacks when compared to explicit representations? A mesh, for example, *explicitly* encodes a surface, an



**Figure 2:** We can use our method to render implicit shapes with color texture, normal map, and displacement in a sphere-tracing pipeline. In this figure, the bunny (left) is defined by a neural pseudo-SDF, while the cylinder (right) is an analytical SDF.

object intrinsically 2D but living in 3D. One can exploit this 2-dimensional nature to compute a 2D parameterization, for example using CETM [SSP08] or boundary-first flattening [SC17], so there exists a global mapping between the surface  $M$  and the uv-plane  $\mathbb{R}^2$ . This parameterization can be used to apply additional rendering properties

$$x \in M \mapsto (u, v) \in \mathbb{R}^2 \mapsto \mathcal{A}(u, v), \quad (5)$$

where  $\mathcal{A} : \mathbb{R}^2 \rightarrow Q$  could be color ( $Q = [0, 255]^3$ ), normal ( $Q = \mathbb{S}^2$ ), or displacement height ( $Q = \mathbb{R}$ ). Since implicit surfaces are merely a subset in  $\mathbb{R}^3$ , one can only query the surface in a volumetric manner. Without first converting the implicit surface into a mesh (via *marching cube* [LC98] or *dual contour* [JLSW02]), one cannot easily parameterize the surface to add rendering details. This meshing step is computationally expensive and can introduce unwanted smoothed features. Furthermore, it depletes the advantage of using analytic SDFs for real-time modeling if the users have to generate the mesh to add details. We thus devise a fast and point-wise method to generate semi-global uv-coordinates on implicit surfaces for texturing purposes while avoiding meshing. By moving along the surface  $M$  using a second-order approximation of the shape, we build local parameterization and provide a way to join multiple adjacent parameterized patches together into a common coordinate system. Built into a sphere-tracing renderer, our method runs in real-time on analytic SDFs and allows users to edit the uv-fields with instantaneous feedback. Finally, since we never mesh the surface, the final textured object remains implicit and can be used in a modeling pipeline, for example to perform topological operations on the shape in real-time, to deform it, or to exploit symmetries.

We summarize our contributions into the following list:

- A fully point-wise and parallelizable local parameterization built from an approximation of the geodesic path between the query point and a reference point.
- Each parameterization patch is defined on a smooth region delimited by sharp features.
- A smooth blending of multiple local parameterization patches

into a *common coordinate system* for parametrizing a larger smooth portion of the surface. Our interpolation scheme is continuous and only blends near the interface between seeds.

- This parameterization system can be used to
  - Apply textures of color, normal, and displacement maps in a sphere-tracing rendering pipeline (see Fig. 2).
  - Encode a complex (possibly explicit) shape as a displacement on a simple implicit surface.
  - Morph some other implicit geometry in the three-dimensional shell around the surface using shell maps [PBFJ05].
- The evaluation of this system runs in parallel per pixel and can be implemented on GPU.

## 2 Previous works

### Purely implicit approaches

We first review techniques that directly aim at computing coordinates or adding details on implicit surfaces without discretizing the surface.

Many such techniques make heavy assumptions on the structure of the function  $f$ . For example, analytical fields, which are structured as the skeleton of simple shapes are used to compute coordinates and blending textures [TW99], add fine geometric details on the surface from a noise defined on the surface [ZBL\*12], or from displacement maps on shapes with canonical coordinates, such as ellipsoids [SP91]. Another structured representation is proposed by Arquès et al. [AMP00] where an implicit surface is defined by a large number of blended discoid primitives. Since this representation is very dense in terms of sampling, they can encode complex geometry and store coordinates at each seed, enabling the use of textures.

Another approach is to define a flow from the surface to an en-globing volume, such as a sphere, on which coordinates are known [ZGVdF98] [TW99]. However, such a mapping process induces a lot of distortion and is valid mostly for simple, blob-like shapes.

### Decal style that works on arbitrary geometry but only locally

On the other hand, some approaches work without assuming *any* structure on the surface, which can then be applied on *any* shape representation, including implicit surfaces. Indeed, in the work of De Groot et al. [DGWB\*14], the authors propose a way to define a local coordinate system around some points in order to apply decals on generic surfaces. They do so by only using the Euclidean distance and an orientation frame carried by each point. This technique has been extended in several works, such as by using curvature information to improve distance estimation [RASS16], or by enhancing the interaction between the decals [NMBS21]. However, despite its efficiency and simplicity, the coordinate computation is too simple to be applied on topologically complex shapes at a large scale, hence being mostly restricted to decals.

### Can be applied to implicit surfaces but need discretization

To bypass the absence of an explicit surface, some works first discretize the shape, using contouring [LC98] [JLSW02] to compute

coordinates by applying some variations of the shortest path algorithm. In the work of Pederson et al. [Ped95], the authors subdivide the shape into parameterized patches, whose borders are geodesics. To compute the geodesic between two points, they first compute the shortest path on the contoured shape, which is then refined by a curve shortening flow restrained on the implicit surface (which is a common way to compute such curves [YZ10]). Finally, the authors match the derivatives of the coordinates between the patches in order to tile the surface. A variation of Djikstra's algorithm is also proposed in the work of Schmidt et al [SGW06], which also propagates exponential coordinates from a seed. Such a system provides an intuitive and controllable way to compute robust coordinates on implicit surfaces. However, since the shape is discretized, the output is not an implicit surface anymore and one loses the benefits of such a representation.

### Dense representations

When the implicit surface is stored in a voxel grid, the geometry editing is simplified, and users can add fine details [EB17] [DHM12]. For example, by defining a coordinate system via a shell of particles, the work of Brodersen et al. [BMPB08] enables the use of shell maps [PBFJ05] to carve or add geometric details by morphing some other geometry on the surface of the shape. By fully embracing such volumetric representation, the work of Tarini [Tar16] bypasses the need of a surface by also computing and storing uv coordinates in voxels. In a similar way, the field of differentiable rendering may also encode shapes in such a volumetric representation [VSJ22]. However, although voxel-based approaches offer very high accuracy, the cost of such accuracy in terms of grid resolution often precludes real-time processing and control.

## 3 A local parameterization around a reference point

The goal of this paper is to extend what is possible when rendering implicit surfaces using sphere tracing. In such a pipeline, the GPU computes, in parallel, for each pixel of the screen the intersection between its ray and the surface. For each ray-surface intersection  $x \in M$ , we would like to query additional rendering attributes  $\mathcal{A}(u, v)$  (color, specularity, etc., see Sec. 5.2). We design a point-wise method of deriving uv-coordinates that is implementable in a GPU renderer.

In this section, we first create a local parameterization around a single reference point  $p \in M$ . Recall that a local parameterization is a mapping from a neighborhood  $U$  of  $p$  to  $\mathbb{R}^2$ :

$$\begin{aligned}\varphi_p : U \subset M &\rightarrow \mathbb{R}^2 \\ x &\mapsto (u, v).\end{aligned}\tag{6}$$

We use an approximation of the exponential map (Sec. 3.1) together with an ambient heuristic (Sec. 3.3) to approximate the geodesic path between the query point  $x$  and the reference point  $p$ . We explain how geodesics embody local parameterization in Sec. 3.4. In Sec. 3.6, we lay out our algorithm and provide additional implementation details.

### 3.1 Local approximation of the exponential map.

#### 3.1.1 Geodesics

Geodesics are the generalization of straight lines to a (curved) manifold. A curve  $\gamma : [0, T] \rightarrow M$  is a geodesic if it is a critical point of the following energy

$$E(\gamma) = \int_0^T \|\dot{\gamma}(t)\|^2 dt. \quad (7)$$

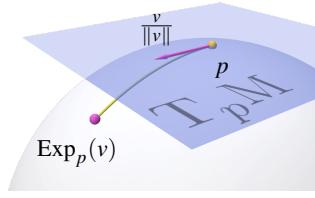
where  $\|\cdot\|$  is the standard Euclidean norm in  $\mathbb{R}^3$ . Such a curve is parameterized by arc length, i.e.,  $\|\dot{\gamma}\| \equiv 1$ . We can understand geodesics as the locally shortest paths on a surface. A geodesic path beginning with point  $p \in M$  with an initial direction  $v \in T_p M$  is the critical point of  $E(\gamma)$  with the following initial values:

$$\begin{cases} \gamma_p^v(0) = p \\ \dot{\gamma}_p^v(0) = \frac{v}{\|v\|}. \end{cases} \quad (8)$$

The *exponential map*  $\text{Exp}_p : T_p M \rightarrow M$  maps any vector  $v$  to the solution of Eq. 8 at  $t = \|v\|$ :

$$\text{Exp}_p(v) = \gamma_p^v(\|v\|). \quad (9)$$

(See inset figure.) From the arc length parameterization, the *geodesic* distance between  $p$  and  $\text{Exp}_p(v)$  is exactly  $d_M(p, \text{Exp}_p(v)) = \|v\|$ .



#### 3.1.2 First order approximation

An implicit surface  $M = f^{-1}(0)$  is uniquely defined by the function  $f$ ; therefore, all its geometric properties can be inferred from  $f$  and its derivatives. For example, the gradient of  $f$  gives rise to the unit normal of  $M$ :

$$n(p) = \frac{\nabla f(p)}{\|\nabla f(p)\|}. \quad (10)$$

The tangent plane  $T_p M$ , seen here as a subset of  $\mathbb{R}^3$ , is the orthogonal complement of  $n(p)$ ,

$$T_p M = n(p)^\perp = \nabla f(p)^\perp. \quad (11)$$

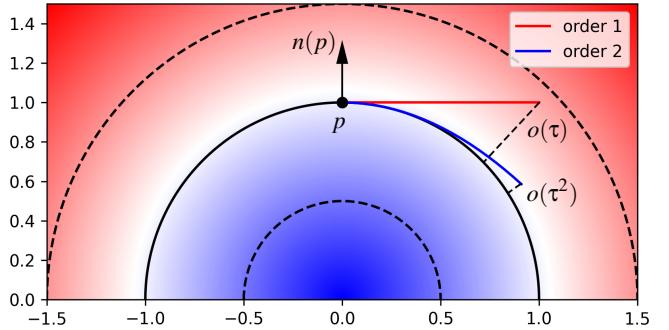
Given an initial direction  $v \in T_p M$ , we can locally approximate the exponential map  $\text{Exp}_p(v)$  à la Forward Euler method:

$$\gamma_p^v(\tau) = p + \tau v. \quad (12)$$

However, one can check with a simple Taylor expansion of  $f$  that Eq. 12 only stays on  $M$  up to  $o(\tau)$ :

$$\begin{aligned} f(p + \tau v) &= f(p) + \langle v, \nabla f(p) \rangle \tau + o(\tau) \\ &= o(\tau), \end{aligned}$$

since  $f(p) = 0$  and  $v \perp \nabla f(p)$ . This first order approximation requires very small time steps to keep the curve close to the surface. Since evaluating  $f$  is the most costly part of the pipeline, this can significantly slow down the process. Next, we proceed to describe a second-order approximation in order to bypass this limitation.



**Figure 3:** Comparison of first vs second order schemes for  $\text{Exp}_p$ . A second-order approximation of the level set allows one to take larger, hence fewer, steps. The walk remains significantly closer to the surface than with the first-order method. The dotted-lines illustrate the size of the error and the location of the projection.

#### 3.1.3 Second order approximation

A second-order approximation of the exponential map on an implicit surface  $M$  is a path  $\gamma_p^v$  such that

$$f(\gamma_p^v(\tau)) = o(\tau^2). \quad (13)$$

To build such  $\gamma_p^v$ , we must use second-order geometric quantities of  $M$ . Let  $n : M \rightarrow \mathbb{S}^2$  denote the Gauss map as given in Eq. 10. The *shape operator* at  $p$  is defined as the differential of  $n$ ,

$$\begin{aligned} S_p : T_p M &\rightarrow T_p M \\ v &\mapsto D_v n(p) = \left. \frac{\partial n}{\partial v} \right|_p. \end{aligned} \quad (14)$$

The eigenvalues  $\kappa_1, \kappa_2$  and eigenvectors  $v_1, v_2$  of  $S_p$  are the principal curvatures and principal curvature directions of  $M$  at  $p$ . For implicit surfaces, we can compute the shape operator  $S_p$  from  $f$  with

$$S_p = \left( I - n(p)n(p)^T \right) \frac{H_f(p)}{\|\nabla f(p)\|}, \quad (15)$$

where  $H_f(p)$  denotes the Hessian of  $f$  at  $p$ . The shape operator gives a local quadratic approximation of the surface  $M$ ,

$$\begin{aligned} \gamma_p^v(\tau) &= p + \tau v + \frac{\tau^2}{2} \langle v, S_p(v) \rangle n(p) \\ &= p + \tau v - \frac{\tau^2}{2\|\nabla f(p)\|} \langle v, H_f(p)v \rangle n(p). \end{aligned} \quad (16)$$

One can use the Taylor expansion of  $f$  at  $p$ , composed with the above, to verify that Eq. 13 is satisfied.

Evaluating the full Hessian in Eq. 16 would be pretty expensive (e.g. using finite difference,  $f$  needs to be evaluated 18 times!). Luckily, we only need the term  $\langle v, H_f v \rangle = (D_v)^2 f$ . The three-point finite difference scheme,

$$\langle v, H_f v \rangle = (D_v)^2 f = \frac{f(p - hv) - 2f(p) + f(p + hv)}{h^2} + o(h^2), \quad (17)$$

only requires 2 additional evaluations of  $f$  since we always need

$f(p)$ . For highly curved surfaces or those  $C^1$  but not  $C^2$  (differentiable but not twice differentiable), one can re-project the approximation of  $\text{Exp}_p(v)$  onto  $M$  using:

$$\Pi_f(x) = x - f(x) \frac{\nabla f(x)}{\|\nabla f(x)\|^2}. \quad (18)$$

As shown in Fig. 3, the second order approximation permits larger steps and, therefore, fewer evaluations are needed to move a fixed distance  $T$ , even on irregular surface. In the next section, we will see how one can define coordinates from the Exponential map in order to parameterize the neighborhood of a point, which will be the cornerstone of our method.

### 3.2 The log map as a local parameterization

As described previously, the exponential map  $\text{Exp}_p$  maps the tangent plane  $T_p M$  to the surface  $M$ . If we select an orthonormal basis  $e_1, e_2 \in T_p M$ , we can then map  $\mathbb{R}^2$  to  $M$  by

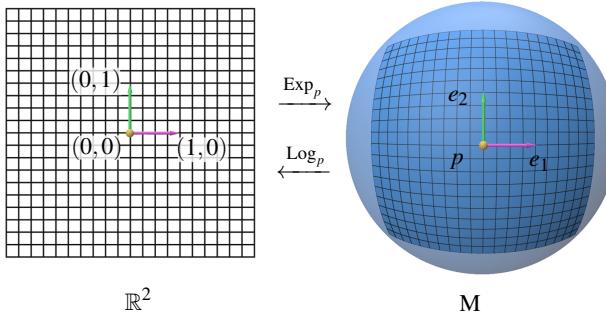
$$\begin{aligned} \mathbb{R}^2 &\rightarrow M \\ (u, v) &\mapsto \text{Exp}_p(ue_1 + ve_2). \end{aligned} \quad (19)$$

In other words, given a 2D coordinate  $(u, v)$ , we can use Eq. 19 to deduce the 3D point location  $\text{Exp}_p(ue_1 + ve_2)$  on  $M$ . This process describes the inverse of our goal Eq. 6. We therefore build the inverse of the exponential map, the *logarithmic (log) map*. It is defined locally on a neighborhood  $U \subset M$  near  $p$ :

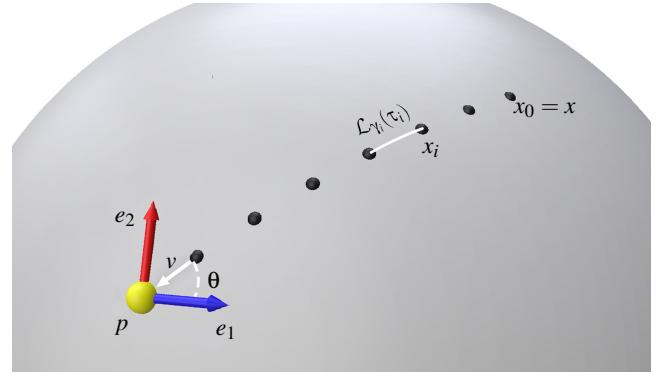
$$\text{Log}_p : U \rightarrow T_p M, \quad (20)$$

such that  $\text{Exp}_p(\text{Log}_p(x)) = x$ . As illustrated on Fig. 4, using the same parameterization of  $T_p M \rightarrow \mathbb{R}^2$  for the fixed orthonormal frame  $e_1, e_2 \in T_p M$ , the Log map effectively maps  $U$  to  $\mathbb{R}^2$ , which is our goal. Several previous works also used the Log map for parameterization [SSC19] [Sch13] [HA19].

However, as is often the case, calculating the inverse map is much harder than calculating the forward one. Indeed, the Log map takes a query point  $x \in M$  and finds the velocity  $v = \dot{\gamma}(0) \in T_p M$  of the minimizing geodesic  $\gamma$  from  $p = \gamma(0)$  to  $x = \gamma(T)$ , which is equivalent to computing a shortest path  $\gamma$  between  $p$  and  $x$  [MS01].



**Figure 4:** With a selected orthonormal frame  $e_1, e_2 \in T_p M$ , the exponential map  $\text{Exp}_p$  maps a flat plane to the surface, while the log map  $\text{Log}_p$  provides each nearby point  $x$  with 2D coordinates.



**Figure 5:** Following a gradient descent of the Euclidean distance while adhering to the surface yields an approximated geodesic between  $x$  and  $p$ . We calculate the coordinates from the arc length of the curve and its incoming angle at endpoint  $p$  with a fixed frame  $e_1, e_2$ .

In the following subsection, we explain how we use a heuristic to approximate the solution to this problem without a discretizing the shape.

### 3.3 Approximating the log map with ambient heuristic

Large-scale approaches to computing geodesics commonly use global methods to explore large portions of the geometry all at once [CLPQ20]. However, we want to avoid meshing the surface and to be able to run the queries in parallel, so we resort to locally approximating the geodesic for each query point  $x$ .

We follow a heuristic to guide the path. We approximate the geodesic from  $x$  to  $p$  with a gradient flow of the Euclidean ambient distance  $d(y, z) = \|y - z\|$  while adhering to the surface  $M$ :

$$\begin{cases} \dot{\gamma}(t) = \Pi_{T_{\gamma(t)} M} (\nabla d(\gamma(t), p)) \\ \gamma(0) = x, \end{cases} \quad (21)$$

The gradient of Euclidean distance  $\nabla d$  has a closed form,

$$\nabla d(\gamma(t), p) = \frac{\gamma(t) - p}{\|\gamma(t) - p\|}. \quad (22)$$

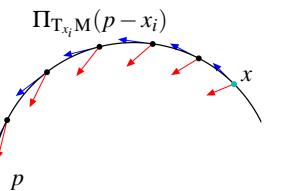
So does the projection operator  $\Pi_{T_{\gamma(t)} M}$  on the tangent plane  $T_{\gamma(t)} M$ ,

$$\Pi_{T_{\gamma(t)} M} = I_3 - n(\gamma(t))n(\gamma(t))^T. \quad (23)$$

Furthermore, we can use the exponential map defined in Eq. 16 to keep the discretized sequence on the surface:

$$x_{i+1} = \text{Exp}_{x_i}(-\tau_i \nabla_{x_i}^M d(x_i, p)). \quad (24)$$

Here  $\tau_i$  denotes the step size at step  $i$ . Our approach doesn't fall into the IVP-BVP (initial value problem vs boundary value problem) dichotomy presented in [CLPQ20] since we use a local



geodesic tracing technique to approximate the geodesic between two points.

There is no guarantee that our approximation (Eq. 24) converges to an actual geodesic between  $x$  and  $p$ . It is also possible that our calculated path from  $x$  to  $y$  differs from that from  $y$  to  $x$ . However, we only use this admittedly flawed approximation of geodesics only to build a coordinate system. From here on, we sometimes omit the adjective “approximated” and refer to these curves as “geodesics.”

### 3.4 From geodesics to coordinates

The previous subsection lets us approximate the geodesic curve  $\gamma$  from  $x$  to  $p$ . Although we cannot convert the curve  $\gamma$  directly to uv-coordinates in  $T_p M$ , we draw inspiration from the conversion of polar to Cartesian coordinates:

$$(u, v) = r(\cos(\theta), \sin(\theta)). \quad (25)$$

Indeed, here we can estimate the length  $r = \mathcal{L}(\gamma)$  and an angle  $\theta$  between the direction of  $\gamma$  at its endpoint  $p$  and the fixed frame  $e_1, e_2 \in T_p M$ . See Fig. 5 for an illustration.

We compute  $\mathcal{L}(\gamma)$  by measuring each of the  $N$  curves  $\gamma_i$ , joining  $x_i = \gamma_i(0)$  and  $x_{i+1} = \gamma_i(\tau_i)$ , from Eq. 24:

$$\mathcal{L}(\gamma) \approx \sum_{i=0}^{N-1} \mathcal{L}(\gamma_i). \quad (26)$$

We then approximate each  $\mathcal{L}(\gamma_i)$  using Simpson’s rule:

$$\mathcal{L}(\gamma) = \int_0^\tau \|\dot{\gamma}(s)\| ds \approx \frac{\tau}{6} \left( \|\dot{\gamma}(0)\| + \|\dot{\gamma}\left(\frac{\tau}{2}\right)\| + \|\dot{\gamma}(\tau)\| \right). \quad (27)$$

To estimate the incoming angle  $\theta$  at  $p$ , we first calculate its unit tangent direction at the end point  $p = \gamma(T)$ :

$$v = \frac{\dot{\gamma}(T)}{\|\dot{\gamma}(T)\|}. \quad (28)$$

Since the frame  $e_1, e_2 \in T_p M$  is orthonormal, we have the following relation:

$$(\cos(\theta), \sin(\theta)) = (\langle -v, e_1 \rangle, \langle -v, e_2 \rangle). \quad (29)$$

Finally, we compute the log map in  $(u, v)$  coordinates using:

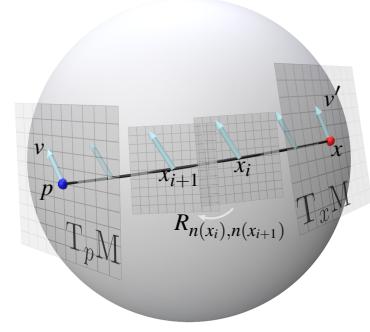
$$\text{Log}_p^{\text{uv}}(x) = \mathcal{L}(\gamma)(\langle -v, e_1 \rangle, \langle -v, e_2 \rangle). \quad (30)$$

The curve-based parameterization has several advantages compared to those using direct estimators for implicit decals [DGWB\*14] [RASS16]. Since we walk on the surface, we obtain a much more meaningful estimation of the geodesic distances, as displayed in Fig. 8, and we can accumulate additional geometric information along the curve as explained in the following subsection.

### 3.5 Geometric information accumulated along the curve

#### 3.5.1 Parallel transport

*Parallel transport* refers to the way a tangent vector evolves when transported along a curve  $\gamma$  from one point of the surface to another.



**Figure 6:** We can discretize parallel transport along the curve built by Alg. 1. We store the product of each rotation matrix  $R_{n(x_i), n(x_{i+1})}$  into one rotation matrix  $R_{p \rightarrow x}$ . A vector  $v \in T_p M$  is then transported to  $v' \in T_x M$  by  $v' = R_{p \rightarrow x}v$ .

Here, we can build the mapping from the tangent plane of the seed  $T_p M$  to the one of the query points  $T_x M$ .

On a classical triangular mesh [CDS10], the smallest rotation matrix between two neighboring faces  $F_i$  and  $F_j$  gives a coherent discretization of the parallel transport between  $T_{F_i} M$  and  $T_{F_j} M$ . This rotation is the same as the one that rotates the face normals  $n_i$  to  $n_j$  along the dihedral angle, given by the following formula:

$$R_{n_i, n_j} = I_3 + J + \frac{1}{1 + \langle n_i, n_j \rangle} J^2, \quad (31)$$

where  $J$  is the skew-symmetric matrix such that  $Ju = (n_i \times n_j) \times u$  for all  $u \in \mathbb{R}^3$ . We take inspiration from this formula and build the parallel transport matrix from reference point  $p$  to query point  $x$  by accumulating the rotation at each iteration between  $x_i$  and  $x_{i+1}$ :

$$R_{T_p M \rightarrow T_x M} = \prod_{i=1}^{n-1} R_{n(x_{i+1}), n(x_i)}. \quad (32)$$

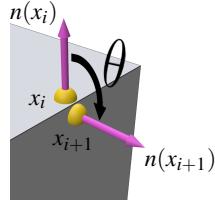
From here on, use the notation  $R_{p \rightarrow x} = R_{T_p M \rightarrow T_x M}$  interchangeably for simplicity. We illustrate the accumulation process in Fig. 6. This will be useful to apply normal maps in Sec. 5.2.1 and to compute optimal frames in Eq. 45 when merging multiple patches in Sec. 4.4. Together with the estimated geodesic distances, we can diffuse tangent vector fields on the surface as displayed in Sec. D of the supplemental.

#### 3.5.2 Detecting sharp features

Until this point, we have built our theory on the assumption that the function  $f$  is sufficiently smooth. However, models found in the wild often contain sharp features where the normal is discontinuous, e.g. the 90-degree edges of a box. Our geodesic tracer would be unstable near these sharp edges and corners because there’s no well-defined tangent plane to project the distance gradient, hence, the uv-patch cannot cross them in a robust way. We argue that this property does not diminish the usefulness of our method. In the context of 3D design, sharp features often divide the shape into semantically separate regions, each having their own uv island. For example, in Fig. 1, sharp features signal different components made of different materials in the robot.

Since our parameterization is based on geodesics connecting query points  $x$  and a reference point  $p$ , we can detect sharp features by testing for normal discontinuity along the curve:

$$n(x_i) \cdot n(x_{i+1}) < 1 - \epsilon_2. \quad (33)$$



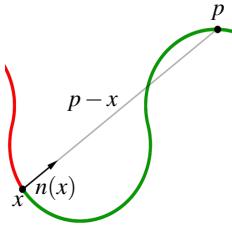
Here  $\epsilon_2 > 0$  is the tolerance to normal changes. This test may give out false positives, i.e., detecting sharp features if the shape smoothly but rapidly curves. One may use different tolerance  $\epsilon_2$  at different regions to accommodate the geometry.

### 3.6 Implementation

We sum up our method for a patch of local parameterization around a single reference point  $p$  in Alg. 1.

#### 3.6.1 Limits of ambient heuristics

The gradient flow of the Euclidean distance does not always produce a valid curve between the query point  $x$  and the reference point  $p$  on the surface. Since we do not perform a global search to find the optimal path, the flow might get stuck in a local minimum. These local minima are the points  $x$  such that  $p - x$  is co-linear with  $n(x)$ , as illustrated in the inset (the region where the heuristic produces a path to  $p$  is highlighted in green). However, since the function  $x \mapsto d(x, p)$  is a continuous function on  $M$  with a global minimizer  $p$ , every point in the attraction basin of  $p$  must be able to follow the gradient flow and reach  $p$ . We describe how to alleviate the locality of such construction in Sec. 4.



#### 3.6.2 Adaptive Time-step

An essential ingredient of the algorithm is to make sure that all walks to a seed take the same number of steps to guarantee GPU optimization and continuity of the uv-field. We detail in Sec. A of the supplemental document why we use:

$$\tau_i = \sqrt{\frac{\sqrt{1 + s_i^2 \alpha^2} - 1}{\frac{\alpha^2}{2}}} \text{ where } s_i := \|x_{i+1} - x_i\| = \frac{\|x_i - p\|}{N - i}. \quad (34)$$

## 4 Semi-global parameterization using multiple patches

Since the previous section describes the method to compute a local parameterization patch given a reference point  $p$  and an orthonormal frame  $e_1, e_2 \in T_p M$ , we call the tuple  $(p, e_1, e_2) \in M \times T_p M^2$  a *seed* of a patch of parameterization. In this section, we investigate using multiple user-specified seeds to reduce distortion and cover more area (Fig. 7).

### 4.1 Finding the geodesic-nearest seed

If there are  $n$  seeds  $\{(p_i, e_{1i}, e_{2i})\}_{i=1}^n$ , a query point  $x$  can move towards any of the  $n$  seeds to obtain its uv-coordinates. Previous

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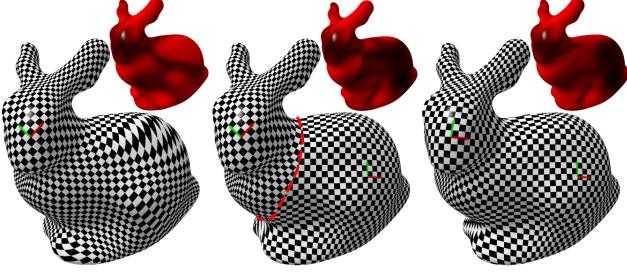
#### Algorithm 1: Log Map around a single reference point

```

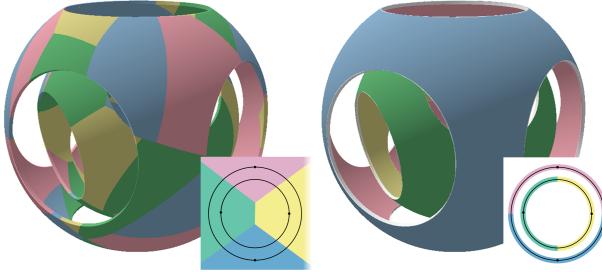
Data: scalar field:  $F$ , query point  $x \in M$ , reference point  $p$ ,
frame  $(e_1, e_2) \in T_p M^2$ , number of iterations  $N$ ,
maximum distance  $D$ , tolerance parameters  $\epsilon_1, \epsilon_2$ 
Result: uv coordinates:  $(u, v) = (r \cos(\theta), r \sin(\theta))$ , parallel
transport rotation matrix  $R_{T_p M \rightarrow T_x M}$ 
if  $d(x, p) > D$  then
| return fail; // exceeds maximal distance
end
 $x_0 = x$ ;
 $r_0 = 0$ ;
 $R_0 = I_3$ ;
// Fixed number of iterations
for  $i \in [0, N - 1]$  do
|  $s_i = \frac{d(x_i, p)}{N - i}$ ;
|  $\tau_i = \text{ComputeTimeStep}(s_i)$ ; // adaptive time
| step (Sec. 3.6.2)
| // Compute direction
|  $v_i = \Pi_{n_{x_i}^\perp}(p - x_i)$ ;
| if  $\|v_i\| < \epsilon_1$  then
| | return fail; // heuristic fails
| | (Sec. 3.6.1)
| end
|  $\gamma_i = \Gamma(F, x_i, \frac{v_i}{\|v_i\|})$ ; // build geodesic
| (Eq. 16)
|  $x_{i+1} = \gamma_i(\tau_i)$ ;
| if  $n_{i+1} \cdot n_i < 1 - \epsilon_2$  then
| | return fail; // sharp feature
| | (Sec. 3.5.2)
| end
| // Accumulate rotation from  $TM_{x_{i+1}}$  to
|  $TM_{x_i}$ 
|  $R_{i+1} = R_i R_{n_{i+1} \rightarrow n_i}$ ;
|  $r_{i+1} = r_i + \mathcal{L}_\gamma(\tau_i)$ ;
| if  $r_{i+1} + d(x_{i+1}, p) > D$  then
| | return fail; // exceeds maximal
| | distance
| end
| end
| if  $d(x_N, p) < r_{min}$  then
| |  $r = r_N + d(x_N, p)$ ;
| | // Use  $x_{N-1} - p$  instead of  $x_{N-1} - x_N$  for
| | numerical robustness
| |  $t = \Pi_{n_p^\perp}(x_{N-1} - p)$ ;
| | return  $(u, v) = (\frac{t}{\|t\|} \cdot re_1, \frac{t}{\|t\|} \cdot re_2)$ ,  $R = R_N R_{n_p \rightarrow n_N}$ 
| end
| return fail; // exceeds maximal steps

```

---



**Figure 7:** Left: using one single seed, Alg. 1 can parameterize a large smooth patch of a surface but may accumulate distortion with distance and curvature [SGW06], which is visible on the bunny's back. Middle: using multiple seeds can reduce distortion but creates discontinuity (red dashed line). Right: We merge the parameterization of adjacent patches to eliminate discontinuity. We visualize distortion using a black-red gradient (red being the most distorted) next to each example.



**Figure 8:** We illustrate in 2D the Euclidean Voronoi cells (top left) and the geodesic Voronoi cells (top right) with the same seed placement on the same geometry. parameterization using the Euclidean nearest seed [DGWB\*14] [RASS16] is only valid near the seeds and can produce erroneous results with too few seeds (bottom left). We use the geodesic nearest seed and the segmentation by sharp feature, which incorporates the geometry of the surface and produces favorable results (bottom right) with the same few seeds used in the bottom left.

works use the seed closest to the query point measured by the Euclidean distance [DGWB\*14] [RASS16]. This strategy can incur significant errors because a small Euclidean distance between  $x$  and  $p_i$  does not guarantee that  $x$  is semantically associated with  $p_i$  or even connected to  $p_i$  on the surface (Fig. 8 left). We replace the Euclidean distance  $d(x, p_i)$  with the geodesic distance  $d_M(x, p_i) = \mathcal{L}(\gamma)$  computed following Alg. 1. The Voronoi cell of a seed  $p_i$ , defined by the set of  $x$  where  $p_i$  is the closest seed to  $x$ , is the region where we compute the uv-coordinates based on  $(p_i, e_{1i}, e_{2i})$ . Using the geodesic closest seed implies partitioning the surface with the *geodesic Voronoi cells*, which are always simply connected. As a result, our strategy produces coherent parameterization on implicit surfaces regardless of geometry, and we can use much fewer seeds compared to previous works.

Since the geodesic distance is always longer than the Euclidean distance

$$d_M \geq d, \quad (35)$$

we can use the Euclidean distance to speed up the subroutine in finding the geodesic-nearest seed. We first find the  $n_{\text{euc}}$  closest seeds w.r.t. the Euclidean distance to the query point  $x$ , sorted by ascending order. Starting with the Euclidean closest seed, we call Alg. 1 to compute the geodesic distance  $d_M(x, p_i)$  between  $x$  and each seed  $p_i$ . According to Eq. 35, if a seed  $p_i$  has a larger Euclidean distance  $d(x, p_i)$  than the current minimal geodesic distance  $d_M(x, p^*)$ ,  $p_i$  would have no chance to be closer to  $x$  than the current best seed  $p^*$ . Therefore, we can terminate the iteration early with very few calls to Alg. 1.

#### 4.2 Merging patches overview

Existing techniques that generate a global uv-field on the entire implicit surface [SP91] [Tig01] can facilitate texture application but suffer from singularities and significant distortion since they are defined generically. On the other hand, techniques using local coordinate systems are much more controllable and have low distortion near each seed [DGWB\*14] [SGW06], but they only permit small and localized texture patches (“decal” style). To benefit from the best of both worlds, we propose a way to merge multiple local coordinate fields into a more global one while maintaining quality and control. Moreover, since even exact log maps [SGW06] accumulate distortion with curvature, our method can be viewed as extending the valid region of the log map.

To generate a more global parameterization, a natural idea is to interpolate the uv-fields from different seeds:

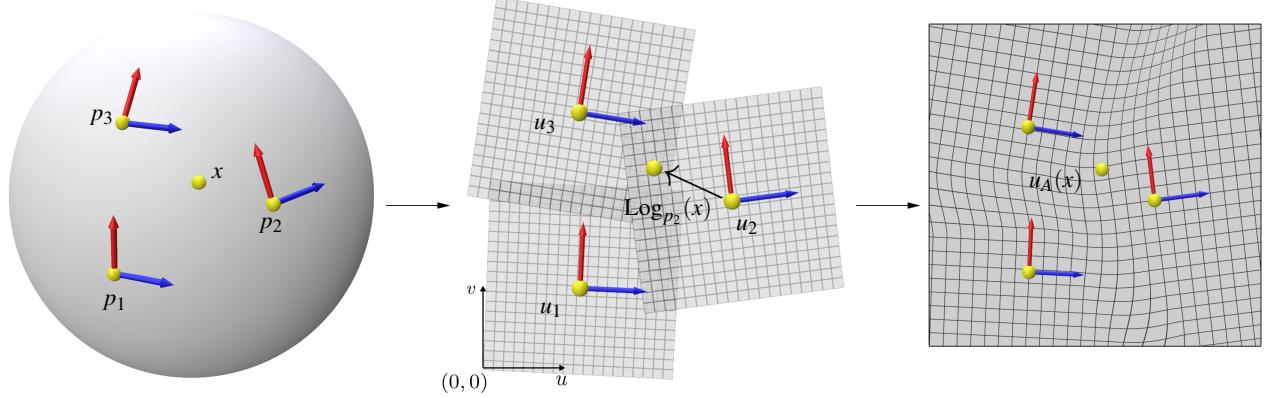
$$u(x) = \frac{\sum_{i=1}^n w_i(x) \text{Log}_{p_i}^{\text{uv}}(x)}{\sum_{i=1}^n w_i(x)}. \quad (36)$$

In particular, one can use the Voronoi diagram of the seeds to guide the interpolation. To keep our evaluation fully point-wise, real-time, and implicit by nature, we cannot afford to compute interpolation weights as in natural neighbors techniques [BU06]. Therefore, we look to kernel-based techniques, i.e. using the distance to each seed to calculate the interpolation weight. However, interpolating uv-fields is more involved than interpolating fixed data, such as colors (Fig. 10). Since each parameterization  $\text{Log}_{p_i}^{\text{uv}}$  is centered around  $(u, v) = (0, 0)$  at its seed  $p_i$ , simply taking a weighted sum of different uv-fields will result in incoherent texture. In particular, the uv-coordinates for every seed are all the same,  $(0, 0)$ , even though they are scattered at different places on the surface. To express each uv-field involved in the interpolation in a common coordinate system, we assign each seed  $p_i$  with an offset location  $u_i$  in a mutually agreeable manner (Fig. 9 middle). We modify Eq. 36 with uv-offsets:

$$u(x) = \frac{\sum_{i=1}^n w_i(x) (\text{Log}_{p_i}^{\text{uv}}(x) + u_i)}{\sum_{i=1}^n w_i(x)}. \quad (37)$$

We discuss the blending weights  $w_i$  in Sec. 4.3 and the uv-offsets  $u_i$  in Sec. 4.4. We illustrate the complete pipeline in Fig. 9.

In practice, we build a graph  $(V, E)$  where the vertices are the seeds location  $V = \{p_i\}_{i=1}^n$  and an edge is included  $E_{ij} \in E$  if the



**Figure 9:** To combine multiple uv-fields, we wish to take weighted averages of coordinates at a point  $x$ , coming from adjacent seeds (left). Since each Log map  $\text{Log}_{p_i}$  is expressed in its own referential, we first assign each seed  $p_i$  an offset coordinate  $u_i \in \mathbb{R}^2$  in a common uv-space (middle). In this common coordinate system, we can blend the coordinates at the interfaces between the seeds using compact support interpolation (right).

user selects  $p_i$  and  $p_j$  for merging. Each edge is weighted by the geodesic distance  $d_M(p_i, p_j)$ . We also attach the uv-offset  $u_i$  to each seed  $(p_i, e_{1i}, e_{2i}, u_i)_i$  into tuples,

$$(p_i, e_{1i}, e_{2i}, u_i) \in M \times (\mathbf{T}_p M)^2 \times \mathbb{R}^2, i = 1, \dots, n. \quad (38)$$

We package these tuples together with the weighted graph into an *Atlas* data structure  $A = (\{(p_i, e_{1i}, e_{2i}, u_i)\}_i, E)$  to be sent to GPU for rendering the textured surface. See Alg. 2 for the complete implementation for computing the blended uv.

Finally, the location of the seams of the final parameterization is very easy to predict since it lies at the Voronoi frontier between two seeds,  $i$  and  $j$ , that are not merged together, i.e. if  $(i, j) \notin E$ . Note that one could use this system on the colors between unmerged fields to make the seams less visible.

### 4.3 Compactly supported weights

One could use the Gaussian kernel based on the geodesic distance between points. However, though it provides controllable and smooth interpolation, it has poor locality, i.e., the weights remain nonzero even if the points are far away. Since we want to avoid frequent calls to Alg. 1, we prefer to blend only at the Voronoi frontier (boundary of Voronoi cells) between two seeds. We first describe a kernel that only blends in the narrow band of the Voronoi frontier (Sec. 4.3.1). We then describe how to blend amongst more than two seeds (Sec. 4.3.2). Finally, we share a simple test using the Euclidean distance to check if a given seed  $p_i$  is needed to compute the final parameterization of the query point  $x$  (Sec. 4.3.2).

**4.3.1 Approximation of the distance to the geodesic Voronoi frontier**

We use the following formula to approximate the distance to the geodesic Voronoi frontier  $V_{ij}$  between  $p_i$  and  $p_j$ :

$$d_{V_{ij}}(x) := d_M(x, V_{ij}) \approx \frac{d_M(x, p_i)^2 - d_M(x, p_j)^2}{2d_M(p_j, p_i)}. \quad (39)$$

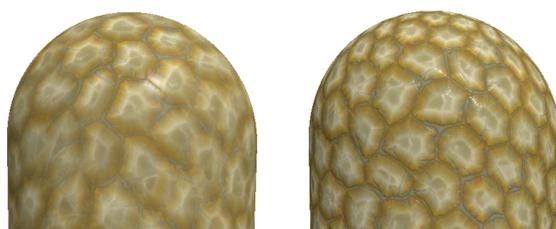
The formula only relies on geodesic distances to each seed and the geodesic distance between seeds. We provide a derivation in Sec. B of the supplemental.

We use this approximated distance as a weight function to blend the uv-fields from these two seeds. Given a user-specified blending width  $2\sigma > 0$ , we define the weight using a smooth blending function  $\omega$  (see Fig. 11):

$$w_{i,j}(x) = \begin{cases} 1, & d_{V_{ij}}(x) < -\sigma \\ 0, & \sigma < d_{V_{ij}}(x) \\ \omega\left(\frac{1}{2} - \frac{d_{V_{ij}}(x)}{2\sigma}\right), & -\sigma \leq d_{V_{ij}}(x) \leq \sigma. \end{cases} \quad (40)$$

To be more precise, we use  $\omega(t) = 3t^2 - 2t^3$ , but any smooth function with  $\omega(0) = 0, \omega(1) = 1$  could work. In practice, we set  $\sigma = \frac{1}{3} \min_{e_{ij} \in E} d_M(p_i, p_j)$  for all seed pairs. Intuitively,  $w_{i,j}$  is the smoothed indicator function of the Voronoi half-space  $H_{i,j} = \{x \in M \mid d_M(x, p_i) < d_M(x, p_j)\}$ .

One can check that if  $d_M(x, p_i) = d_M(x, p_j)$  then  $d_{V_{ij}}(x) = 0$ , and if  $d_M(x, p_i) < d_M(x, p_j)$  then  $d_{V_{ij}}(x) < 0$  and vice-versa. Hence,  $\{|d_{V_{ij}}(x)| \leq \sigma\}$  always contains the approximated geodesic



**Figure 10:** Techniques like triplanar mapping directly blend the color, resulting in blurred texture and incoherence (left). We extend several uv-fields into a common uv-coordinate system, enabling structured textures on a large scale (right).

**Algorithm 2:** Compute blended uv from an atlas

---

**Data:** Atlas  $A = (\{(p_i, e_{1i}, e_{2i}, u_i\}_i, E)$ , query point  $x \in M$ , bandwidth  $\sigma > 0$ ; parameters for Alg. 1: number of steps  $N$ , maximal distance  $D > 0$ , tolerance  $\epsilon_1, \epsilon_2$

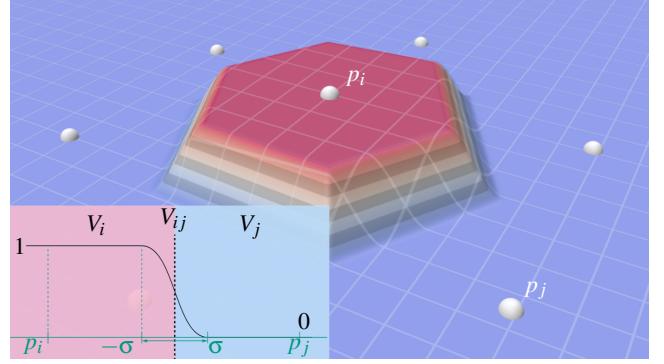
**Result:** Blended coordinates:  $u(x) \in \mathbb{R}^2$

```

 $\{p_{i_1}, \dots, p_{i_n}\} = \text{SortSeed}(x, A); \quad // \text{sorted in ascending } \|x - p_{i_k}\|$ 
 $L_1 = \text{LogMap}(x, p_{i_1}, N, D, \epsilon_1, \epsilon_2); \quad // \text{Alg. 1}$ 
 $L^* = L_1;$ 
 $i^* = i_1;$ 
 $d_M^* = \|L_1\|; \quad // +\infty \text{ if Alg. 1 failed}$ 
// Find the geodesic nearest seed  $p_{i^*}$ 
for  $k \in [2, n]$  do
  if  $\|x - p_{i_k}\| \leq d_M^*$  then
     $L_k = \text{LogMap}(x, p_{i_k}, N, d_M^*, \epsilon_1, \epsilon_2);$ 
    if  $\|L_k\| \leq d_M^*$  then
       $i^* = i_k; \quad // \text{current best index}$ 
       $d_M^* = \|L_k\|; \quad // \text{current geodesic distance}$ 
       $L^* = L_k; \quad // \text{current uv}$ 
    end
  end
end
if  $|Neighbors(E, i^*)| = 0$  then
  // No blending required
  return  $L^* + u_{i^*}; \quad // \text{offseted uv-coordinates}$ 
end
else
  // Compactly supported blending
   $L = (0, 0);$ 
   $w_{i^*} = 1;$ 
   $W = 0;$ 
  for  $j \in Neighbors(E, i^*)$  do
     $d_{j,i^*} = d_M(p_j, p_{i^*}); \quad // \text{edge weight of } e_{j,i^*} \in E$ 
    if  $(d_M^*)^2 + 2\sigma d_{j,i^*} \geq \|x - p_j\|^2$  then
      // passes test from Sec. 4.3.3
       $L_j = \text{LogMap}(x, p_j, N, d_M^* + 2\sigma);$ 
       $w_{i^*,j} = w(d_M^*, \|L_j\|, d_{j,i^*}); \quad // \text{blending weight by Eq. 40}$ 
       $w_{i^*} = \min(w_{i^*}, w_{i^*,j}); \quad // \text{Eq. 41}$ 
       $w_{j,i^*} = 1 - w_{i^*,j}; \quad // \text{Eq. 42}$ 
      // Numerator of Eq. 37
       $L = L + (w_{j,i^*})(L_j + u_j);$ 
      // Denominator of Eq. 37
       $W = W + w_{j,i^*};$ 
    end
  end
  return  $\frac{L + w_{i^*} L^*}{W + w_{i^*}};$ 
end

```

---



**Figure 11:** Blending weight  $w_{i,j}$  between two seeds  $p_i, p_j$  (top) and blending weight  $w_i = \min_{j, e_{ij} \in E} w_{i,j}$  in relation to all surrounding seeds (bottom). The final weight  $w_i$  is continuous, compactly supported, and only takes values other than 0 or 1 on a narrow band near the Voronoi frontiers between seeds.

Voronoi frontier  $\{d_M(x, p_i) = d_M(x, p_j)\}$  and has non zero width, which is all we need to guarantee that the parameterizations are blended continuously between seeds.

#### 4.3.2 Combining weights from multiple seeds

If a query point  $x$  falls in the blending area of more than two seeds, we need a weight  $w_i(x)$  for each of these relevant seeds  $p_i$ . Since the Voronoi cell  $V_i$  is the intersection of all the Voronoi half-spaces  $V_{i,j}$ , we take inspiration from  $\min(1_A, 1_B) = 1_{A \cap B}$ , where  $1_A$  is the indicator function of  $A$ , and propose the following weight function:

$$w_i(x) = \min_{j : e_{ij} \in E} w_{i,j}(x), \quad (41)$$

We plot the graph of this function in Fig. 11 (bottom). This formula is continuous and only depends on distances to nearby seeds selected for blending. Furthermore, to implement things efficiently in shaders, if  $p_i$  is the closest seed to  $x$ , we skip Eq. 41 for other seeds  $p_j$  and use the follow approximation:

$$w_j(x) \approx w_{j,i}(x) = 1 - w_{i,j}(x). \quad (42)$$

This assumption makes the evaluation linear in the number of seeds to blend (no need to compute  $w_{j,i,j}$ ) and prevents caching. This assumption might not hold at the intersection of multiple Voronoi frontiers, but we found the discontinuity barely visible in all of our examples. The final blending then follows Eq. 37 using these weights.

#### 4.3.3 A cheap test to avoid unnecessary computations

Suppose  $p_i$  is the closest seed to the query point  $x$ . If  $w_{i,j}(x) = 1$  for a neighbor seed  $p_j$ ,  $e_{ij} \in E$ , we would not need to compute  $\text{Log}_{p_j}^{uv}(x)$  since it will not influence the final uv at  $x$ . This happens when  $d_{V_{ij}}(x) < -\sigma$ , i.e.  $x$  is sufficiently far away from  $p_j$ . We can use the Euclidean distance and Eq. 39 to make a cheap test:

$$\begin{aligned} w_{i,j}(x) = 1 &\iff \frac{d_M(x, p_i)^2 - d_M(x, p_j)^2}{2d_M(p_i, p_j)} < -\sigma \\ &\iff d_M(x, p_i)^2 + 2\sigma d_M(p_i, p_j) < d_M(x, p_j)^2. \end{aligned} \quad (43)$$

Since  $\|x - p_j\|^2 \leq d_M(x, p_j)^2$ , we can test for the sufficient condition

$$d_M(x, p_i)^2 + 2\sigma d_M(p_i, p_j) < \|x - p_j\|^2, \quad (44)$$

which saves us a call to Alg. 1 to compute  $d_M(x, p_j)$  when this condition is met.

#### 4.4 Optimal frames and uv-offsets

Even on a flat plane, two seeds with inconsistent frames  $(e_1, e_2) \neq (e'_1, e'_2)$  can cause the merged parameterization to have severe distortion. We therefore optimize the frame placement and uv-offsets together to minimize distortion. We let the user specify the position of the seeds  $\{p_i\}_{i=1}^n$  together with the blending graph  $E$  that indicates what seeds should be blended.

The *distortion* of a parameterization can be understood as a measurement of how it differs from an isometry (a locally rigid map). In other words, we will measure how the parametrized surface behaves differently from the uv-plane. We start by observing that, on a flat plane, parallel transport is always the identity map  $R_{x \rightarrow y} = I_3$ . Since we restrict our attention to ortho-normal frames, given  $e_1$ , we can uniquely determine  $e_2$  by a 90° rotation on the tangent plane, using the formula  $e_2 = n \times e_1$ . We introduce the optimization variables  $t_i \in \mathbb{R}^3$  for the frame vector  $e_1$  at each seed. To penalize distortion, we select the optimal frames by minimizing a simple Dirichlet energy on the blending graph  $A$ , weighted by the inverse squared geodesic distance to favor interactions with closest neighbors:

$$E_{\text{frame}}(\{t_i\}_{i=1}^n) = \sum_{e_{ij} \in E} \frac{\|R_{i \rightarrow j} t_i - t_j\|^2}{d_M(p_i, p_j)^2}. \quad (45)$$

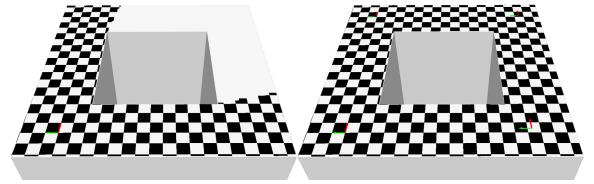
Here we use  $R_{i \rightarrow j} = R_{p_i \rightarrow p_j} = R_{T_{p_i} M \rightarrow T_{p_j} M}$  to simplify the notation. Since this energy is invariant by global rotations, we fix one frame  $t_1$  for one seed  $p_1$ .

If  $M$  is a flat plane, the global minimizer of the energy would be constant field  $t_i = t, i = 1, \dots, n$  and  $E_{\text{frame}} = 0$ . Once the optimal  $\{t_i\}_{i=1}^n$  are computed, we set  $e_{1i}$  by projecting  $t_i$  to  $T_{p_i} M$  and normalizing it. When  $M$  is curved, the optimal frame field will change as little as possible between two neighboring seeds. This energy is similar to the one used in the work of Knöppel et al. [KCP13], except that we apply it to a very small graph  $A$  and we fix one frame instead of finding the smallest energy eigenvector.

Once we computed the frames, we then use Alg. 1 to compute the uv-coordinates in reference to all neighboring seeds that require blending  $\{\text{Log}_{p_i}^{\text{uv}}(p_j)\}_{e_{ij} \in E}$ . Again, if  $M$  is flat, the uv-offsets  $u_i$  are trivial:

$$\text{Log}_{p_j}(p_i) = u_i - u_j. \quad (46)$$

We follow the same spirit and select the uv-offsets that minimize



**Figure 12:** The parameterization merging also solves some issues of the Euclidean heuristic: even if each local uv-field cannot cross the hole in the center (left), merging parameterizations from four seeds allows us to parameterize around it (right).

deviation from the above equation:

$$E_{\text{offset}}(\{u_i\}_{i=1}^n) = \sum_{e_{ij} \in E} \frac{\|u_i - u_j - \text{Log}_{p_j}(p_i)\|^2}{d_M(p_i, p_j)^2}. \quad (47)$$

Similarly, we fix the uv-offset  $u_0 = (0, 0)$  for one seed  $p_0$  to remove redundant degrees of freedom.

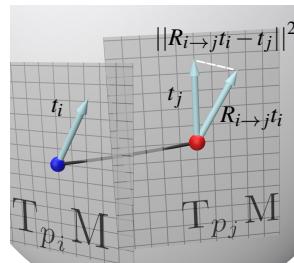
Both energies  $E_{\text{frame}}$  and  $E_{\text{offset}}$  are quadratic and can be minimized by solving a linear system. The corresponding matrices are standard graph Laplacians, except that the frame energy matrix is defined by block where the parallel transport matrices are stored in the cross terms:

$$L_{ij}^{\nabla} = -\frac{R_{i \rightarrow j}}{d_M(p_i, p_j)^2}, \quad (48)$$

in a fashion similar to the discrete connection Laplacians [KCP13].

It is important to note that the main advantage of our method resides in the fact that the linear systems are of size proportional to the number of seeds in the blending graph, a small number in practice. In our experience, we never needed more than 10 seeds per patch, whereas linear systems used to parameterize meshes are often applied on meshes with several thousands of vertices. We need to update the frames and uv-offsets if the user makes any change to the implicit surface, moves a seed, or changes the blending graph, but the systems' size makes them instantaneously solvable. Hence, the editing of the merged uv-field is also real-time!

Finally, not all configuration of seeds can be merged since we may face topological obstructions (e.g. trying to map a plane to a cylinder). We detail how to detect them in Sec. C of the supplemental.



## 5 Applications

### 5.1 The parameterized surfaces remain implicit

Why remain in implicit (point-wise) settings when one could just convert to meshes and parameterize them [SGW06] [Ped95]? We recount the advantages of remaining in a fully implicit setting First, the instantaneous computation per pixel per frame allows users to modify *any* parameter of our method or of the underlying model and get immediate feedback. This includes parameters of Alg. 1, seed placement, merging graph, etc... Artists can have the highest level of control during the design process. Another advantage is



**Figure 13:** Applications: (row A) When rendering an implicit shape deformed by a warping operator  $\omega$ , one can compute the uv-coordinates on the rest pose (left) and transfer the texture to the deformed pose (right). A highly deformed mesh would show rough edges where the resolution is not high enough, while implicit surfaces can fully preserve their geometry. (row B) Conversion of a textured mesh into an implicit shape as a displacement of a sphere (middle). One can then trivially apply standard implicit modeling operators, here a time-dependent twist and a subtraction by a cylinder (right). (row C) Using implicit shell-maps, a base geometry (cylinder) can be tiled with another one (bunny) by using a modular operator on the computed uv (left). A tangent vector field (here, the gradient of a scalar field) is visualized by placing 3D arrows and using a modular operator on the uv for the duplication (right).

that the output remains implicit, so one could use our method at any node of a CSG construction tree to build independent textured objects before combining them. One can also perform topological changes on textured shapes, which is difficult on meshes (Fig. 13 right). Finally, while a high-resolution mesh is required for deformation, our method can generate parameterization on implicit surfaces warped by  $\omega : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  [SJNJ19] with no additional cost. Indeed, one can first texture a shape  $M = \{f = 0\}$  in a rest pose, then, when rendering the deformed shape  $\omega(M) = \{f \circ \omega^{-1} = 0\}$  with sphere-tracing, when an intersection  $x$  is found, i.e.,  $f(\omega^{-1}(x)) = 0$ , Alg. Eq. 2 can be called on the function  $f$  directly but queried at the point  $\omega^{-1}(x)$ . In this way, as illustrated on Fig. 13, the walk Eq. 21 is performed on the rest pose while the uv appropriately follows the shape from  $M$  to  $\omega(M)$ . In a similar fashion, if an object is symmetric, for example w.r.t. the  $yz$ -plane, one can symmetrize the texturing by querying the uv at  $(|x|, y, z)$ , hence dividing the number of required seeds by 2.

## 5.2 Rendering details on implicit surfaces

In this section, we review several ways to use Implicit UV to add details on implicit surfaces. Applying color texture is straightforward. Given a texture picture  $\mathcal{A} : \mathbb{R}^2 \rightarrow [0, 255]^3$ , the color at query point  $x \in M$  is computed by

$$x \mapsto u(x) \mapsto \mathcal{A}(u(x)). \quad (49)$$

To apply normal and displacement maps, we utilize the parallel transport from Sec. 3.5.1.

### 5.2.1 Normal maps

Recall that normal maps  $N : \mathbb{R}^2 \mapsto \mathbb{S}^2$  are textures that contain three channels  $(n_1, n_2, n_3)$  consisting of the 3D coordinates of the material's normal vector at each point of the texture. They capture sub-resolution details that may change how light reflects off the surface. Normal maps were previously only available to meshes and other

explicit representations due to their requirement for parameterizations. We aim to extend them to implicit surfaces rendered with sphere-tracing. Since normals are vectors, they are stored in coordinates  $(n_1, n_2, n_3)$  with a fixed basis  $\{t, b, n\}$ . At a given location  $x$  on a surface with the local tangent vector  $t(x)$ , bi-tangent vector  $b(x)$ , and normal vector  $n(x)$ , the normal map augments the rendering normal into

$$\tilde{n}(x) = n_1 t(x) + n_2 b(x) + n_3 n(x). \quad (50)$$

On meshes, the standard procedure uses the uv-coordinates on a triangle to estimate the basis  $t(x)$  and  $b(x)$  and uses the triangle normal for  $n(x)$ . The frame  $\{t(x), b(x), n(x)\}$  is stored in a TBN matrix and

$$\tilde{n}(x) = \text{TBN}(x)N(u, v). \quad (51)$$

On implicit surfaces, the tangent frame  $t_i$  and normal  $n_i$  at seed  $p_i$  are a part of the Atlas data structure. Therefore, we transport this frame from the nearest seed  $p_i$  to the query point  $x$  to obtain the rendering normal:

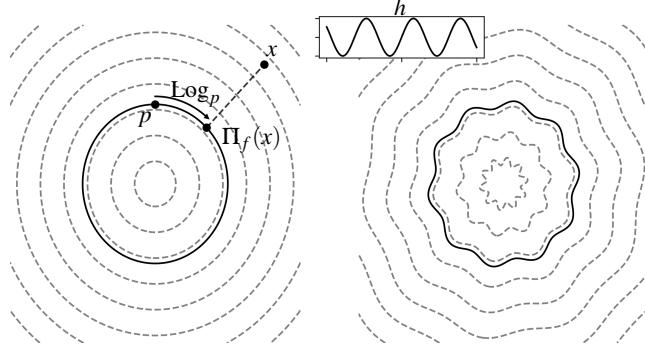
$$\tilde{n}(x) = R_{p_i \rightarrow x} \text{TBN}_i N(u(x)). \quad (52)$$

### 5.2.2 Displacement maps

When  $f$  is an SDF, we can also use our method to add high frequency details by modifying the distance field query. Unlike most common methods for implicit surfaces, our method does not involve warping and is suited for structured displacement maps.

The idea is very simple: we want to displace the surface  $M$ , in the normal direction  $n(x)$  by an amount  $h(x)$ . The problem is loading the height map  $h : \mathbb{R}^2 \rightarrow \mathbb{R}$  from a 2D image on query points on a curved geometry  $x \in M \subset \mathbb{R}^3$ .

To illustrate the process, observe that the graph of a function  $h : \mathbb{R}^2 \rightarrow \mathbb{R}$ ,  $M = \{(x, y, z) \mid z = h(x, y)\}$  can be seen as the 0-level



**Figure 14:** To apply a displacement in the normal direction of a signed distance field (left), we first project the query point  $x$  onto the surface  $\Pi_f(x)$ , then compute the uv (here with one seed  $p$ ) then subtract the height at said uv (right).

set of a function  $f_h : \mathbb{R}^3 \rightarrow \mathbb{R}$  by

$$f_h(x, y, z) = z - h(x, y). \quad (53)$$

The above can also be seen as displacing the  $(x, y)$  plane  $P_{xy} = \{(x, y, z) \mid z = 0\}$  by  $h$  in its normal direction  $e_z$ . The (undisplaced) signed distance function to  $P_{xy}$  is  $d(x, y, z) = z$  to  $P_{x,y}$ , the parameterization of  $P_{xy}$  is

$$\begin{aligned} u : P_{xy} &\rightarrow \mathbb{R}^2 \\ (x, y, 0) &\mapsto (x, y). \end{aligned} \quad (54)$$

Considering the projection to  $P_{xy}$ ,  $(x, y) = \Pi(x, y, z)$ , we can rewrite Eq. 53 as:

$$f_h(x, y, z) = d(x, y, z) - (h \circ u \circ \Pi)(x, y, z). \quad (55)$$

We can directly generalize Eq. 55 to curved surfaces defined by SDFs. Given an SDF  $f : \mathbb{R}^3 \rightarrow \mathbb{R}$ , our parameterization system  $u(x)$ , and a height map  $h : \mathbb{R}^2 \rightarrow \mathbb{R}$ , the modified SDF is

$$f_h(\xi) = f(\xi) - h \circ u \circ \Pi_f(\xi), \quad (56)$$

where the projection  $\Pi_f$  can be computed by:

$$\Pi_f(\xi) = \xi - f(\xi) \nabla f(\xi), \quad (57)$$

when  $\xi$  is close to the surface  $M = f^{-1}(0)$ . We demonstrate this computation in 2D in Fig. 14.

To accelerate the computation, we simplify the function  $f_h$  away from the  $\epsilon$ -shell of the original surface by

$$f_h(\xi) \approx \begin{cases} f(\xi) - \epsilon & \text{if } |f(\xi)| > \epsilon\tau \\ f(\xi) - (h \circ u \circ \Pi_f)(\xi) & \text{else,} \end{cases} \quad (58)$$

for some multiplier  $\tau > 1$ .

Zanni et al. had a similar expression in [ZBL\*12], except they only considered noise functions defined on the surface instead of the height map term  $h \circ u$ . Sclaroff et al. also had an alternative formulation for displacement using warping [SP91]. Compared to previous works, our formulation generalizes more easily to other co-dimensions since it does not require normals on the shape. For example, as displayed in the inset, applying this displacement to the SDF of a 1D object (here a line) yields the surface of revolution. When no normal map is associated with the displacement, we provide a simple approximation of the normal of the displaced shape in Sec. E of the supplemental.



### 5.3 Encoding complex geometries as displacement maps of simple implicit surfaces

The displacement map presented in Eq. 58 can be used to add local geometric details on an implicit surface. However, one could also use it to encode the entire shape as a deformation of a simple one, as displayed on Fig. 13. This approach appears to have many advantages since the complexity of the uv computation is proportional to the complexity of the shape we walk on. For simple shapes like spheres, the displacement computation is almost free. Of course, one direct limitation is that the target shape  $T$  must belong to the set of shapes  $\Phi_n(B)$  that are a deformation in the normal direction of the base shape  $B$ , i.e.:

$$\Phi_n(B) = \{S \in \Sigma^2(\mathbb{R}^3), \text{ s.t. }, \exists h : B \rightarrow \mathbb{R}^+, S = B + hn_B\}, \quad (59)$$

where  $\Sigma^2(\mathbb{R}^3)$  is the set of all 2-manifolds of  $\mathbb{R}^3$  and  $B + hn_B = \{x + h(x)n_B(x), x \in B\}$ . For example, when  $B = \mathbb{S}^2$  is a sphere centered at the origin,  $\Phi_n(\mathbb{S}^2)$  is the set of star-shaped objects. For applications like character animation, this limitation is not that restrictive, since the head could be a deformed sphere and the arms and legs deformed cylinders. The conversion process is straightforward since it only requires computing ray intersection queries from a sample of  $B$  to  $T$  and then storing the ray's length in the height map at the uv computed by Alg. 1. The conversion process is parallel and can be made almost instantaneous with a BVH structure when  $T$  is a mesh. Note that one can also capture the normals and the colors of the shape  $T$  in other maps. The resulting shape can then be rendered in real-time in a sphere-tracing pipeline while benefiting from the modeling possibilities offered by implicit approaches, such as animation or topological changes. This could be an interesting way to encode shapes since it benefits from the advantages of both explicit and implicit representations while being efficient, controllable, and lightweight without training neural networks.

### 5.4 Implicit shell maps

Getting access to uv-coordinates on a surface enables querying 2D data, but one can also use them to define a 3D coordinate system that follows the geometry of said surface. Such a coordinate system

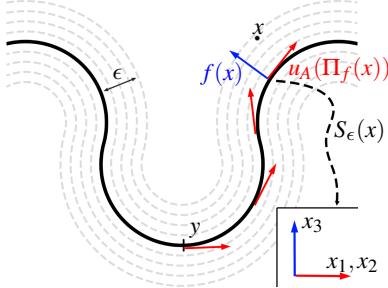
is known as a *shell map* [PBFJ05], and it is defined on an offset of the original surface, a *shell space*. An  $\epsilon$ -shell space is defined as :

$$M^\epsilon = \{x \in \mathbb{R}^3 \text{ s.t. } d(x, M) \leq \epsilon\}. \quad (60)$$

We exploit the fact that, when  $M$  is the 0 iso-surface of a signed distance function  $f$ , such an offset is trivially defined as  $M^\epsilon = \{-\epsilon \leq f \leq \epsilon\}$ . In such a shell, we implicitly define a shell map from a 2D coordinate system on  $M$ ,  $x \in M \mapsto (u(x), v(x)) \in \mathbb{R}^2$  in the following way:

$$S_\epsilon : M^\epsilon \rightarrow \mathbb{R}^2 \times [-1, 1]$$

$$x \mapsto \begin{pmatrix} u(\Pi_f(x)) \\ v(\Pi_f(x)) \\ f(x) \end{pmatrix} (61)^{-1}$$



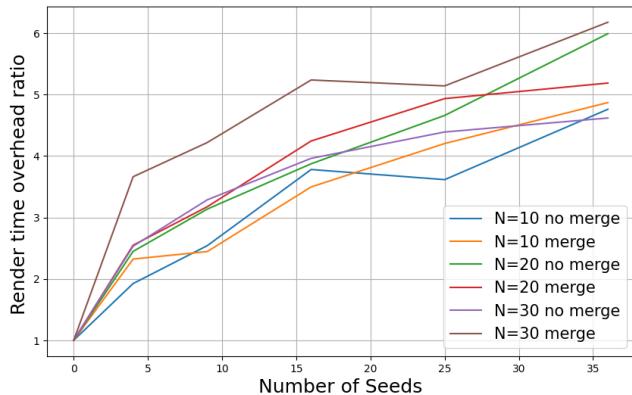
As an application, one can copy the geometry induced by another implicit function  $g$  in the shell-space, by defining a new function  $f_g^\epsilon$  as:

$$f_g^\epsilon(x) = \min(f(x), \epsilon g(S_\epsilon(x))) \quad (62)$$

Defining shell maps on implicit representations has been explored. In [BMPB08], the uvs and the shell coordinates are defined using a cage of particles placed on the surface. However, since the surface is represented by voxel grids, their method is computationally heavy and not suited for real-time processing and evolving geometries. Our method remains fully implicit. Therefore, we can directly use the above embedding by using our Atlas system in a sphere-tracing pipeline as displayed in Fig. 13.

## 6 Performance

The performance of our pipeline is highly dependent on all of the following factors:



**Figure 15:** Plot of the render time (shown as ratio to render time without seeds) with increasing number of seeds, with and without merging. The seeds are placed uniformly on a flat geometry and rendered from a fixed camera. We tested different numbers of geodesic tracing iterations  $N = 10, 20, 30$  and with or without merging. Timings were captured on a MacBook Pro M1 Max.

- Geometry of the surface  $M = f^{-1}(0)$ ,
- Evaluation complexity of  $f$ ,
- Screen resolution,
- Distance between camera and  $M$ ,
- Number of steps  $N$ , number of seeds  $n$ , seed placement, number of blending pairs  $|E|$ , blending width  $\sigma$ ,

We implemented our method in GLSL and ran all our experiments on a laptop with an RTX 4080 laptop GPU on a Full HD screen. In our experience, most color-textured (including normal and roughness maps) examples achieved more than 40 frames per second, including the blue alien toy (Fig. 1 bottom left), bunny (Fig. 7). Even for models as complex as the robot in Fig. 1 (top row), which consists of 42 seeds and 9 blending pairs, the fully color-textured visualization still reaches 25 fps. After adding displacement maps, the performance of rendering the column (Fig. 2 right) dropped from 60 fps to 40 fps and the bunny (Fig. 2 left) to 10 fps, though the sphere-displaced-to-head model (Fig. 13) remained at 60 fps. As displayed in Fig. 15, the most impacting factor on the performances is the number of seeds. Due to parallelism and optimized tests (Eq. 44), the overall complexity of the system is sub-linear in the number of seeds.

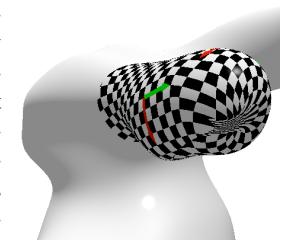
## 7 Limitations and Future works

In this paper, we propose a tool to perform *implicit* queries of *explicit* data. We use an ambient space heuristic to compute an approximation of the geodesic between two points and use it to derive the 2D coordinates in a point-wise manner. Our method enables one to apply additional details using texture mapping while still benefiting from the modeling advantages of implicit surfaces.

The limitations of our method mainly stem from our heuristical approximation of the log map. Since no global shortest path computation is performed, the gradient descent might get stuck in local minima when the geometry has a lot of concavities. While the merging system can help to overcome such locality, as displayed in Fig. 12, the segmentation imposed by the sharp edges remains. For example, as displayed on the inset, narrow regions delimited by sharp features may require many seeds for full coverage.



While the fact that the approximated geodesics comes from a flow guarantees local-injectivity for single seeds, the merging system cannot control the global injectivity of the final uv-field. Moreover, since the computation of the offsets  $u_i$  and frames only takes into account the local geometry around the seeds, high distortion and sub-optimal seams may appear where the geometry is very different from the source seeds as displayed in the inset.



The ambient heuristic and the point-wise nature of our method can also affect the merging of uv-fields. In the scenario where two seeds  $p_i, p_j$  are selected for merging, if a query point  $x$  can reach

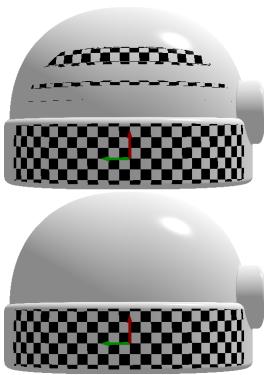
one but not the other, even though we can compute the geodesic between  $p_i$  and  $p_j$ , we still cannot evaluate the merged uv-coordinates for  $x$ .

Since the method is fully pointwise, complete spatial coherence is hard to guarantee. For example, the sharp feature detection might not give spatially coherent results when the number of steps  $N$  for Alg. 1 is too low. As displayed in the inset, for  $\epsilon_2 = 0.015$ , the jump between the two parts of the head is too small to be detected by some points for  $N < 14$ , but is uniformly detected when  $N \geq 14$ .

While the performance of rendering a fully textured model remains real-time on most shapes, displacement mapping (Sec. 5.2.2) and implicit shell maps (Sec. 5.4) are often slower. These two operations are highly dependent on the complexity of the base SDF because they involve recomputing the UV at each sphere-tracing iteration in the shell. Therefore, complex SDFs might not be rendered in real-time when displaced. Having a way to start the computation of a coordinate from the one of the previous iterations could greatly speed up the process. Using accelerating structures on the seeds and multi-scale rendering could also greatly improve the performance.

The interactivity granted by the implicit nature of our method allows one to place seeds and modify the uv-fields easily. An interesting future work could be the automatic placement of the seeds and the generation of the merging graph, for example using a variational approach.

Finally, the displacement of a simple shape into a more complex one (Sec. 5.3) appears to be a promising representation of shapes, falling in the intersection of the implicit and explicit categories. Fitting a complex geometry into a simple base SDF, possibly using a small neural network [CB24], with a height map using our method could potentially express a wide range of shapes.



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