

## Decorating 3D models with Poisson vector graphics

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### ABSTRACT

This paper proposes a novel method for decorating 3D surfaces using a new type of vector graphics, called Poisson Vector Graphics (PVG). Unlike other existing techniques that frequently require local/global parameterization, our approach advocates a parameterization-free paradigm, affording decoration of geometric models with any topological type while minimizing the overall computational expenses. Since PVG supports a set of simple discrete curves, it is straightforward for users to edit colors and synthesize geometry details. Meanwhile, the details could be organized by Poisson Region (PR), leading to much smoother decoration than those of Diffusion Curve (DC). Consequently, it is an ideal tool to create smooth relief. It may be noted that, DC is adequate to create sharp or discontinuous results. But PR is superior to DC, supporting level-of-details editing on meshes thanks to its smoothness. To render PVG on meshes efficiently, we develop a Poisson solver based on harmonic B-splines, which could be constructed using geodesic Voronoi diagram. Our Poisson solver is a local solver for rendering with more flexibility and versatility. We demonstrate the efficacy of our approach on synthetic and real-world 3D models.

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## 1. Introduction and motivation

Despite many novel techniques being developed for 3D graphics modeling in recent decades, decorating 3D surfaces of arbitrary topological type still remains a common yet challenging task in geometric modeling and computer graphics applications. Texture mapping is still a popular technique for enhancing the realism of 3D surfaces, and the other commonly-used technique is bump mapping. Based on the local details of the texture, Sander et al. [1] reallocated the object parameterization to locally provide more details on the mesh. Carr and Hart [2] used the surface painting to paint a texture directly onto a surface with a texture atlas. To provide a more clear shadow map result near the camera according to the current viewpoint, Stamminger and Drettakis [3] changed the texture space by perspective shadow map. Dachsbacher and Stamminger [4] rendered procedural terrain via geometry image warping. However, since local or global parameterization is required, texture mapping is oftentimes computationally expensive. From the authoring perspective, 3D editing tools could not be easily supported by texture mapping, as texture mapping frequently fails

to handle models with complex geometry, and it also tends to produce high distortion.

An alternative approach is to directly construct color and displacement functions on 3D surfaces using vector graphics. Orzan et al. [5] proposed diffusion curves, which diffuse the user-specified colors from the curves and produce smooth color distribution in space except at the curves. Jeschke et al. [6] rendered surface details using diffusion curves. Due to the non-vanishing Dirichlet boundary conditions applied to diffusion curves, their method can produce only sharp features.

Strongly inspired by the existing works on decoration of arbitrary mesh, we take a different approach and propose a 3D mesh decoration algorithm based on Poisson vector graphics in order to produce smooth details. PVG [7] is a natural extension to the popular diffusion curves with two new primitives – Poisson curve (PC) and Poisson region, which is a good 2D graphic authoring tool based on Poisson's equation. It extends diffusion curve from Laplacian's equations to Poisson's equation, allowing non-zero Laplacians away from curves. In this paper, we generalize PVG from planar domain to 3D mesh decoration. Diffusion curves based decoration algorithm suffers from  $C^0$  or  $C^{-1}$  continuity. In contrast, thanks to the  $C^1$  continuity of Poisson regions, our algorithm is adequate to produce smooth local details.

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To render vector graphics, Orzan et al. [5] designed a multi-grid solver to use a coarse version of the sections. Jeschke et al. [8] used finite differences with varying step size to accelerate the calculation process. Boye et al. [9] presented a finite element method (FEM) which can convert a diffusion curve image into a mesh-based representation. It may be noted that, a *global* solver is not necessary for mesh rendering, since only faces within or adjacent to the painting/decoration region shall be further processed. Based on the vector solver of PVG, we devise a *local* Poisson solver to render mesh details efficiently, while avoiding to process global details. With the local solution, only the vertices that contribute to the results are evaluated. Our contributions are as follows:

- At the theoretic and computational front, we propose an efficient Poisson solver directly on 3D meshes, which is a local solver for rendering with more flexibility.
- In the application of 3D decorations, PVG provides users with simple inputs (i.e., a set of sparse curves) to produce detailed geometries. In particular, smooth 3D decorations could directly benefit from PR within the framework of PVG, which could not be properly enabled by using conventional DCs.
- Thanks to the smoothness property inherited from PR, our new method will afford level-of-details editing directly on complex geometries.

The remaining of the paper is organized as follows. Section 2 reviews the related works and Section 3 introduces the preliminary knowledge and required terminologies on Poisson vector graphics and harmonic B-splines. Section 4 presents our PVG mesh solver in details and Section 5 addresses our specific algorithm, followed by experimental results and discussions in Section 6. Finally, Section 7 concludes the paper and points out the future work.

## 2. Related work

This section briefly reviews the work relevant to ours.

### 2.1. Surface decoration

Besides texture mapping, there are a few other alternatives for decorating 3D surfaces. Bump mapping [10–12] simulates 3D surface details by perturbing normal vectors while the original geometry remains unchanged. In contrast to bump mapping that can only control shading normals, parallax mapping [13,14] can also modify the texture coordinates along the view direction to obtain mesostructure normals and color data based on the depth values using an approximate solution. Relief texture mapping [15,16] uses textures enhanced with depth information to create the illusion of complex geometric details onto flat polygons with locating the intersection of the height field and the ray to generate pixel shader. In contrast to bump mapping, displacement mapping [17–19] actually changes model's geometry by moving vertices along surface normals.

Using texture and displacement mappings, Elber [20] synthesized detailed geometry on the surfaces of objects by mapping the geometric texture. Zhou et al. [21] proposed a mesh quilting algorithm to seamlessly synthesize geometric texture sample inside a thin shell around an arbitrary surface through local stitching and deformation. Lai et al. [22] converted the 2D texture into geometric image to reconstruct synthesized geometry. Bhat et al. [23] presented an example-based method for synthesizing geometric textures such as pits and grooves on surfaces by extending the neighborhood-based texture synthesis algorithms for volumetric models. By reconstructing the fine-scale geometric details over a simple proxy of the original model, Toledo et al. [24] proposed a suitable representation for highly tessellated models using a set

of geometry textures. Landreneau and Schaefer [25] developed a method for generating scales and scale-like structures on a polygonal mesh. Taking a user-sketched lateral line as input (which controls the distribution and orientation of scales), it computes a vector field over the surface to control an anisotropic centroidal Voronoi tessellation and then automatically fills each Voronoi cell with oriented scales and generates a fully connected 2-manifold mesh that is suitable for subsequent post-processing applications. Their method is effective to generate pre-defined patterns repetitively on 3D surfaces, while our method supports free-form vector graphics and can also provide the user more control of the displacement.

### 2.2. Vector graphics

Diffusion curves, proposed by Orzan et al. [5], are 2-sided curves with colors defined on either side. By diffusing these colors over the image, the resulting image includes sharp boundaries along the curves with smoothly shaded regions between them. Note that diffusion curve images are harmonic functions of colors, the maximum principle states that a non-constant harmonic function cannot attain a maximum (or minimum) at an interior point of its domain. Therefore, the follow-up work focuses on providing more degrees of freedom for controlling color gradient. Bezerra et al. [26] proposed diffusion barriers, diffusion anisotropy, and spatially varying color strength to control the diffusion process. Their approach is able to diffuse both colors and normal maps, hereby producing interesting non-photorealistic effects. However, it is non-intuitive to specify the boundary condition for normals and normals can only be diffused within a closed diffusion curve. Using thin-plate splines (TPS), Finch et al. [27] extended diffusion curves to provide smooth interpolation through color constraints. Although TPS allows more user control and is able to mimic smooth shading, it often produces unwanted local extremals (hereby unpredicted colors) due to the violation of the maximal principle of harmonic equation. Moreover, solving a bi-Laplace's equation is more computationally expensive than solving Laplace's equation, and it may suffer from serious numerical issues since the system is less well-conditioned. Lieng et al. [28] proposed shading curves, which associate shading profiles to each side of the curve. These shading profiles, which can be manually manipulated, represent the color gradient out from their associated curves. Lecot and Levy [29] developed a vectorization method which is based on a two-level variational parametric segmentation algorithm, minimizing Mumford and Shah's energy and operating on an intermediate triangulation, well adapted to the features of the image. Using holomorphic 1-form, Lai et al. [30] developed an automatic method for converting raster images to high-quality gradient meshes with non-trivial topologies. Xie et al. [31] automatically generated sparse diffusion curve vectorizations of raster images by fitting curves in the Laplacian domain. Their method is highly efficient, combines Laplacian and bi-Laplacian diffusion curve representations, and generates a hierarchical representation that accurately reconstructs both vector art and natural images.

## 3. Preliminary

In this section, we briefly introduce harmonic B-splines, diffusion curves and Poisson vector graphics. For details, we refer readers to [32,33]. To ease reading, we list the main notations in Table 1.

**Table 1**  
Notations.

Item	Description
$\Omega$	2D domain
$\mathcal{M}$	Input 3D triangle mesh
$U \subset \mathcal{M}$	A simply connected region on the mesh
$\mathbf{x}, \mathbf{y}$	Vertices of $\Omega$ or $\mathcal{M}$
$\mathbf{n}$	Normal vector
$\mathbf{t}_i$	The $i$ th knot of $\Omega$ or $\mathcal{M}$
$\bigcup_i \mathcal{V}_i$	Voronoi diagram
$N_1(\mathcal{V}_i)$	1-ring neighboring cells of $\mathcal{V}_i$
$n$	Number of knots (Voronoi cells)
$A_{\mathcal{M}}$	The total area of mesh $\mathcal{M}$
$A_{\mathcal{V}_i}$	The area of Voronoi cell $\mathcal{V}_i$
$A_{\mathbf{x}}$	The area of Voronoi region of vertex $\mathbf{x}$
$e_{ij}$	Voronoi edge between cells $\mathcal{V}_i$ and $\mathcal{V}_j$
$d_{ij}$	Geodesic distance between $\mathbf{t}_i$ and $\mathbf{t}_j$
$f$	Laplace constraints of PCs and PRs
$g$	The Dirichlet boundary condition
$u(\mathbf{x})$	The solution of Poisson equation
$\Delta$	The Laplace–Beltrami operator
$\phi_{\mathbf{y}}(\mathbf{x})$	Green's function centered at vertex $\mathbf{y}$

### 3.1. Harmonic B-splines

Let  $\Omega \subset \mathbb{R}^2$  be a 2D compact domain and  $\mathcal{T} = \{\mathbf{t}_i | \mathbf{t}_i \in \Omega\}_{i=1}^n$  a set of knots. Taking  $\{\mathbf{t}_i\}$  as the generators, we construct a Voronoi diagram  $\Omega = \bigcup_{i=1}^n \mathcal{V}_i$ , where  $\mathcal{V}_i$  is the Voronoi cell of knot  $\mathbf{t}_i$ .

For arbitrary points  $\mathbf{x}, \mathbf{y} \in \Omega$ , Green's function of the Laplace operator  $\Delta$  satisfies

$$\Delta \phi_{\mathbf{y}}(\mathbf{x}) = \delta_{\mathbf{y}}(\mathbf{x}), \quad (1)$$

where  $\delta_{\mathbf{y}}(\mathbf{x})$  is the Dirac delta function centered at  $\mathbf{y}$ . In 2D space, an analytic solution to Eq. (1) is

$$\phi_{\mathbf{y}}(\mathbf{x}) = \frac{1}{2\pi} \log(\|\mathbf{x} - \mathbf{y}\|).$$

For a Voronoi cell  $\mathcal{V}_j$ , applying Green's theorem to (1) yields

$$\int_{\mathcal{V}_j} \Delta \phi_{\mathbf{y}}(\mathbf{x}) d\sigma = \int_{\partial \mathcal{V}_j} \frac{\partial \Delta \phi_{\mathbf{y}}(\mathbf{x})}{\partial \mathbf{n}} ds, \quad (2)$$

where  $\mathbf{n}$  is the outward unit normal to the boundary  $\partial \mathcal{V}_j$ ,  $d\sigma$  and  $ds$  are the area and line integral elements, respectively.

Then we define a function  $\psi_j$  for each Voronoi cell  $\mathcal{V}_j$  as

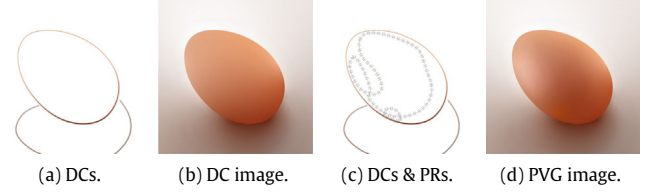
$$\psi_j(\mathbf{x}) = \sum_i w_{ij} \phi_{\mathbf{t}_i}(\mathbf{x}), \quad (3)$$

where  $w_{ij}$  is the discrete Laplacian weight and  $\sum_i w_{ij} \phi_{\mathbf{t}_i}(\mathbf{x})$  is a boundary sum that approximates the line integral on the right hand side of Eq. (2).

Feng and Warren [32] showed that the functions  $\psi_j$  are *approximately* local, nonnegative, and satisfying partition of unity, sharing many properties of the popular B-spline's basis functions. Therefore they named it *harmonic* B-spline. Harmonic B-spline has two salient features that distinguish itself from the conventional B-spline. First, there is no restriction on knots and splines can be constructed on a set of fully irregular knots. Second, it is parameterization free, since evaluating Green's function  $\phi_{\mathbf{y}}(\mathbf{x})$  requires only the distance  $\|\mathbf{x} - \mathbf{y}\|$ , which can be measured in a coordinate-free manner.

### 3.2. Diffusion curves & Poisson vector graphics

Orzan et al. [5] developed diffusion curves, which are 2-sided curves with colors defined on either side. By diffusing these colors over the image, the final result includes sharp boundaries along the curves with smoothly shaded regions between them (see .



**Fig. 1.** Diffusion curves and Poisson vector graphics [7]. PVG complements DC by two new primitives, Poisson regions and Poisson curves, which provide additional control on colors. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

Fig. 1(a)(b)). Mathematically speaking, diffusion curve image is the solution of the following Laplace equation  $\Delta u(\mathbf{x}) = 0$ ,  $u(\mathbf{x})|_{\partial \Omega} = g$ , where  $g$  is user-specified boundary color. Since the solution of Laplace equation is a harmonic function, which, by the maximum principle, retains its maximum and minimum on the domain boundary, implying that  $u$  cannot have local maxima or minima, unless it is constant. As a result, diffusion curves do not have enough degrees of freedom to control color gradient.

Poisson vector graphics [7] extends the DC framework by adding two new geometric primitives, called Poisson curves and Poisson regions. The former is to model color discontinuity across curves, while the latter is to design smooth shading within the user specified regions. PVG solves a Poisson's equation with piecewise constant Laplacians  $f$ , hereby taking DC as a special case with  $f \equiv 0$ . Extending the zero Laplacian to a piecewise constant function  $f$  brings 4 unique advantages. First, users can explicitly control the local and/or global shading profiling via manipulating  $f$  (which is a scalar for each color channel). Second, users can easily control the color extrema, which are either on the curves (for PC and DC) or inside a region (for PR). Third, PVG allows intersection among the geometric primitives. Fourth, PVG natively supports seamless cloning. Although a PVG can have an arbitrary number of PCs and PRs, it must contain at least one diffusion curve, serving as the boundary condition  $g$ . Hou et al. [7] demonstrated that PVG is a simple yet powerful authoring tool that can produce photo-realistic vector graphics from scratches. See Fig. 1(c)(d) for a simple example of PVG.

### 3.3. 2D PVG Solver

To render PVG, one solves the following Poisson equation

$$\Delta u(\mathbf{x}) = f, \quad u(\mathbf{x})|_{\mathbf{x} \in \partial \Omega} = g, \quad (4)$$

where  $f$  is the Laplacian constraint of Poisson curves and regions, and  $g$  is the Dirichlet boundary condition of colors. Using harmonic B-splines [32], Hou et al. [7] developed a novel random-access solver for the above Poisson equation.

Green's third identity provides an analytical solution for Laplace/Poisson's equations in the form of integral,

$$u(\mathbf{x}) = \iint_{\Omega} \phi_{\mathbf{y}}(\mathbf{x}) \Delta u(\mathbf{y}) d\sigma_{\mathbf{y}} + \oint_{\partial \Omega} \left( u(\mathbf{y}) \frac{\partial \phi_{\mathbf{y}}(\mathbf{x})}{\partial \mathbf{n}} - \phi_{\mathbf{y}}(\mathbf{x}) \frac{\partial u(\mathbf{y})}{\partial \mathbf{n}} \right) dl_{\mathbf{y}}, \quad (5)$$

where  $d\sigma$  and  $dl$  are the surface and line elements,  $\mathbf{n}$  is the outward pointing unit normal of  $dl$ .

Hou et al. [7] rasterized the image domain  $\Omega$  with user-specified resolution and then discretized the geometric primitives (DCs/PCs/PRs) using quad-tree. Taking the quad-tree nodes as generators, they computed a Voronoi diagram to partition the domain  $\Omega = \bigcup_i \mathcal{V}_i$ . For each Voronoi cell  $\mathcal{V}_i$ , they constructed a harmonic

B-spline basis function  $\psi_i(\mathbf{x})$ , whose knots are the generators of  $\mathcal{V}_i$  and its 1-ring neighboring cells.

For basis function of a boundary Voronoi cell  $\mathcal{V}_i$ , they set its control point  $\lambda_i$  using the Dirichlet boundary condition  $g$ . For basis functions of internal Voronoi cells, they computed their control points by solving a sparse linear system of size  $k \times k$ , where  $k$  is the number of interior quad-tree nodes. Finally, the solution is given by  $u(\mathbf{x}) = \sum_{i=1} \lambda_i \psi_i(\mathbf{x})$ . Since harmonic B-spline basis functions are approximately local, evaluating  $u(\mathbf{x})$  involves only the basis functions close to  $\mathbf{x}$ .

#### 4. PVG mesh solver

Let  $\mathcal{M}$  be a closed manifold mesh and  $U \subset \mathcal{M}$  a simply connected region enclosing the PVG primitives. To render PVG on meshes, we adopt the harmonic B-spline based solver similar to [7]. For a closed mesh  $\mathcal{M}$ , Green's function of the Laplace–Beltrami operator is

$$\Delta \phi_{\mathbf{y}}(\mathbf{x}) = \delta_{\mathbf{y}}(\mathbf{x}) - \frac{1}{A_{\mathcal{M}}}, \quad (6)$$

where  $A_{\mathcal{M}}$  is the area of  $\mathcal{M}$ . The addition of the area term is necessary to ensure the existence of a solution  $\mathbf{y}(\mathbf{x})$  in the compact case. Using Green's third identity on closed manifolds, we can write the solution  $u(\mathbf{x})$  as

$$u(\mathbf{x}) = \frac{1}{A_{\mathcal{M}}} \iint_U u(\mathbf{x}) d\sigma + \iint_U \Phi_{\mathbf{y}}(\mathbf{x}) \Delta u(\mathbf{x}) d\sigma. \quad (7)$$

Comparing to the planar case (Eq. (5)), we have the additional term  $\frac{1}{A_{\mathcal{M}}} \iint_U u(\mathbf{x}) d\sigma$  which is the average of  $u(\mathbf{x})$  in region  $U$ . Such a term reveals the difference of Green's function defined on closed and open domains.

We select a set of knots  $\{\mathbf{t}_j\}_{j=1}^n$  which is a subset of mesh vertices and construct a Voronoi diagram on  $\mathcal{M}$ . The knot selection criteria are as follows: (1) the outer boundary of  $\mathcal{V}$  coincides with the boundary of  $U$ . (2) the regions on the boundary (i.e.,  $\partial \mathcal{V}_j \cap \partial U \neq \emptyset$ ) are small enough. (3) the function  $f$  in every  $\mathcal{V}_j$  is a constant. Similar to [7], we discretize  $u(\mathbf{x})$  as

$$u(\mathbf{x}) \approx \frac{1}{A_{\mathcal{M}}} \iint_U u(\mathbf{x}) d\sigma + \sum_{\mathcal{V}_j} A_{\mathcal{V}_j} f|_{\mathcal{V}_j} \bar{\phi}_{\mathcal{V}_j}(\mathbf{x}). \quad (8)$$

Feng and Warren [32] defined the basis function for Voronoi cell  $\mathcal{V}_j$  as

$$\psi_j(\mathbf{x}) = \frac{A_{\mathcal{V}_j}}{A_{\mathcal{M}}} + \sum_{\mathcal{V}_i \in \mathcal{N}_1(\mathcal{V}_j)} a_{ij} (\bar{\phi}_{\mathcal{V}_i}(\mathbf{x}) - \bar{\phi}_{\mathcal{V}_j}(\mathbf{x}))$$

where the coefficient  $a_{ij} = -\frac{\|e_{ij}\|}{d_{ij}}$ , and the Voronoi edges  $e_{ij}$  are fallen into three disjoint classes geometrically: inner boundaries  $\mathcal{E}_{ib}$ , outer boundaries  $\mathcal{E}_{ob}$  and interior edges  $\mathcal{E}_{ie}$ .

Then we can derive

$$\begin{aligned} & \sum_{\mathcal{V}_j} \lambda_j \psi_j \\ &= \sum_{\mathcal{V}_j} \lambda_j \frac{A_{\mathcal{V}_j}}{A_{\mathcal{M}}} + \sum_{e_{ij} \in \mathcal{E}_{ob}} a_{ij} (\bar{\phi}_{\mathcal{V}_i} - \bar{\phi}_{\mathcal{V}_j}) \lambda_j \\ &= \sum_{\mathcal{V}_j} \lambda_j \frac{A_{\mathcal{V}_j}}{A_{\mathcal{M}}} + \sum_{\mathcal{V}_j \in \mathcal{I}} \left( \bar{\phi}_{\mathcal{V}_j} \sum_{\mathcal{V}_i \in \mathcal{N}_1(\mathcal{V}_j)} a_{ij} (\lambda_i - \lambda_j) \right) \end{aligned} \quad (9)$$

For  $\lambda_j$  of  $\mathcal{V}_j \in \mathcal{B}$ , we simply set  $\lambda_j = g|_{\mathcal{V}_j}$ . To compute the coefficients of  $\bar{\phi}_{\mathcal{V}_j}$  ( $\mathcal{V}_j \in \mathcal{I}$ ) satisfying  $A_{\mathcal{V}_j} f|_{\mathcal{V}_j}$ , we solve the following equation

$$\mathbf{L}^{\mathcal{I}} \boldsymbol{\lambda}^{\mathcal{I}} = \mathbf{b} - \mathbf{L}^{\mathcal{B}} \boldsymbol{\lambda}^{\mathcal{B}}, \quad (10)$$

where  $b_j = A_{\mathcal{V}_j} f|_{\mathcal{V}_j}$ . It is apparent that  $\lambda_j$  is the average color or displacement in  $\mathcal{V}_j$ . It is easily verified that the third and fourth terms of Eqn. (9) are identical to inner boundaries and outer boundaries of Eqn. (8). Since  $\lambda_j$  is the average color or displacement of  $\mathcal{V}_j$ ,  $\sum_{\mathcal{V}_j \in \mathcal{I} \cup \mathcal{B}} \lambda_j \frac{A_{\mathcal{V}_j}}{A_{\mathcal{M}}}$  is the average color/displacement of region  $U$ . Thus, Eqn. (9) is the solution of the Poisson equation (4).

Since the basis function  $\psi_j(\mathbf{x})$  decays to zero quickly, we evaluate  $u(\mathbf{x}) = \sum_{\mathcal{V}_j \in \mathcal{N}(\mathbf{x})} \lambda_j \psi_j(\mathbf{x})$  using only the basis functions in a local neighborhood of  $\mathbf{x}$ .

**Remark.** Our solver can be extended to handle open meshes. By removing the constant term and adding the boundary integral to Eqn. (7), we can express the solution as

$$\begin{aligned} u(\mathbf{x}) &= \frac{1}{A_{\mathcal{M}}} \iint_U u(\mathbf{x}) d\sigma + \iint_U \Phi_{\mathbf{y}}(\mathbf{x}) \Delta u(\mathbf{x}) d\sigma \\ &+ \oint_{\partial U} \left( u(\mathbf{x}) \frac{\partial \Phi_{\mathbf{y}}(\mathbf{x})}{\partial \mathbf{n}} - \Phi_{\mathbf{y}}(\mathbf{x}) \frac{\partial u(\mathbf{x})}{\partial \mathbf{n}} \right) dl. \end{aligned} \quad (11)$$

The basis function for Voronoi cell  $\mathcal{V}_j$  is

$$\psi_j(\mathbf{x}) = \sum_{\mathcal{V}_i \in \mathcal{N}_1(\mathcal{V}_j)} a_{ij} (\bar{\phi}_{\mathcal{V}_i}(\mathbf{x}) - \bar{\phi}_{\mathcal{V}_j}(\mathbf{x})),$$

so we obtain the solution  $u(\mathbf{x}) = \sum_{\mathcal{V}_j \in \mathcal{N}(\mathbf{x})} \lambda_j \psi_j(\mathbf{x})$  in a similar form as the case of closed meshes.

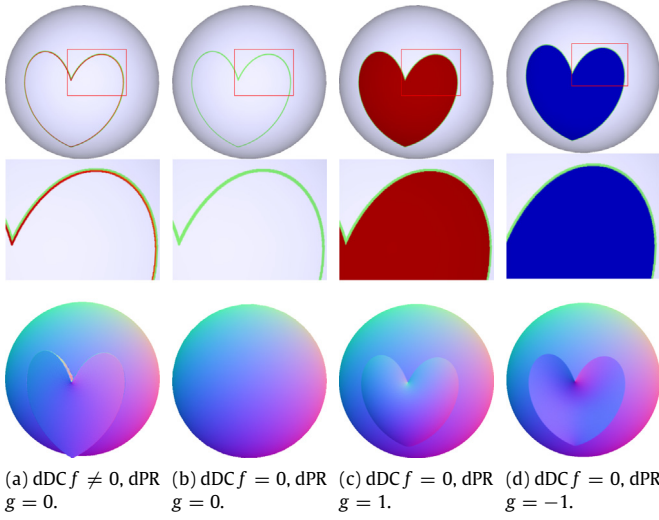
#### 5. Algorithm

##### 5.1. Overview

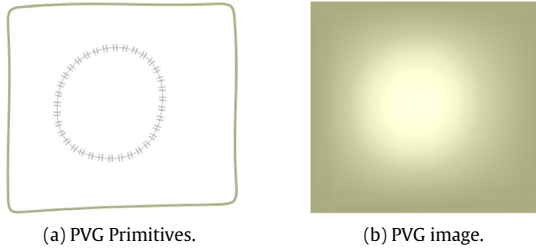
PVG consists of three types of primitives: diffusion curve, Poisson curve and Poisson region. DC is a curve with color (cDC) or displacement (dDC) specified by the user, meanwhile, it guarantees a unique solution of the Poisson solver as Dirichlet boundary condition. DC is like a radiative heat transport diffusing color or height on the mesh. PC is a double-sided curve with opposite Laplacian values and can create high contrast effect along the curve. PR is a region whose boundary and inside region have reverse Laplacians and can generate a controllable soft boundary there. PR is effective to produce the effects like highlight and smooth convex or concave bump effect. In short, DCs specify color and height on the mesh as the boundary condition, PCs and PRs control tone by offsetting Laplacians.

In this paper, we adopt diffusion curves and Poisson regions to represent both colors and displacements on 3D surfaces, and use prefixes ‘‘c’’ and ‘‘d’’ to distinguish them. For example, cDC is a color field represented by diffusion curves and dPR is displacement generated by Poisson regions. Fig. 5 shows a simple example with 5 cDCs, 4 dDCs and 1 dPR. Besides, from Fig. 5(e)(g), we can see that how PRs make the difference from traditional DCs to achieve smooth tune result. As shown in Fig. 2, DC alone produces only sharp feature, whereas DC+PR can make a smooth displacement field. Furthermore, PRs are continuous across their boundaries for 2D cases, thanks to the analytic solution of Green's functions (see Fig. 3). Unfortunately, we do not have such a luxury for general 3D meshes, the  $C^1$ -continuity does not hold any longer. Nevertheless, editing PRs is still more flexible than DCs, since PRs can intersect with each other, whereas DCs cannot (see Fig. 4).

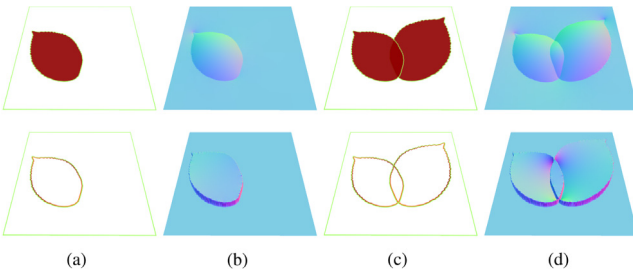
Fig. 6 illustrates the pipeline of our surface decoration algorithm. Taking the user-sketched curves as input, we take a subset of mesh vertices as knots and then construct a geodesic Voronoi diagram. Then, for each Voronoi cell, we construct a harmonic B-spline basis function  $\psi$ . By solving small sparse linear system, we obtain the control coefficients  $\lambda$  and finally express the color/displacement function using  $\sum_i \lambda_i \psi_i$ .



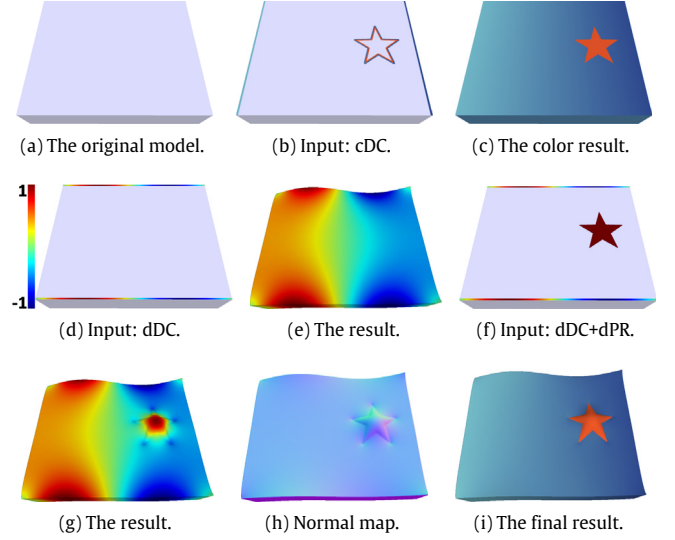
**Fig. 2.** Diffusion curves vs Poisson regions. Diffusion curves are associated with the Dirichlet boundary condition  $f$ , while Poisson regions are assigned Laplace constraints  $g$ . We solve the Poisson equation  $\Delta u = g$  and  $u|_{\partial\Omega} = f$  to compute the displacement field. (a) Although the displacement generated by DC is smooth within  $\Omega$ , it is discontinuous along the domain boundary  $\partial\Omega$ . (b) The only exception is  $f \equiv 0$ , which produces a constant function  $u \equiv 0$ . (c)–(d) Combining DC and PR, we can define displacement which is continuous both inside  $\Omega$  and across the boundary  $\partial\Omega$ . We visualize the values of  $f$  and  $g$  using color maps: warm colors are positive values and cold colors are negative values, and green is 0. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)



**Fig. 3.** On 2D domain, PR (dashed curve) is  $C^1$  continuous across its boundary, thanks to the analytical solution of Green's functions of the 2D Laplacian operator. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)



**Fig. 4.** PVG allows intersecting primitives, which is a desired feature for editing. Row 1 shows two PRs intersecting each other. When two diffusion curves are intersecting, their associated boundary conditions are often “competing”, resulting in non-smooth artifacts (see row 2). (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)



**Fig. 5.** Generating a flag model from a cuboid model with simple input based PVG. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

#### Algorithm 1 Decorating 3D meshes with PVG

**Require:** The triangle mesh  $\mathcal{M}$ , the PVG primitives (i.e., diffusion curves and Poisson regions) associated with color and/or displacement constraints

**Ensure:** The mesh with details and colors decorated by PVG

- 1: Compute discrete Green's function  $\phi_{\mathbf{y}}(\mathbf{x})$  using [34]
- 2: Sample knots  $\{\mathbf{t}_i\}_{i=1}^n$  from mesh's vertices
- 3: Construct geodesic Voronoi diagram on  $\mathcal{M}$ , and  $\mathcal{M}$  is divided into a set of sub-regions  $\mathcal{M} = \bigcup_{i=1}^n \mathcal{V}_i$
- 4: // Construct the harmonic B-spline basis function  $\{\psi_i\}_{i=1}^n$
- 5: **for** each Voronoi cell  $\mathcal{V}_j$  **do**
- 6:  $\psi_j(\mathbf{x}) = \frac{A_{\mathcal{V}_j}}{A_{\mathcal{M}}} + \sum_{\mathcal{V}_i \in N_1(\mathcal{V}_j)} a_{ij}(\bar{\phi}_{\mathcal{V}_i}(\mathbf{x}) - \bar{\phi}_{\mathcal{V}_j}(\mathbf{x}))$
- 7:  $\bar{\phi}_{\mathcal{V}_j}(\mathbf{x}) = \frac{1}{A_{\mathcal{V}_j}} \sum_{\mathbf{y} \in \mathcal{V}_j} A_{\mathbf{y}} \phi_{\mathbf{y}}(\mathbf{x})$
- 8: **end for**
- 9: // Compute the control points  $\lambda = \{\lambda_i\}_{i=1}^n$
- 10: **for** each Voronoi cell  $\mathcal{V}_i$  **do**
- 11: Compose  $i$ th row of sparse matrix  $\mathbf{L} = \{a_{ij}\}_{i,j=1}^n$
- 12: **for** every neighbor  $\mathcal{V}_j \in N_1(\mathcal{V}_i)$  **do**
- 13:  $a_{ij} = -\frac{\|e_{ij}\|}{d_{ij}}$
- 14: **end for**
- 15:  $a_{ii} = \sum_{\mathcal{V}_j \in N_1(\mathcal{V}_i)} a_{ij}$
- 16:  $b_i = A_{\mathcal{V}_i} f|_{\mathcal{V}_i}$
- 17: **end for**
- 18: Solve  $\mathbf{L}\lambda = \mathbf{b}$
- 19: // Compute the decorated mesh
- 20: **for** every vertex  $\mathbf{x} \in \mathcal{V}_i$  **do**
- 21:  $u(\mathbf{x}) = \sum_{\mathcal{V}_j \in N(\mathbf{x})} \lambda_j \psi_j(\mathbf{x})$
- 22: **end for**

#### 5.2. Computing discrete Green's function

Define  $g_{ij} = \phi_{\mathbf{v}_i}(\mathbf{v}_j)$  the Green's function centered at vertex  $\mathbf{v}_i$  evaluating at vertex  $\mathbf{v}_j$ . We denote by  $\mathbf{G} = (g_{ij})_{m \times m}$  the discrete Green's function on closed mesh  $M$ , where  $m$  is the number of vertices. Lipman et al. [34] proved that  $\mathbf{A}^{-1}\mathbf{L}\mathbf{G} = \mathbf{I} - \frac{1}{n}\mathbf{1}\mathbf{1}^T$ , where  $\mathbf{1}$  is the column vector of all ones. Denote by  $\mathbf{M}_j$  the  $j$ th column of matrix  $\mathbf{M}$ . They showed that  $\mathbf{G}_j = \mathbf{x} - \frac{\mathbf{1}^T \mathbf{x}}{\mathbf{1}^T \mathbf{1}} \mathbf{1}$ , where  $\mathbf{x}$  is a

particular solution to  $\mathbf{L}\mathbf{A}^{-1}\mathbf{L}\mathbf{x} = (\mathbf{I} - \frac{1}{n}\mathbf{1}\mathbf{1}^T)_j$ . To get the particular solution, replace the first row and the first column of  $\mathbf{A}^{-1}\mathbf{L}$  by zeros and set the diagonal entry at their intersection to 1. Also replace the first row of  $(\mathbf{I} - \frac{1}{n}\mathbf{1}\mathbf{1}^T)_j$  by 0. As pointed out in [34], this kind of linear system can be solved very efficiently by first performing Cholesky factorization of  $\mathbf{L}\mathbf{A}^{-1}\mathbf{L}$  and then performing two backward substitutions for every given vector on the right hand side. Green's functions on open mesh can be similarly solved by  $\mathbf{A}^{-1}\mathbf{L}\mathbf{G} = \mathbf{I}$ .

### 5.3. Computing geodesic Voronoi diagram

While user drawing DC and PR's boundary on the mesh, each curve assigned with double-side color or Laplacian weight are represented by a B-spline curve. According to the work in [35], de Casteljau's algorithm is generalized into 2D manifold which replaced the linear interpolation with the smooth geodesic interpolation. With this framework, we represent PVG primitives by B-spline curves on the mesh. It requires extensive computation of pairwise geodesic paths. Here we adopt an efficient approach [36] to compute approximate geodesic distances between any two points on the mesh. Given the user-sketched B-spline curves, we assign the vertices belonging to the curves the color or Laplacian constraints.

To get high quality solution, we need to sample the knots with sufficient density so that the Laplacian in each Voronoi cell is a constant. We first include all the key vertices involved in the input PVG in the knot set. Then we add additional vertices that are evenly selected with the specified knot density. Next, using the knots as generators, we construct a geodesic Voronoi diagram on the triangle mesh.

The algorithm of building Voronoi diagram in  $\mathbb{R}^2$  has been well implemented such as by using dual Delaunay triangulation. But in  $\mathbb{R}^3$ , the situation becomes more complicated and one of the most differences is that the bisector is a point trace on the mesh which have equal geodesic distance to two knot neighbors rather than a straight line. First, the knots are used as sources to create multi-source exact geodesic distance field on the mesh by applying the improved CH algorithm [37]. After that, each vertex is assigned a geodesic distance to its closest knot. Then, we check each triangle edge of the mesh and label it if its endpoints have different closest knots. And if a labeled edge's incident triangle face involves three labeled edges, it contains a branch point inside, otherwise it is crossed through by a segment of bisector. Finally, based on the labeled edges and faces, we obtain the geodesic Voronoi diagram (Fig. 6(b)) on the mesh through the marching algorithm [38].

### 5.4. Computing control coefficients

According to the PVG constraints defined on the vertices and the constructed geodesic Voronoi diagram, we attempt to solve a sparse linear system to require control coefficients  $\lambda$  for expressing the harmonic B-spline function in the further step. As for constructing this linear system

$$\mathbf{L}\lambda = \mathbf{b} \quad (12)$$

here  $\mathbf{L} = \{a_{ij}\}$  is a sparse symmetric matrix whose size is  $n \times n$  ( $n$  is the number of knots  $\{\mathbf{t}_i\}$ ), and  $\mathbf{b}$  is an  $n$ -dimensional vector defined as each knot's Laplacian value multiplying by its area. Based on the position relationships of knots in Voronoi diagram, they can be obtained specifically as follows:

$$a_{ij} = \begin{cases} -\frac{\|e_{ij}\|}{d_{ij}}, & \text{if } i \neq j \text{ and } \{t_i, t_j\} \text{ are neighbors} \\ 0, & \text{if } i \neq j \text{ and } \{t_i, t_j\} \text{ are not neighbors} \\ \sum_{\mathcal{V}_j \in N_1(\mathcal{V}_i)} a_{ij}, & \text{if } i = j; i, j \in [1, n] \end{cases} \quad (13)$$

$$b_i = A_{\mathcal{V}_i} f|_{\mathcal{V}_i}, i \in [1, n] \quad (14)$$

where  $a_{ij}$  ( $i, j \in [1, n]$ ) is only related to the  $i$ th knot  $\mathbf{t}_i$  and its 1-ring Voronoi neighborhood. Note that no matter how the knots are distributed, on average, each knot has six neighbors (since the average neighboring Voronoi cells is 6). Therefore, for each row of  $\mathbf{L}$ , there are roughly seven non-zero values.

Moreover, allow for  $\text{rank}(\mathbf{L}) = (n - 1)$ , and this linear system Eqn. (12) has infinitely many solutions, we use DCs to assign the Dirichlet boundary condition. By moving them to the right hand side of this equation as Eqn. (10), we can attain a unique solution. That is, the average color (displacement) of each Voronoi cell is obtained (Fig. 6(d)), which will serve as the control coefficients for the next step of constructing harmonic B-spline function. In addition, we know that any PVG consists of at least one DC to ensure a unique solution.

### 5.5. Constructing basis functions

In this paper, we would like to develop a method for decorating 3D meshes in both colors and displacements, making that the Poisson solver is working on all vertices of the model. Of course, the more dense mesh model you use, the higher the resolution result you get. To achieve a good rendering result, the 3D mesh model should have sufficiently high resolution. In our implementation, we use the midpoint subdivision to increase the model's resolution if it is not high enough.

Therefore, the main goal of the Poisson solver is to reduce the computational cost. Here, we propose an efficient method for solving the Poisson equation on meshes through constructing harmonic B-splines. Harmonic B-spline function provides the local computation for solving the Poisson equation on the mesh instead of a global solution covering all vertices as the conventional finite element method does.

As for the construction of harmonic B-spline function on the 3D mesh  $\mathcal{M}$ , the above computation has supplied its control coefficients. According to Green's third identity and previous derivation, the  $j$ th basis function can be defined on the closed mesh by

$$\psi_j(\mathbf{x}) = \frac{A_{\mathcal{V}_j}}{A_{\mathcal{M}}} + \sum_{\mathcal{V}_i \in N_1(\mathcal{V}_j)} a_{ij}(\bar{\phi}_{\mathcal{V}_i}(\mathbf{x}) - \bar{\phi}_{\mathcal{V}_j}(\mathbf{x})), j \in [1, n] \quad (15)$$

and for the open mesh it is:

$$\psi_j(\mathbf{x}) = \sum_{\mathcal{V}_i \in N_1(\mathcal{V}_j)} a_{ij}(\bar{\phi}_{\mathcal{V}_i}(\mathbf{x}) - \bar{\phi}_{\mathcal{V}_j}(\mathbf{x})), j \in [1, n] \quad (16)$$

where  $\mathcal{V}_i$  is 1-ring neighbor of  $\mathcal{V}_j$ ,  $A_{\mathcal{M}}$  is the total area of the mesh,  $A_{\mathcal{V}_j}$  is the area of  $j$ th Voronoi cell  $\mathcal{V}_j$ , and is the average value of Green function  $\phi$  in Voronoi cell  $\mathcal{V}_j$ :

$$\bar{\phi}_{\mathcal{V}_j}(\mathbf{x}) = \frac{1}{A_{\mathcal{V}_j}} \int_{\mathcal{V}_j} \phi(\mathbf{x}, \mathbf{y}) d\delta_{\mathbf{y}} \quad (17)$$

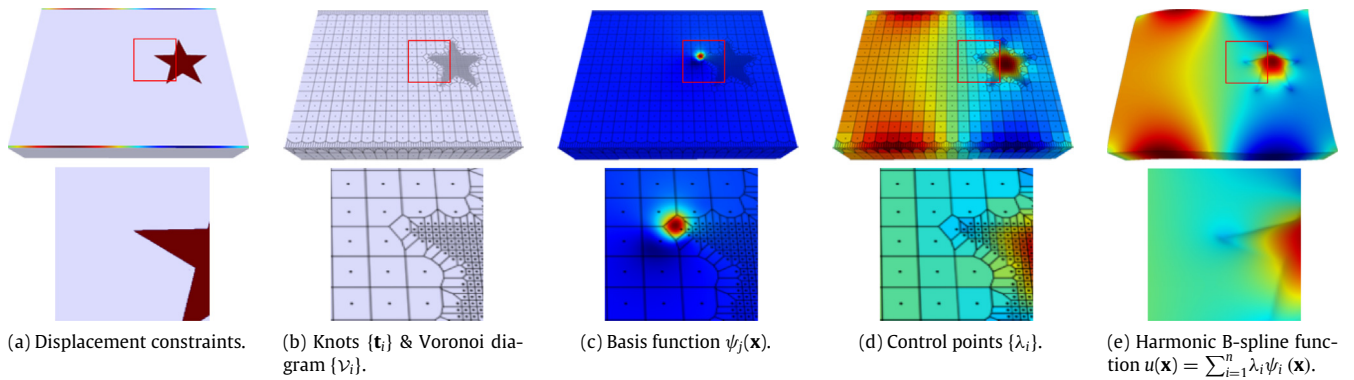
whose discrete form can be expressed as

$$\bar{\phi}_{\mathcal{V}_j}(\mathbf{x}) = \frac{1}{A_{\mathcal{V}_j}} \sum_{\mathbf{y} \in \mathcal{V}_j} A_{\mathbf{y}} \phi_{\mathbf{y}}(\mathbf{x}). \quad (18)$$

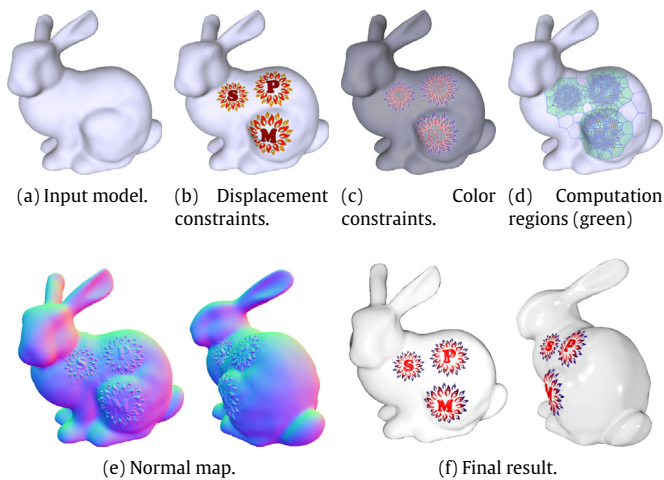
Here  $\phi_{\mathbf{y}}(\mathbf{x})$  is the Green's function on the mesh, which can be obtained by the method in [34]. Note that Green's function on a specific 3D mesh only needs to be calculated once, which is done in the preprocessing step.

Then we can derive the harmonic B-spline function as follows:

$$u(\mathbf{x}) = \sum_{j=1}^n \lambda_j \psi_j(\mathbf{x}) \quad (19)$$



**Fig. 6.** Illustration of our Poisson solver on the box mesh  $\mathcal{M}$ . (a) shows the domain  $U$  (because dDCs are not closed, here  $U = \mathcal{M}$ ) and the displacement constraints including one dPR (red star) and four dDCs (another two on the bottom) along edges of the cuboid model, dPR assigns Laplacian constraints  $f$  and dDCs give the Dirichlet boundary condition  $g$ . (b) To ensure that the Laplacian in each Voronoi cell is a constant, we sample  $n$  knots consisted of all the key vertices involved in the dPR and dDCs and some other vertices evenly distributed with specified knot density. Then we use these interior and boundary knots as generators to construct geodesic Voronoi diagram, which partitions  $U$  into a set of disjoint sub-regions,  $U = \bigcup_{i=1}^n \mathcal{V}_i$ . (c) For a Voronoi cell  $\mathcal{V}_j$ , we define the harmonic B-spline basis function  $\psi_j(\mathbf{x})$ , whose knot  $\mathbf{t}_j$  is the generator of  $\mathcal{V}_j$ . (d) If  $\mathcal{V}_j$  is a boundary Voronoi cell, we simply set its control point  $\lambda_j$  using the given boundary condition  $g$ . For internal Voronoi cell  $\mathcal{V}_j$ , we compute its control point  $\lambda_j$  by solving a sparse linear system, whose size is much smaller than the number of vertices of the mesh. (e) We express the solution by the harmonic B-spline function  $u(\mathbf{x}) = \sum_{i=1}^n \lambda_i \psi_i(\mathbf{x})$ . Since the basis functions are approximately local, we can evaluate  $u(\mathbf{x}) = \sum_{\mathcal{V}_i \in \mathcal{N}(\mathbf{x})} \lambda_i \psi_i(\mathbf{x})$  using only the basis functions close to  $\mathbf{x}$ . (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)



**Fig. 7.** Bunny with the SPM logo. With a few diffusion curves and Poisson regions, we obtain colorful and smooth reliefs. Thanks to the *local* feature of our solver, we only need to construct the geodesic Voronoi diagram and harmonic B-splines for the region of interest. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

where  $u(\mathbf{x})$  is each vertex's color (displacement). These  $\mathbf{x}$  belonging to the knot set which  $u(\mathbf{x})$  are already obtained (by Eqn. (12)) or known (specified by DCs) and there is no need to compute again  $u(\mathbf{x})$ . And  $\lambda_j$  is the control coefficients required from Eqn. (12).

Since the basis function  $\psi(\mathbf{x}, \mathbf{y})$  decays to zero quickly, we can evaluate Eqn. (19) by using the basis of a neighborhood around  $\mathbf{x}$ :

$$u(\mathbf{x}) = \sum_{\mathcal{V}_j \in \mathcal{N}(\mathbf{x})} \lambda_j \psi_j(\mathbf{x}). \quad (20)$$

After solving this harmonic B-spline function Eqn. (20), each vertex's color/displacement can be obtained (Fig. 6(e)). Thanks to the random-access solver, users can decide the region of interest, hereby reducing the computational cost. For instance, applying the visibility algorithm such as back-face culling [39] or Z-buffer [40], we can check whether a point  $\mathbf{x}$  is visible. If not, we do not need to compute  $u(\mathbf{x})$  at all.

## 6. Results

We implemented our algorithm in C++ and evaluated it on a laptop with an Intel i7 CPU 2.80 GHz. Our method allows user to decorate 3D meshes in 3 ways: 1) user sketches PVG primitives (DCs, PCs and PRs) directly on the mesh, and then tunes the Dirichlet boundary condition  $g$  and Laplacian of colors/displacements  $f$ ; 2) user attaches a pre-defined 2D PVG to the region of interest on the 3D model using *local* parameterization; and 3) user tessellates a pre-defined 2D PVG to the entire 3D model using global parameterization.

In Fig. 7, we decorate the Bunny model with the SPM logo and some colorful patterns. Using PVG, mesh decoration becomes easier, since users can sketch arbitrary curves on the 3D model directly and then tune the colors and displacements via setting the associated functions  $f$  and  $g$ . Besides, as the input affecting a small part of the mesh, our Poisson solver can provide a local computation method for the useful regions which lightens the calculation burden a lot in actual situations. As shown Fig. 7(d), based on the input, we can obtain a connected region with open boundary which need to be calculated. After building a geodesic Voronoi diagram with the  $n$  selected knots from that area, we can calculate  $n$  control points of harmonic B-splines function via equation  $\mathbf{L}\mathbf{a} = \mathbf{b}$ . However, only vertices whose Voronoi cells contain input area (the green regions) will be actually calculated their  $u(\mathbf{x})$ .

Fig. 8 shows the result of a stele model decorated with concave Chinese calligraphy characters and convex patterns of auspicious clouds and dragon. First, we use two dDCs to generate a whole concave effect of the middle plaque with the inconsecutive displacement boundaries. Then, we raise the dragon a bit by using the two dDCs. Next, we emboss the middle part of the dragon and the auspicious clouds by dPRs. Finally, we carve the Chinese characters using dPRs.

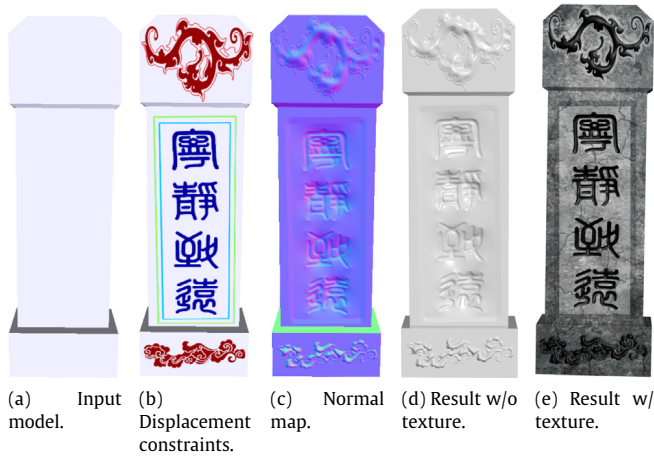
Fig. 9 shows the result of a vase model decorated with blue and white porcelain patterns and smooth petal bumps. On one hand, the vase is decorated with blue and white porcelain patterns in the RGB field. On the other hand, all the petals are smoothly embossed on the mesh.

Considering that the mesh need to be decorated by massive repetitive patterns sometimes, our system also accepts global parameterization as the PVG input to generate rendering result (see

**Table 2**

Statistics. We report the number of diffusion curves and Poisson regions used. The prefixes “c” and “d” distinguish the primitives for color and displacement.

Triangle mesh	#cDC	#cPR	#dDC	#dPR
Bunny (Fig. 7)	82	0	82	81
Stele (Fig. 8)	0	0	22	31
Vase 1 (Fig. 9)	385	0	155	155
Vase 2 (Fig. 9)	264	0	264	264
Rhino (Fig. 10)	0	0	3038	3038
Sofa (Fig. 10)	27144	0	27144	27144
Cactus (Fig. 11)	854	0	122+854	122+854
Tree (Fig. 13)	1 + 119 × 2	1 + 119 × 2	1 + 119 × 2	1 + 119 × 2
Piggy bank (Fig. 12)	17(+3)	66(+15)	17(+3)	17(+3)



**Fig. 8.** Decorating the stele model with concave Chinese calligraphy characters and convex patterns of auspicious clouds and dragon. By simply adjusting the Laplacian values, we can control the height of the concave or convex reliefs. Compared the concave Chinese calligraphy characters and the convex patterns of auspicious clouds and dragon produced by dPR with those created by dDC, it is apparent that our PR could produce smooth details while DC could only produce sharp features. We visualize the displacement constraints by color map in (b), where warm and cold colors indicate positive and negative displacements respectively. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

Furthermore, the combination of sketching and global parameterization can design various gorgeous decoration effects (see Fig. 11). In our current implementation, the bottleneck is the pre-computation stage, including constructing the geodesic Voronoi diagram and computing Green’s functions. For the Piggy Bank model with roughly 40 PVG primitives, pre-computing takes 138.4s (see Fig. 12). Local updating/editing PVG is much more efficient than re-computing for the entire model, since Voronoi diagram can be updated locally and we only need to re-compute the basis functions for the affected Voronoi cells. As shown in Fig. 12, changing colors takes only 0.758s, and editing the geometries of PRs takes 2.889s.

Thanks to the smoothness provided by PRs, our system also allows multi-level sculpting for the meshes (see Fig. 13). Table 2 reports the statistics of the testing models. We observe that PVG can decorate 3D meshes with various effects on the sides of color and geometry by sketching and global parameterization. Particularly, in the field of 3D geometric detail synthesis, compared with the result of the sharp displacement boundaries by only using DCs, our PRs can give a smooth bump effect.

**Discussion.** Jeschke et al. [6] proposed a surface details decoration method with diffusion curves. Due to the Dirichlet boundary

condition applied to diffusion curves, their method can produce only sharp features. In contrast, our method is able to produce both sharp features and smooth displacements, since diffusion curves are a special case of PVG.

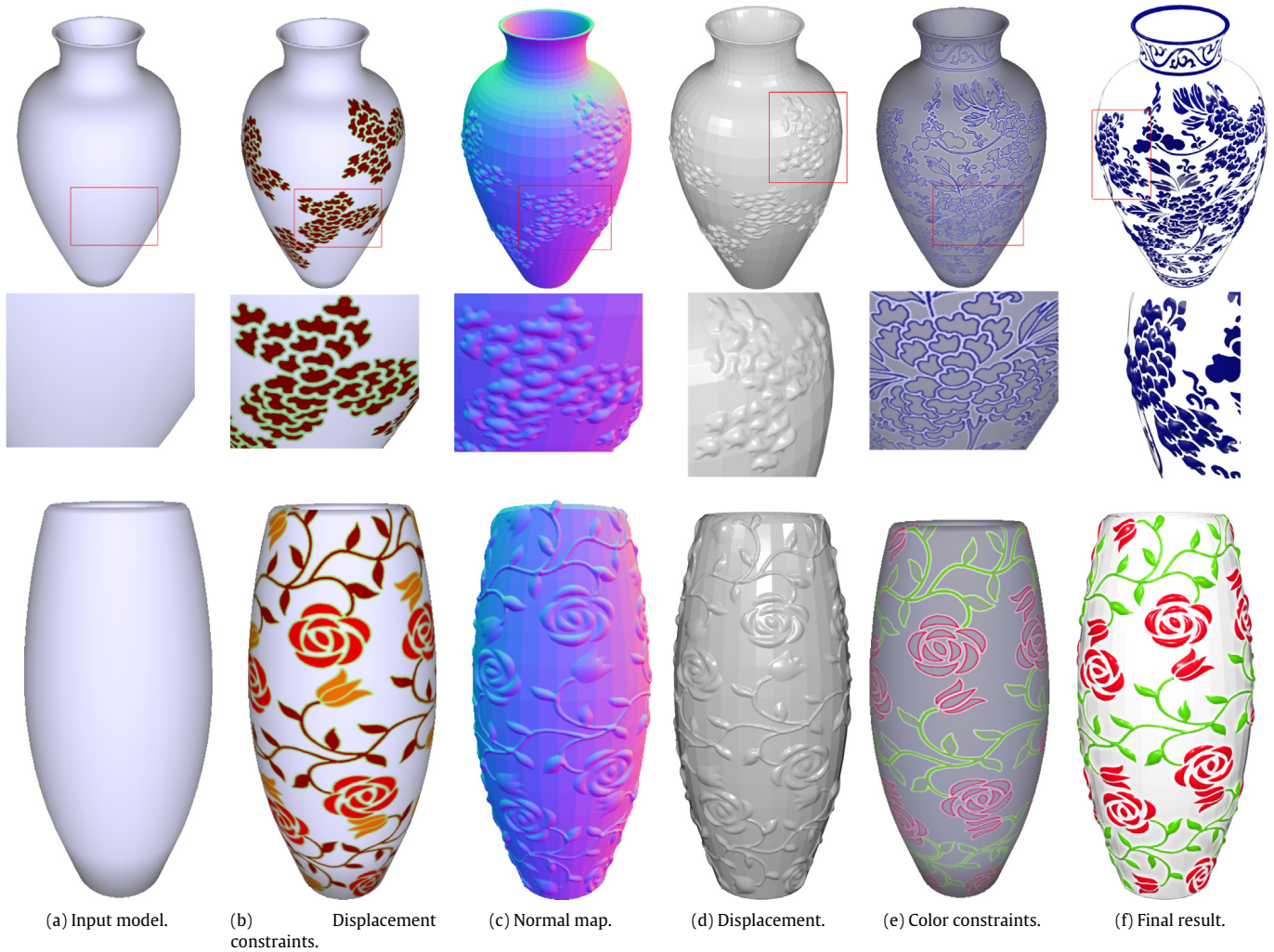
## 7. Conclusion & future work

Poisson vector graphics, a powerful extension to the popular diffusion curves, are equipped with two new types of primitives, namely Poisson curves and Poisson regions. This paper has advocated a new decoration approach that functions directly over meshes of arbitrary topological type within the PVG framework. DCs could be utilized to specify color and height on the mesh, serving as the boundary condition. At the same time, both PCs and PRs could control tone by offsetting Laplacians. By using them in concert, we have detailed a parameterization-free decoration method in this paper, which is affording users to directly work on 3D meshes, by sketching out a set of sparse curves intuitively and conveniently enabled by PVG. At the algorithmic level, to render PVG on 3D meshes efficiently, we have designed an efficient Poisson solver based on harmonic B-spline functions in this paper. First, after sampling the knots, we correspondingly built the geodesic Voronoi diagram on the mesh. Second, we used the Voronoi cells to construct a sparse linear system whose solution provides the control coefficients. Finally, we constructed harmonic B-spline functions and locally synthesized the entire result. With the local computation provided by harmonic B-spline functions, we could flexibly and selectively determine which positions of the mesh model are needed for calculation.

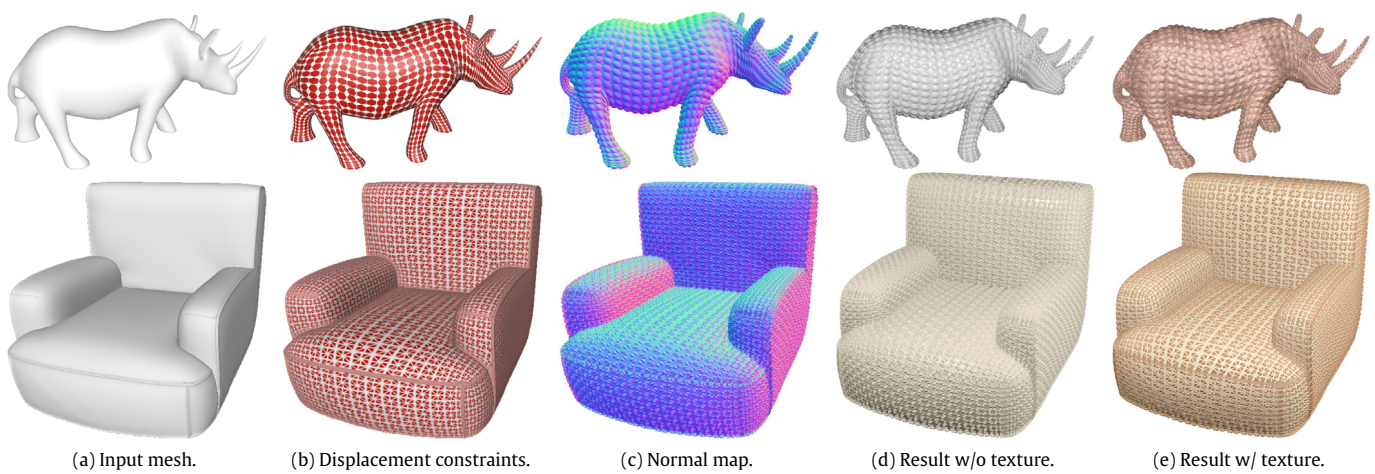
In strong contrast to previous works only supported by the utility of DCs, our new system could achieve the decoration results which are embodied not only with the color shading but also with the smooth embossing and carving visual effects. Additionally, because of the smoothness property directly inherited from PRs, our system also support multi-level editing on meshes. Since our system is solely based on a local solver, our system could significantly reduce the computational burden and provide more versatility such as ignoring calculation of invisible parts of models of current interest.

In our current implementation of the PVG solver, we computed Green’s functions for mesh vertices only and then use linear interpolation to approximate the function value for points inside triangular faces. As a result, we require the input mesh is of relatively good triangulation and high resolution. It is highly desirable to develop a more accurate method to compute Green’s functions for non-vertex points. Also, our paper defines the displacement field using scalar functions. In the future, we plan to extend our method

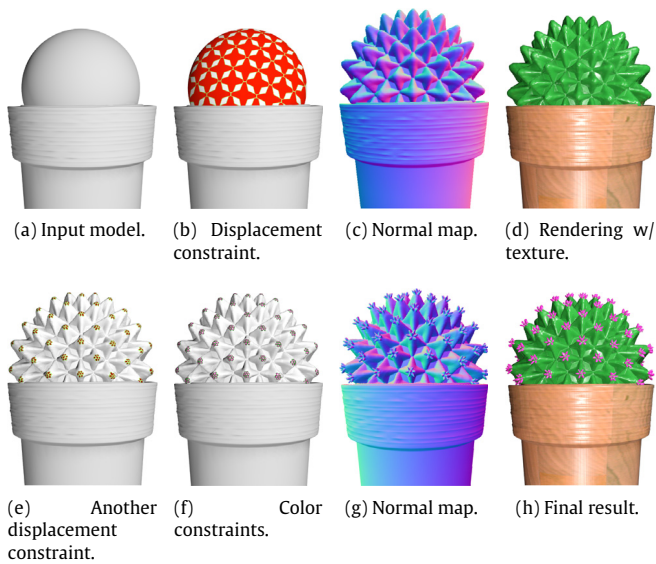




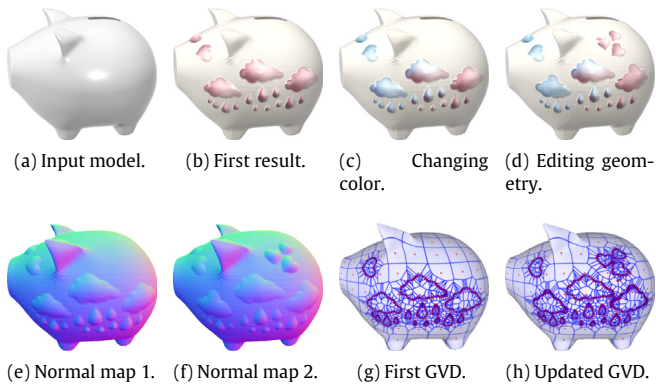
**Fig. 9.** Decorating the vases with smooth reliefs generated by Poisson regions. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)



**Fig. 10.** Decorating the Rhino and Sofa models with repetitive patterns. To simplify the procedure, our system provides global parameterization so that the user-specified pattern can be generated on the entire meshes in a seamless manner. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)



**Fig. 11.** Transforming a sphere into a cactus. We define PVG on 2 separate layers, where the first layer models the stems and the second layer the sharp spines. Since PR is continuous, the two layers can be added together in a seamless manner. Note that it is difficult to use only DCs for LOD modeling, since it is discontinuous along the curves. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

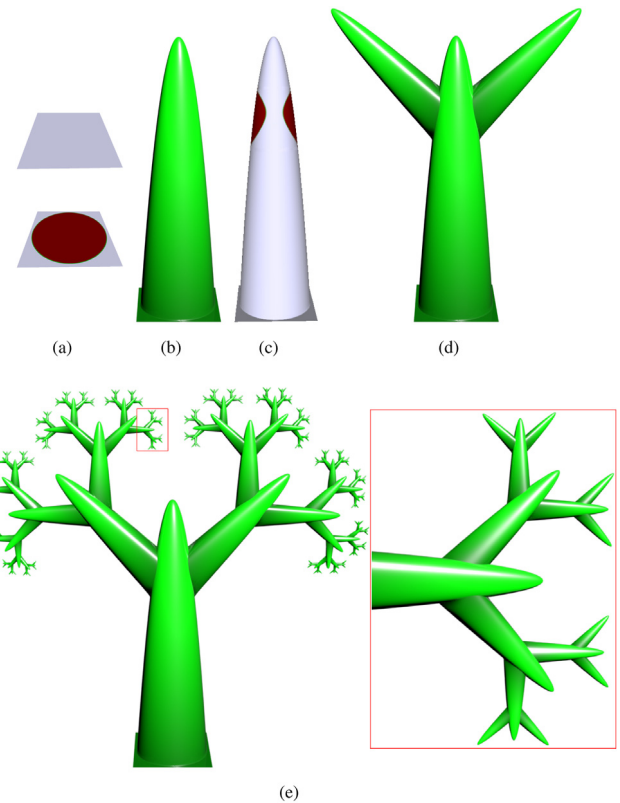


**Fig. 12.** Decorating and editing a piggy bank model. (a) shows the input mesh. (b) shows the first result of decorating with PVGs. (c) is the second result after changing color whose Voronoi diagram is not changed. (d) is the final result after changing DC's shape of two clouds whose Voronoi diagram is locally updated with some additional knots and only the vertices  $\mathbf{x}'$  inside these new Voronoi cells are recalculated by using equation  $u(\mathbf{x}') = \sum_{\mathbf{v}_j \in N(\mathbf{x}')} \lambda_j \psi_j(\mathbf{x}')$ . There are 1,073 knots in (g), and 214 additional knots are added in (h). (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

to vector-valued functions so that we can decorate 3D surfaces with non-height function based displacement.

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**Fig. 13.** Modeling a tree using multilevel PVG. Given a rectangular domain (a), user sketches a circular PR enclosed by a DC whose Dirichlet boundary condition is 0. The resulting PVG is the tree's trunk (b). Then user sketches 2 other sets of DC and PR (c) to define two branches (d). Repeating this procedure a few times, we obtain the tree (e). Similar to other figures, we visualize the values of  $f$  and  $g$  using colors, i.e., green for 0 and red for positive value. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

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