DMESH: FAST DEPTH-IMAGE MESHING AND WARPING

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In this paper we present a novel and efficient depth-image representation and warping technique called DMesh which is based on a piece-wise linear approximation of the depth-image as a textured and simplified triangle mesh. We describe the application of a hierarchical multiresolution triangulation method to generate adaptively triangulated depth-meshes efficiently from reference depth-images, discuss depth-mesh segmentation methods to avoid occlusion artifacts and propose a new hardware accelerated depth-image rendering technique that supports per-pixel weighted blending of multiple depth-images in real-time. Applications of our technique include image-based object representations and the use of depth-images in large scale walk-through visualization systems.

Keywords: Image based rendering; depth-image warping; multiresolution triangulation; level-of-detail; hardware accelerated blending.

1. Introduction

In recent years a new rendering paradigm called Image Based Rendering (IBR), is based on the reuse of image data rather than geometry to synthesize arbitrary views, has attracted growing interest. Since IBR works on sampled image data, and not on geometric scene descriptions, the rendering cost is largely independent of the geometric scene complexity, and depends mainly on the resolution of the sampled data. In fact, one of the goals of IBR is to de-couple 3D rendering cost from geometric scene complexity to achieve better display performance in terms of interactivity. Target applications include interactive rendering of highly complex scenes, display of captured natural environments, and rendering on time-budgets.

There exist many approaches to accelerate rendering for interactive navigation such as reducing the geometric scene complexity using different levels-of-detail (LODs) and exploiting frame-to-frame coherence. Although these approaches achieve significant improvements, they are still limited when the complexity of the
scene has increased far beyond the image-space resolution. In this case, IBR techniques offer advantages over traditional geometric rendering.

In this paper we expand on the technique of depth-image warping.\textsuperscript{5,18} Images with depth per-pixel have been used to represent individual objects\textsuperscript{17,20,31} or to approximate small parts of a large scene in interactive walk-through applications.\textsuperscript{1,2,27,28,30,32} We present an alternative depth-image warping technique, called \textit{DMesh}, based on adaptive triangulation, segmentation and simplification of the depth-buffer, and rendering this \textit{depth-mesh} with the color texture of the input depth-image (see also Ref. 16). In addition to reducing the rendering cost from the geometric scene complexity to the resolution of the depth-images, adaptively triangulating the depth-map further reduces the rendering cost down to the complexity of the depth-variation within this image.

DMesh offers several improvements and alternatives compared to previous depth-image warping techniques. The main contributions include:

- A fast depth-buffer triangulation and simplification technique based on a hierarchical quadtree triangulation algorithm that performs adaptive (and view-dependent if desired) depth-meshing at interactive frame-rates for high-resolution depth-images (i.e. with $512^2$ pixels or more).
- An efficient and simple multiresolution mesh segmentation method.
- Depth-image warping is efficiently performed by rendering a comparatively small and bounded set of textured triangle-strips instead of warping a large number of individual pixels.

Due to its efficiency, our approach is applicable in various rendering systems\textsuperscript{1,2,28,30,32} which represent small parts of a large scene by image-based approximations and that dynamically update the image-based parts frequently at run-time.

The remainder of the paper is organized as follows. In Sec. 2 we briefly review related work in depth-image warping and Sec. 3 describes the basic quadtree triangulation approach used in DMesh. Section 4 describes our depth-meshing method, mesh segmentation is explained in Sec. 5, and Sec. 6 explains the rendering algorithm. In Sec. 7 we provide experimental results supporting our claims and Sec. 8 concludes the paper.

2. Related Work

The notion of \textit{depth} per pixel has been introduced as disparity between images\textsuperscript{5} and used for image synthesis by interpolation between pairs of input images. The depth information — distance from the center of projection along the view direction to the corresponding surface point — allows to re-project pixels from a depth-image to arbitrary new views. A unique evaluation order\textsuperscript{18} guarantees correct back-to-front
drawing order when (forward) warping pixels from the input depth-image to the frame buffer of a new view.

A major problem of depth-image warping techniques are disocclusion artifacts in novel viewpoints that may occur due to regions being visible from the new viewpoint that were not visible from any reference viewpoint. This kind of exposure artifacts cannot be handled solely by algorithmic improvements in IBR methods but must be addressed by additional sample data. In general, if something is not visible in any of the input data sets — reference depth-images — it cannot be reconstructed. Storing multiple depth and color values per pixel has been proposed, also in the layered depth-image (LDI) approach, to cope with such disocclusion artifacts. Such LDIs are used together with an automatic preprocess image placement method to support interactive rendering at guaranteed frame rates, and to represent objects using a bounding box of LDIs. The idea of LDIs has further been extended to LDI-trees for improved control over the sampling rate during image synthesis.

Exposure artifacts from reprojecting a depth-image to an output image with higher resolution, stretching the input samples over a wider output range, is another major problem of depth-image warping. This occurs when zooming in on the depth-image or viewing it from different angles. An efficient approach to cope with this type of exposure artifacts is to represent the depth-image as a triangulated surface. The triangulated depth-buffer provides a connected surface approximating the 3D scene and supports automatic pixel interpolation for exposed or stretched regions due to parallax changes as well as hardware accelerated warping by rendering a textured triangle mesh.

Despite hardware accelerated rendering of textured depth-meshes, the performance of depth-image warping is still determined by the size of the reference images. Therefore, high-resolution depth-images of 512\(^2\) pixels or more impose a strong limiting factor on the rendering speed. To reduce rendering cost, the segmented depth-buffer can be triangulated at a fixed minimum grid resolution. Other approaches go a step further and propose an automatic simplification of the high-resolution depth-mesh. Disocclusion artifacts can be addressed by the use of multiple layers of triangulated depth-images.

In a similar way as our approach, the view-based rendering method uses weighted depth-meshes as well as per-pixel quality factors and normalization to blend multiple reference views together to synthesize new images. In contrast to our DMesh approach, however, the generation of the depth-meshes is performed by manual selection of feature points and a Delaunay triangulation to approximate the depth field. Furthermore, their reference depth-mesh blending algorithm and the per-pixel normalization is performed in software.

As depth-images basically represent colored points in 3D they are related to point based approaches (i.e. Refs. 4, 11, 23–25, 29 and 35) which cope with exposure artifacts due to undersampling and zooming by rendering points as oriented disks, surface elements with non-zero extent. When rendered, the tightly packed surface elements appear to represent a smooth continuous surface.
3. Quadtree Triangulation

In DMesh we use the restricted quadtree triangulation (RQT) method\textsuperscript{21,34} to generate a triangulated surface from the depth-buffer. Figure 1 shows the basic recursive quadtree subdivision and triangulation that introduces vertices from the grid in two steps.

To avoid cracks in the triangulated surface from unrestricted adaptive subdivision and triangulation of the quadtree as shown in Fig. 2, RQT subdivision is constraint such that the levels of adjacent quadtree nodes differ by at most one.

An efficient variation of this constraint is the dependency relation shown in Fig. 3.\textsuperscript{14} This relation specifies for each vertex \( v \) on level \( l \) two other vertices \( v_a, v_b \) on level \( l \) as in Figs. 3(b) and (d), or on level \( l - 1 \) as in Figs. 3(a) and (c), that

\[
\text{level } l-1 \quad \text{level } l \quad \text{level } l \quad \text{level } l+1
\]

\[(a) \quad (b) \quad (c) \quad (d)\]

Fig. 1. Recursive quadtree subdivision and triangulation. Refinement points are shown as grey circles in (b) for a diagonal edge bisection and (c) for vertical and horizontal edge bisections. The bisected edges are denoted by dashed lines.

\[
\text{level } l \quad \text{level } l \quad \text{level } l+1 \quad \text{level } l+1
\]

\[(a) \quad (b) \quad (c) \quad (d)\]

Fig. 2. Cracks (shaded in grey) resulting from an unrestricted quadtree subdivision.

\[
\text{level } l \quad \text{level } l \quad \text{level } l+1 \quad \text{level } l+1
\]

\[(a) \quad (b) \quad (c) \quad (d)\]

Fig. 3. Dependency relation of a RQT. The center vertex (a) depends on the inclusion of two corners of its quad region. The boundary edge midpoints (b) depend on the center vertex. Dependencies within and between the next higher resolution levels are shown in (c) and (d).
must be included in the triangulation such that \( v \) itself can be selected without introducing a crack.

Note that this hierarchical triangulation allows the entire triangle mesh to be represented by one single generalized triangle strip \(^{21}\) which we exploit for rendering efficiency.

4. Depth-Image Meshing

4.1. Overview

The depth values of a depth-image can be considered to be a 2.5-dimensional (projective) height-field with a regular grid structure similar to terrain elevation models. Given the depth values in the \( z \)-buffer for a particular reference image, we can calculate for each pixel \((i, j)\) its corresponding 3D coordinate \( P_{i,j} \in \mathbb{R}^3 \) in the camera coordinate system as illustrated in Fig. 4. Using the coordinates of points \( P_{i,j} \), a quadtree based multiresolution triangulation hierarchy as outlined in Sec. 3 can be constructed on the grid of pixels of the reference depth-image. We call this triangulation of a depth-image a depth-mesh, and the representation of an object by multiple depth-image triangulations a depth-mesh object. In this section we focus on how a single depth-mesh is initialized from a given depth-image, and how adaptively triangulated depth-meshes are generated efficiently.

Figure 5 illustrates the stages to generate the multiresolution depth-mesh data structures. After rendering a reference view, the depth-buffer has to be converted into 3D points given in the camera coordinate system. On the resulting grid of 3D points the quadtree hierarchy is initialized by computing an error-metric and quality measure for each point as described below. Furthermore, per-triangle segmentation is performed as outlined in Sec. 5.

4.2. Initialization

For a multiresolution triangulation using the method outlined in Sec. 3, an error-metric is required to determine the approximation error of a simplified triangle...
As object-space geometric error metric we use a point-to-surface distance. The error of a refinement point is its distance along the elevation axis, the z-axis, to the refined edge, thus a point-to-line distance function. In general, for refinement points \( p \) bisecting an edge between points \( a \) and \( b \) the approximation error is calculated as the 3D point-to-line distance:

\[
d = \frac{|(b - a) \times (b - p)|}{|b - a|}.
\]

Equation (1) is used for refinement points bisecting a diagonal edge as in Fig. 1(b).

For performance reasons, vertical and horizontal refinement points as shown in Fig. 1(c) use an approximative 2D point-to-line distance function instead of Eq. (1). The approximation error \( d \) of a refinement point \( p \), bisecting a vertical (or horizontal) edge between two points \( a \) and \( b \) is calculated as the 2D distance of \( p \) to the line \( \overline{ab} \) in the orthogonal projection onto the \( y, z \)-plane (or \( x, z \)-plane). Figure 6 shows an example configuration of projecting points \( a \), \( b \) and \( p \), from vertically aligned pixels, onto the \( y, z \)-plane. Thus using the line equation \( z = m \cdot y + b \) for \( \overline{ab} \)
projected onto the \( y, z \)-plane, with \( m = (z_b - z_a)/(y_b - y_a) \), the approximation error for vertical refinement points (and analogously for horizontal refinement points) is given by:

\[
d = \frac{m(y_p - y_a) - (z_p - z_a)}{\sqrt{1 + m^2}}.
\]

For field-of-view angles of less than 50°, Eq. (2) introduces an error of less than \( 1 - \cos 25° = 9.4\% \).\(^a\) Note that at most one third of the refinement points are center vertices bisecting a diagonal edge as shown in Fig. 1(b), all others are vertical or horizontal refinement points as in Fig. 1(c). Therefore, because less than 33% of points are diagonal refinement points, we achieve a significant speed-up using the approximate Eq. (2) instead of Eq. (1) for vertical and horizontal refinement points. Note also that working with squared distances \( d^2 \) eliminates the need to compute expensive square root operators in Eqs. (1) and (2).

To achieve a conservative monotonic error metric in the quadtree hierarchy, this distance error metric is maximized in such a way that center vertices (Fig. 1(b)) store the maximal error of all points within that subtree of the quadtree hierarchy.

Additionally for adaptive meshing and rendering purposes we compute and store the following information in the hierarchy. For each depth-mesh vertex \( p \) we store the surface normal \( n_p \) estimated from the depth-buffer, and we calculate a static per-vertex quality measure \( q_p = |n_p \cdot v_p^T| \) with the vector \( v_p \) being the direction from the camera to the point \( p \) in the local camera coordinate system. This quality metric ranks the fidelity of the sampling. Depth-mesh regions perpendicular to the view direction are of highest quality while areas parallel to the view direction are not well sampled. A per-vertex quality measure allows smooth interpolation and blending over depth-mesh triangles in contrast to a per-triangle quality measure.\(^7\)

Note that the use of normals for each depth-mesh point allows interactive re-lighting of the displayed depth-meshes. Given that the reference views were rendered without illumination but only basic color texturing or per-vertex color not modified by any lighting calculations.

### 4.3. Adaptive triangulation

The multiresolution quadtree triangulation hierarchy imposed on the depth-buffer as explained above can be used in different ways to generate adaptively triangulated depth-meshes. The basic algorithm is to perform a recursive top-down traversal of the restricted quadtree triangulation hierarchy\(^{21,34}\) and select points adaptively according to some error tolerance criterion. To guarantee a crack-free triangulation, for each selected point the dependent points are selected as well (see Fig. 3 and Refs. 14, 21). The so selected set of points define a conforming restricted quadtree

\(^a\)Thus this could conservatively be taken into account and added to the result of Eq. (2) by multiplying with 1.1 if desired.
triangulation which can be represented by a single triangle strip. This triangle strip can also be generated in a single top-down traversal.\textsuperscript{21}

For rendering from novel viewpoints, we can distinguish between view-independent as well as view-dependent (Sec. 4.4) reference depth-mesh simplification. A view-independent depth-mesh simplification does not take the novel viewpoint into account for which depth-image warping is performed. We call such a simplification static since the reference depth-mesh has to be simplified only once and then can be warped to any novel view without re-creating it. In this case a reference depth-mesh is simplified according to an object-space geometric approximation error threshold $\varepsilon$. Hence given an error tolerance $\varepsilon$ and since we initialized the error-metric as squared distances, a static simplified depth-mesh is extracted by selecting all points with

$$d^2_p > \varepsilon^2.$$  (3)

Figure 7 shows an example of a triangulated reference depth-image of size $513^2$ pixels viewed from the actual reference viewpoint. While traditional depth-image warping methods have to warp 263,169 individual pixels, DMesh — using a small error tolerance — warps this reference depth-image to any new viewpoint by rendering 32,961 textured triangles (using hardware acceleration). This dramatically demonstrates the power of adaptive depth-mesh simplification.

4.4. View-dependent triangulation

In addition to static depth-mesh simplification, the quadtree triangulation hierarchy also allows to simplify a reference depth-mesh specifically for any given novel viewpoint which the reference view has to be warped too. Such a view-dependent simplification can be achieved by changing the point selection criterion to be dependent not only on the distance error-metric $d_p$ of a depth-mesh point $p$ but also taking into account the novel viewpoint location $e$. We call such a reference depth-mesh simplification dynamic because it has to be performed for every new viewpoint. Thus using an image-space approximation error threshold $\tau$ and given the viewpoint $e$, a depth-mesh point is selected if $f(d_p, p, e) > \tau$. We choose $f$ to
be the geometric error $d_p$ projected onto the image plane of the new view. Thus at rendering time, for each frame an adaptively triangulated depth-mesh can be extracted, given the viewpoint $e$ in the depth-mesh's local camera coordinate system and an image-space error tolerance $\tau$, by selecting all points with

$$\left(\frac{d_p}{|p-e|}\right)^2 > \tau^2.$$  

(4)

As outlined before, the vertex selection is performed recursively top-down in the quadtree hierarchy. In addition to the static simplification, the view-dependent simplification can also take view-frustum culling into account. Hence only points within the view-frustum and projected error exceeding the tolerance are selected. The vertex dependency rules (see Fig. 3 and Refs. 14, 21) are also adhered to and a triangle strip is implicitly generated using the selected vertices.²¹

5. Depth-Mesh Segmentation

5.1. Overview

The triangulation of the $z$-buffer may introduce surface interpolations between different object surfaces and the background as shown in Fig. 8 that cause artificial occlusion or rubber sheet artifacts when warped to new viewpoints. Therefore, it is important to determine whether triangles represent rubber sheets or not.

If not performed in a preprocess, this segmentation of the triangulated depth-mesh has to be done very quickly. Therefore, sophisticated range-image segmentation methods cannot be applied since most of these approaches are far too slow for our purpose.¹² In DMesh, we perform an efficient per-triangle segmentation on the full resolution depth-mesh once during the depth-mesh initialization phase and propagate the results up the multiresolution hierarchy. We will briefly review fast segmentation alternatives such as connectedness and disparity, and then propose the simple orthogonality segmentation test. It is important to note here that any of these three methods can be used efficiently within our proposed multiresolution depth-meshing and warping approach.

Fig. 8. Rubber sheet artifacts showing in (a) are removed in (b) by appropriate segmentation.
5.2. Segmentation techniques

The notion of connectedness\textsuperscript{16} is defined to disambiguate the connected and disconnected mesh regions. Each triangle in the mesh is designated as either low-connectedness or high-connectedness, indicating whether or not the triangle is believed to represent part of a single actual surface. Given a surface normal for each pixel by interpolation, the connectedness between vertical or horizontal pixel neighbors is calculated by comparing the real z-difference with an estimated value from curve fitting.

Figure 9(a) illustrates the basic fitting of a quadratic curve \( f(x) = Ax + Bx + Cx^2 \) using the pixels \( P_1 = (x_1, y_1, z_1) \) and \( P_2 = (x_2, y_2, z_2) \) within a row of the depth-image, and given their corresponding normal vectors \( N_1 = (nx_1, ny_1, nz_1) \) and \( N_2 = (nx_2, ny_2, nz_2) \). Given the reference viewpoint \( e \), the two points and normals are first transformed into the coordinate system with the z-axis aligned with the vector from \( e \) to \( P_1 \) and view-up vector being \((0, 1, 0)\). Using the normals to define the first derivatives \( f'(x) \), the curve parameters can be evaluated to:

\[
A = f(0) = z_1,
B = \frac{\partial f'(0)}{\partial x} = \frac{nx_1}{-nz_1},
C = \frac{\partial^2 f(0)}{2} = \frac{\partial f'(x_2) - \partial f'(0)}{2x_2} = \frac{-nx_2/nz_2 - B}{2x_2}.
\] (5)

We observed that endpoints of stretched edges have almost identical normals — steep rubber sheet triangles — if the normal is estimated from the z-buffer. This makes it easy to fit a quadratic curve according to Eq. (5) such that the points are on, or very close to the curve. To prevent the quadratic curve \( f(x) \) from perfectly fitting vertices of rubber sheet triangles we used a sheared coordinate system as illustrated in Fig. 9(b) with the x-axis parallel to the edge from \( P_1 \) to \( P_2 \). Thus \( x_2 \) changes to \( \hat{x}_2 = |P_2 - P_1| \).

The connectedness is then computed as the ratio of the absolute error in the range estimation to the parameter range \( \hat{x}_2 \):

![Figure 9. Basic quadratic curve fitting in (a), and modified sheared coordinate system in (b).](image-url)
If the ratio of Eq. (6) is greater than some threshold \( \kappa \) then the two pixels are considered to belong to different objects or surfaces. Triangles with such low connectedness \( r > \kappa \) are removed from the depth-mesh. For each right-triangle in our depth-mesh we check the connectedness according to Eq. (6) for the endpoints of the vertical and horizontal edges and if one has low connectedness the triangle is removed. This connectedness is more computing intensive than other methods discussed below due to the fact that normals have to be computed for the points of the depth-image and the complexity of evaluating Eq. (6).

Also changes in disparity values associated with pixels of the depth-image can be used to segment the reference image. Given the distance \( d \) of the image plane from the viewpoint (focal length), the disparity \( \delta \) of a pixel with depth \( z \) is given by \( \delta = d/z \). For three consecutive pixels (in a row or column of the depth-image) with \( z_1, z_2, z_3 \) that are on a line or plane in 3D it holds that \( \delta_2 - \delta_1 = \delta_3 - \delta_2 \). As pointed out in Ref. 19 for curved surfaces, pixels form a piecewise bilinear approximation and are considered to represent a discontinuity if

\[
|\frac{\delta_2 - \delta_1}{z_2} - \frac{\delta_3 - \delta_2}{z_3}| > \sigma
\]

for a given disparity threshold \( \sigma \). In that case the corresponding triangles are removed from the depth-mesh. Using \( \delta_i - \delta_j = \frac{d \cdot (z_j - z_i)}{(z_i z_j)} \), Eq. (7) can be rewritten to:

\[
\frac{d}{z_2} \left| \frac{z_1 - z_2}{z_1} - \frac{z_2 - z_3}{z_3} \right| > \sigma .
\]

As pointed out in Ref. 19 Eq. (7) must vary the threshold parameter \( \sigma \) with the \( z \)-distance of the pixels (but is not further defined). As can be seen from Eq. (8), this adjustment is necessary because of the \( d/z_2 \) factor. In our implementation of this segmentation measure we drop this factor to get a projection-normalized decision inequality:

\[
\left| \frac{z_1 - z_2}{z_1} - \frac{z_2 - z_3}{z_3} \right| > \sigma .
\]

This modified disparity measure of Eq. (9) is an extremely simple measure, it only compares the difference in \( z \)-values between adjacent pixels. In our implementation of disparity, for each right-triangle in our depth-mesh we evaluate Eq. (9) for all three corners and if two or more exceed the disparity threshold \( \sigma \) we remove the corresponding triangle from the depth-mesh.

As an alternative to the connectivity and the disparity measures discussed above we propose an orthogonality test on triangles that is almost as simple to compute as disparity but semantically more powerful. We can observe that the rubber sheet triangles introduced during the triangulation of the depth-image have the following property: the triangle normal is almost perpendicular to the vector
from the viewpoint to the center of the triangle as shown in Fig. 10. Let \( v \) be the vector from viewpoint to the center of triangle \( t \) and \( n_t \) be the normal of \( t \). The following inequality using an angular threshold \( \omega \) can be used to determine if a triangle is a rubber sheet triangle:

\[
\frac{|v|}{|n_t|} \cdot n_t < \cos(90^\circ - \omega) .
\]  

(10)

Compared to the disparity test of Eq. (9) which only compares \( z \)-ranges of two adjacent points, the orthogonality test of Eq. (10) is more powerful since it considers the orientation of a plane in 3D space with respect to the viewpoint.

Note that the proposed orthogonality segmentation criterion (and other segmentation methods as well) may also remove depth-mesh triangles that truly represent object surfaces nearly parallel to the view projection. Even though not intended, this is not considered to be a major drawback since these object surfaces are extremely undersampled in that case. Like other disocclusion artifacts, such strong undersampling is better solved by use of additional sample data, thus using more reference images.

In all segmentation methods, additional care should be taken not to remove very small triangles which represent rough surface features but do not constitute a discontinuity. Therefore, in addition to Eqs. (6), (9) and (10) we also consider the depth-range \( \Delta z \) of triangles at distance \( z \) in the camera coordinate system and we only remove triangles which span a depth-range larger than some threshold \( \lambda \):

\[
\frac{\Delta z}{z} > \lambda .
\]  

(11)

5.3. Multiresolution segmentation

The results of the segmentation process as described in the previous section, if a triangle is considered a rubber sheet, are computed at initialization-time of a reference depth-mesh at full resolution. These segmentation results at the leaf level of the quadtree triangulation hierarchy are then propagated upwards to the parent nodes. The segmentation result are stored as an 8-bit boolean flag \( S \) for the triangles \( a_1, a_2, b_1, b_2, c_1, c_2, d_1, d_2 \) of each quadtree node as shown in Fig. 11(a).
same quadtree node, however, can also be triangulated with fewer triangles as in Fig. 11(b). For these aggregate triangles a, b, c or d the segmentation is derived from the eight sub-triangles. For an aggregate triangle \( t = a \) (as well as b, c or d) the segmentation is given by \( S(t) = S(t_1) \lor S(t_2) \). Similarly, the parent node on level \( l-1 \) determines the segmentation flags from the aggregate triangles of its child node on level \( l \). For example, the segmentation flag of triangle \( a_1 \) on level \( l-1 \) in Fig. 11(c) is set to \( S(a_1_{l-1}) = S(a_2) \lor S(b) \) according to the aggregate triangles a and b in the child node on level \( l \).

Thus the segmentation of one triangle causes all parent triangles in the quadtree hierarchy to be segmented as well. Figure 12 illustrates this behavior of a segmented triangle that causes all parent triangles to be segmented too.

After the initialization of the segmentation at full resolution and the propagation to coarser levels in the multiresolution hierarchy, the segmentation of an adaptively simplified depth-mesh can then be determined extremely efficiently for each quadtree node by a simple boolean expression. This works effectively for both static as well as dynamic view-dependent depth-mesh simplification.

In DMesh, the removal of segmented triangles is directly integrated with the triangle strip generation. As mentioned before, given a set of selected points that specify a conforming restricted quadtree triangulation, a recursive top-down
quadtree traversal \(^{21}\) allows to incrementally generate a single generalized triangle strip,\(^{1}\) see also Fig. 13(a). In DMesh this process is modified to take the segmentation flags of each quadtree node into account. While incrementally creating the triangle strip sequence the segmentation flags of each visited quadtree node are checked, and if necessary the strip is broken up into multiple smaller triangle strips as illustrated in Fig. 13(b). Using this procedure the initial single triangle strip is almost arbitrarily broken into smaller strips. However, as shown in Sec. 7 it still maintains a stripification that is superior to an indexed triangle list in terms of rendering performance.

6. Depth-Mesh Rendering

6.1. Overview

In DMesh, depth-image warping is efficiently performed by hardware supported rendering of textured polygons instead of projecting individual pixels from a reference depth-image to new views. The approximate depth-image consisting of a segmented triangulation of the depth-buffer, as outlined in the previous sections, is rendered using the color values of the reference frame-buffer as texture. Rendering a depth-image with a resolution of \(2^k \times 2^k\) pixels involves warping of \(2^{2k}\) pixels with traditional depth-image warping techniques (or rendering about \(2 \cdot 2^{2k}\) triangles\(^{16}\)). With the proposed technique, instead of warping the \(2^{18} = 262,144\) pixels of a 512 \(\times\) 512 reference image, a textured depth-mesh with only a few thousand textured triangles can be rendered using hardware accelerated 3D graphics at a fraction of the cost of per-pixel depth-image warping.

The depth-mesh initialization, segmentation and adaptive triangulation as outlined in Secs. 4 and 5 is performed in the reference view coordinate system. Whenever a depth-mesh has to be rendered, the coordinate system transformation of that reference view is used as local model-view transformation to place the depth-mesh correctly in the world coordinate system.

To support interactive relighting in depth-mesh rendering, the reference views should initially be captured without any illumination enabled. Thus only basic color

\(^{b}\)Generalized triangle strips may include swap operations.
texturing or per-vertex colors should be enabled when rendering the reference views without applying any modulation from lighting calculations. This way, the color frame-buffer of a reference view contains the “flat” unlit colors. Using these flat colors as texture and using normals at each depth-mesh point, lighting is applied when rendering the depth-meshes for synthesis of new viewpoints.

To reduce disocclusion artifacts, most image warping techniques render multiple reference images that have to be merged to synthesize a new view. We present a novel and highly efficient blending algorithm that exploits graphics hardware acceleration and that supports per-pixel weighted blending of reference depth-images. Blending of $n$ reference depth-meshes to synthesize a new view consists of the following basic steps:

(i) Select $n$ reference depth-meshes $M_i$ ($i = 1 \ldots n$) and textures $T_i$ to be used for the current view, and calculate their positional blending weights $w_i$ with respect to the current viewpoint $e$.

(ii) Adaptively triangulate the depth-meshes $M_i$ for the current viewpoint $e$, and generate the segmented triangle strip representations $S_i$. (If static depth-meshes are used this triangulation is only performed once at depth-mesh initialization, see also Sec. 4.)

(iii) Render the triangle strips $S_i$ without illumination and texturing to initialize the $z$-buffer $Z_e$ for the current viewpoint $e$.

(iv) Render the triangle strips $S_i$ again with their textures $T_i$ and per-vertex quality $\rho$ as alpha values enabled. The result is rendered into separate color-with-alpha frame-buffers $C_i$. Depth-buffer evaluation using $Z_e$ is set to read-only at this stage.

(v) Synthesize the new image $I$ from buffers $C_i$ using positional weights and alpha-blending:

$$ I = \sum_{i=1}^{n} w_i \cdot C_i. $$

(vi) The image $I$ contains the per-pixel weighted result. Note that the final alpha blending factor per pixel may be less than 1.0 at this stage and a normalization of the corresponding color yields the final image.

Figure 14 illustrates the data flow and rendering stages of our algorithm. The following sections explain the different stages in more detail and show how our algorithm exploits hardware acceleration.

Without restricting its generality, the proposed rendering algorithm is tested and explained in more detail below for an example image-based object (IBO) representation but is applicable to other rendering systems that make use of depth-image warping as well. An IBO\textsuperscript{20} is constructed by generating six layered depth images\textsuperscript{c}.

\textsuperscript{c}The per-vertex quality measure $\rho$ will be Gouraud interpolated across triangle faces and yield a per-pixel quality measure in the alpha-channel.
Fig. 14. Depth-mesh rendering and blending stages.

Fig. 15. Selection of reference views is dependent on geometric configuration of the new viewpoint $e$ and its view direction $n$ with respect to the reference view points $r_i$ and directions $v_i$. 
6.2. Depth-mesh selection and triangulation

During rendering, for each frame DMesh selects the most appropriate depth-meshes $M_i$ to be used for the current view. For a depth-mesh object as explained above, we select up to five depth-meshes $M_i$ out of the six reference views that may be visible from the novel rendering viewpoint $e$, see also Fig. 15. A reference view $i$ is selected if the angle $\beta_{v_i,n}$ between the reference view direction $v_i$ and the new view direction $n$ is less than $135^\circ$. Based on this angle, an angular weight

$$w_i(\beta_{v_i,n}) = \begin{cases} 
1.0 & \beta_{v_i,n} < 45^\circ \\
\cos(\beta_{v_i,n} - 45^\circ) & \beta_{v_i,n} \geq 45^\circ
\end{cases}$$

is assigned to the depth-mesh $M_i$. This angular blending weight avoids popping artifacts and guarantees smooth interpolation between contributing depth-meshes when rotating around the image-based object.

Furthermore, an additional positional weight $w_i(d_{r_i,e})$ based on the euclidean distance $d_{r_i,e} = |e - r_i|$ from the reference depth-mesh viewpoint $r_i$ to the novel viewpoint $e$ is associated with depth-mesh $M_i$. The final blending weights $w_i$ are computed as $w_i = w_i(\beta_{v_i,n}) \cdot w_i(d_{r_i,e})$ and are then normalized such that $\sum_i w_i = 1.0$. Given a continuous motion of the viewpoint, the above weight factors also change smoothly and provide smooth blending changes during animation.

For static depth-mesh simplification, the depth-meshes $M_i$ are adaptively triangulated only once at initialization time according to an object-space error threshold $\varepsilon$ (see Sec. 4.3). Otherwise, each selected depth-mesh $M_i$ is simplified view-dependently for the given new viewpoint $e$ according to an image-space geometric error tolerance of $\tau$ pixels (see Sec. 4.4). The adaptively triangulated depth-meshes $M_i$ are segmented and represented by a set of triangle strips $S_i$ which contain vertices with texture coordinates into the reference view color texture $T_i$ and with per vertex quality values $\rho$ (in the RGBA color components).

6.3. Rendering using an $\varepsilon$-z-buffer

To achieve a smooth blending between overlapping depth-meshes $M_i$ the rendering must allow some tolerance in the z-buffer visibility test. Multiple depth-meshes that for a pixel cover the same surface region within some tolerance $\varepsilon$ should be blended together. Only if the z-buffer values are sufficiently different, larger than $\varepsilon$, the front-most depth-mesh determines the final color of that pixel, see also Fig. 16. The blending of overlapping depth-mesh parts does not constitute a simple blurring of available image information. In fact, in the overlap region the depth-meshes capture
Fig. 16. $\varepsilon$-z-buffering of multiple reference depth-meshes.

the same object surface colors. Furthermore, a quality measure is assigned to each point of a depth-mesh as outlined in Sec. 4.2. A weighted blending that takes the relative positions of the reference views as well as the quality of the depth-image samples into account does not represent a blurring but favors the best possible information for image synthesis.

This $\varepsilon$-z-buffer rendering is achieved by first drawing the depth-mesh triangle strips $S_i$ without any shading, illumination or texturing enabled to initialize the z-buffer to $Z_e$ for the desired viewpoint $e$. Since the buffers require to be cleared during the following steps, the stencil buffer is set to store which areas of the frame buffer will be overwritten by the rendered depth-meshes. From here on we will assume that the stencil test is activated to block rendering in areas of the image where the depth-meshes do not cover any part of the new view.

In a second pass and using the previously computed z-buffer $Z_e$ in read-only mode, the meshes $S_i$ are rendered again into individual color buffers $C_i$ with a small negative offset $\varepsilon$ along the view-direction. To initialize the background in the stencil area of the depth-meshes for each buffer $C_i$, a background quad is rendered with its alpha channel set to the corresponding blending weight $w_i$. After this, illumination, shading, alpha-blending, texturing and per-vertex color components are enabled and the $S_i$ are rendered a second time. Since each vertex RGBA color is set to its quality measure ($\rho, \rho, \rho, \rho$) illumination, Gouraud shading and texture modulation with the depth-meshs texture $T_i$ will result in per-pixel weighted colors of the warped depth-mesh in $C_i$. (Technically, at this point the rendered color buffer $C_i$ is copied into texture memory for further processing as outlined in the next section.) If not using any per-vertex quality measure, ($\rho, \rho, \rho, \rho$) can be set to a fixed constant which will result in a positional only weighting. Furthermore, uniform weighting can be achieved by setting all $w_i$ to the same constant value.

At this point every pixel $p$ in the frame buffer $C_i$ with interpolated quality $\overrightarrow{p}$ from lighting, Gouraud shading and texture coordinates $s, t$ will finally store the desired weighted and lit color $(w_i \cdot \overrightarrow{p} \cdot \text{Red}(T_i(s, t)), w_i \cdot \overrightarrow{p} \cdot \text{Green}(T_i(s, t)), w_i \cdot \overrightarrow{p} \cdot \text{Blue}(T_i(s, t)), w_i \cdot \overrightarrow{p})$, or short $(w_i \cdot \overrightarrow{p} \cdot R_p, w_i \cdot \overrightarrow{p} \cdot G_p, w_i \cdot \overrightarrow{p} \cdot B_p, w_i \cdot \overrightarrow{p})$. 
Note that rendering the selected depth-meshes \( S_i \) twice is not very costly as demonstrated by our experiments in Sec. 7 because rendering of textured triangle strips is extremely efficient.

6.4. Blending and normalization

As outlined above, the images \( C_i \) now contain the quality and positional weighted contributions of the selected depth-meshes \( M_i \). The final rendering stages must now perform the image composition and normalization of the color values. The two steps involved are the summation of the weighted colors by \( I = \sum C_i \) followed by normalizing each color component. The image composition operation can be performed efficiently by alpha-blending \( n \) quadrilaterals using \( C_i \) as their texture maps.

The accumulation of images \( C_i \) yields an image \( I \) with intermediate pixel colors \( c' = (\alpha \cdot R, \alpha \cdot G, \alpha \cdot B, \alpha) \), the \( \alpha \) component contains the accumulated blending weight. These color values constitute the proportionally correctly blended values. However, the \( \alpha \) values are not necessarily 1.0 as required. To get the final desired color \( c = (R, G, B, 1.0) \) each color component of \( c' \) has to be multiplied by \( \alpha^{-1} \). This normalization is performed as an image post-process stage on the intermediate blending result \( I \).

Without any hardware extensions to perform complex per-pixel manipulations this normalization step has to be performed in software. However, widely available graphics accelerators now offer per-pixel shading operators that can be used more efficiently. In our current implementation, we perform this normalization in hardware using nVIDIA’s OpenGL Texture Shader\(^{10} \) extension.

To compensate the color intensity deficiency we perform a remapping of the \( R \), \( G \) and \( B \) values based on the value of \( \alpha \). During initialization time we construct a texture encoding in \((s, t)\) of a look-up table of transparency and luminance values respectively, from 0 to 256 possible values. The pixels of this texture encode the new intensity for a given luminance \( t \) compensated with the \( \alpha \) transparency \( s \).

Based on this alpha-luminance map, we proceed to correct each of the \( R \), \( G \) and \( B \) channels of every pixel. Using nVIDIA’s texture shader extension operation GL\_DEPENDENT\_AR\_TEXTURE\_2D\_NV the graphics hardware can remap the \( R \) and \( \alpha \) by a texture lookup with coordinates \( s = a \) and \( t = R \) into our alpha-luminance map. At this point, rendering a quadrilateral with the intermediate image \( I \) as texture-one and the alpha-luminance map as texture-two, and setting the color mask to block the \( G \), \( B \) and \( \alpha \) channels will compensate the red channel by \( \alpha^{-1} \) and store it in a new image \( I_R \). Note that only the \( R \) and \( \alpha \) channels are used by this dependent texture replace operation. Therefore, we need to remap the \( G \) and \( B \) channels to the \( R \) channel of two additional buffers \( I_G \) and \( I_B \) while copying the \( \alpha \) channel as well. This is done by rendering two quads and using nVIDIA’s register combiners. Then the dependent texture replace operation is also performed.
on the images $I_G$ and $I_B$. Thus by separating the $RGB$ channels into three different images and using the $\alpha R$-dependent texture replace operation we get the corrected $RGB$ values (in the red channel of three new images $I_R$, $I_G$ and $I_B$). Figure 17 illustrates this normalization process. Finally the three separated and corrected color channels are composited into the final image again using nVIDIA’s register combiners by rendering three quads with textures $I_R$, $I_G$ and $I_B$.

7. Experiments

In this section we report experimental results of our depth-meshing and warping algorithm. As an example to demonstrate the approach we use an image-based object (IBO) which consists of a bounding box of six depth-meshes around the object as outlined in Sec. 6. All experiments were performed on a a Dell 1.5GHz Pentium4 with nVIDIA GeForce4 4600Ti graphics card.

Using per-vertex depth-mesh normals and unlit captured reference images we perform full interactive relighting with diffuse as well as specular components when rendering the depth-meshes. This real-time lighting calculation is included in all frame rate timings reported in this section.

7.1. Depth-mesh initialization

Table 1 shows timing results for the generation of depth-meshes with our approach given a depth-image size of $513^2 = 263,169$ pixels. The table shows the size of the test models, the time it takes to render the geometry, the cost to capture
Table 1. Depth-mesh construction cost for different polygonal models. Given is the model size, geometric rendering cost, z-buffer capture and conversion, and multiresolution model construction.

<table>
<thead>
<tr>
<th>Model</th>
<th>Triangles</th>
<th>Render</th>
<th>z-buffer</th>
<th>Quadtree</th>
</tr>
</thead>
<tbody>
<tr>
<td>David</td>
<td>8,254,150</td>
<td>5862 ms</td>
<td>38 ms</td>
<td>641 ms</td>
</tr>
<tr>
<td>DHead</td>
<td>4,000,885</td>
<td>3731 ms</td>
<td>40 ms</td>
<td>660 ms</td>
</tr>
<tr>
<td>Dragon</td>
<td>871,414</td>
<td>634 ms</td>
<td>40 ms</td>
<td>650 ms</td>
</tr>
<tr>
<td>Female</td>
<td>605,086</td>
<td>688 ms</td>
<td>37 ms</td>
<td>654 ms</td>
</tr>
</tbody>
</table>

and convert the z-buffers into 3D points, and the cost to initialize the six depth-meshes \( M_i \) around the object. This initialization includes the calculation of the error metric, per-vertex quality measure, triangle segmentation flags, and other multiresolution parameters as described in Sec. 4. In the case of static depth-meshes the initialization time will also include the adaptive vertex selection and generation of triangle strips \( S_i \) for a given object-space error tolerance \( \varepsilon \).

Not shown due to negligible cost is the capturing of the color textures \( T_i \) corresponding to particular depth-meshes \( M_i \). The segmentation angle tolerance \( \omega \) was set to 5° and the depth-range threshold \( \lambda \) to 100 of the objects maximal extent. As expected, one can observe that the depth-mesh initialization is constant for a given reference image resolution. The quadtree hierarchy construction grows with \( O(n \cdot \log n) \) for a reference depth-image of \( n \) pixels.

As can be seen from Table 1, the initialization of a depth-mesh structure is very efficient. In this case the timing includes generating six depth-meshes around an object. The timings show that the proposed depth-meshes can indeed be generated quickly enough to be used in rendering systems\(^{1,2,28,30,32}\) where they have to be updated dynamically at run-time. The major costs are rendering the actual geometry (not really part of the depth-mesh generation) and the construction of the depth-mesh quadtree triangulation data structure. For a single 513\(^2\) depth-image the initialization takes about 150 ms.

We also performed several tests using the different segmentation methods outlined in Sec. 5.\(^{22}\) In this paper we only want to summarize these results as the choice of segmentation method is independent of the algorithms and data structures of DMesh. The connectedness measure is by far the most expensive of the three approaches, while disparity and orthogonality are close in time cost with disparity being the fastest method of all. As can be seen in Fig. 18, despite its processing cost the connectedness based segmentation is not very good and the disparity or orthogonality tests provide much better segmentation at lower cost.

7.2. View-independent depth-mesh rendering

In Table 2 we report the rendering performance of DMesh using static, view-independently simplified depth-meshes. We used an object-space error tolerance \( \varepsilon \) of 1/1000 of the length of the bounding box diagonal. Selection cost and weight calculation of reference views was omitted due to negligible cost. Adaptive triangulation
Fig. 18. Segmentation results using connectedness, disparity and orthogonality test (image pairs from left to right), displayed from the actual reference viewpoint as well as slightly rotated.

Table 2. Static DMesh rendering performance of image based objects. It shows the total number of triangles rendered for the selected depth-mesh triangle strips $S_i$ and the timing of the different rendering stages.

<table>
<thead>
<tr>
<th>Model</th>
<th>Geometry</th>
<th>DMesh</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Time FPS</td>
<td>Triangles in $S_i$</td>
</tr>
<tr>
<td>David</td>
<td>4247 ms 0.24</td>
<td>185,173</td>
</tr>
<tr>
<td>DHead</td>
<td>5595 ms 0.18</td>
<td>382,356</td>
</tr>
<tr>
<td>Dragon</td>
<td>633 ms 1.6</td>
<td>176,136</td>
</tr>
<tr>
<td>Female</td>
<td>1029 ms 0.97</td>
<td>335,050</td>
</tr>
</tbody>
</table>

of the reference depth-meshes (Step 2 from Sec. 6.1) has been performed once at initialization of the reference views and is not included in these timings. The cost of generating the $z$-buffer $Z_e$ includes rendering the segmented triangle strips $S_i$ of all selected depth-meshes once (Step 3 from Sec. 6.1). The rendering of images $C_i$ corresponds to Step 4, and blending includes Steps 5 and 6 from Sec. 6.1.

As one can see from Table 2, the static DMesh depth-meshing and warping algorithm provides stable and very high frame rates for arbitrary complex objects, and that the rendering performance is completely independent from the size of the input polygonal model. The total depth-mesh rendering and blending cost is in the order of 5 to 15 ms. The weighted blending of the individual warped depth-images $C_i$ is extremely efficient and scales to arbitrary number of input images $C_i$ as long as the pixel fill rate of the graphics hardware is not exceeded. The average length of triangle strips for these depth-meshes range from 20 to 40 triangles per strip.

Images of static DMesh rendering are shown in Fig. 19, comparing the geometric rendering (left) to a synthesized view from multiple reference depth-meshes (right). Each color in the occupancy images (bottom row) illustrates the use of different depth-meshes by varying colors and shows any combination of two or more overlapping depth-meshes by coloring this area orange-brown.

7.3. View-dependent depth-mesh rendering

Table 3 shows the rendering performance of view-dependent depth-mesh image based objects, and corresponding screenshots are given in Fig. 20. We used an image-space error tolerance $\tau$ of 1 pixel. The selection of the reference views has negligible costs and is omitted here. The synthesis cost of generating the $z$-buffer $Z_e$
DMesh: Fast Depth-Image Meshing and Warping

Fig. 19. Static DMesh rendering of complex polygonal objects. Left image shows the actual polygonal object and the right image shows the weighted blending of multiple textured depth-meshes. Various blending combinations of different depth-meshes are shown on the far right by different shading (pseudo colors), darkest shade (orange-brown) denotes two or more overlapping depth-meshes.

Fig. 20. View-dependent DMesh rendering of complex polygonal objects. Left image shows the actual polygonal object and the right image shows the weighted blending of multiple textured depth-meshes. Various blending combinations of different depth-meshes are shown on the far right by different shading (pseudo colors), darkest shade (orange-brown) denotes two or more overlapping depth-meshes.

Table 3. View-dependent DMesh rendering performance of image based objects. It shows the total number of triangles rendered for the selected depth mesh triangle strips $S_i$ and the timing of the different rendering stages.

<table>
<thead>
<tr>
<th>Model</th>
<th>Geometry</th>
<th>DMesh</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Time</td>
<td>EPS</td>
</tr>
<tr>
<td>David</td>
<td>4268 ms</td>
<td>0.23</td>
</tr>
<tr>
<td>DHead</td>
<td>5548 ms</td>
<td>0.18</td>
</tr>
<tr>
<td>Dragon</td>
<td>639 ms</td>
<td>1.5</td>
</tr>
<tr>
<td>Female</td>
<td>1028 ms</td>
<td>0.97</td>
</tr>
</tbody>
</table>

includes the view-dependent selection of vertices from the depth-meshes $M_i$, as well as rendering the segmented triangle strips $S_i$ once (Steps 2 and 3 from Sec. 6.1). The rendering of images $C_i$ corresponds to Step 4, and blending includes Steps 5 and 6 from Sec. 6.1.

It can be observed from Table 3 that the view-dependent DMesh rendering method also provides interactive frame rates for very complex objects. The total depth-mesh rendering and blending cost is in the order of 100 to 200 ms. The average length of triangle strips in $S_i$ for these depth-meshes range from 15 to 30 triangles per strip. Compared to the static depth-meshes, the view-dependent ones provide similar quality renderings at lower triangle counts. However, since
the view-dependent depth-mesh simplification was performed for each frame in the experiments the overall rendering time is much slower than for static depth-meshes. This is expected since the view-dependent mesh simplification and generation of triangle strips is an expensive task.

Note however, that infrequent updates to the view-dependent depth-mesh simplification as performed in IBR-based rendering systems\textsuperscript{1,2,28,30,32} may still be handled efficiently enough if not performed for each individual frame. Once a view-dependent depth-mesh is selected its rendering is about as efficient as static depth-mesh rendering.

Figure 20 shows images of view-dependent DMesh rendering, comparing the geometric rendering (left) to a synthesized rendering from multiple reference depth-meshes (right). Each color in the occupancy images (bottom row) illustrates the use of different depth-meshes by varying colors and shows any combination of two or more overlapping ones as orange-brown.

7.4. Other experiments

In Fig. 21 we compare static and view-dependent depth-mesh simplification for far, medium and near range rendering. The wire frame snapshots only show one out of the up to five rendered depth-meshes. Table 4 presents the experimental results, showing the varying number of rendered triangles for view-dependent (dynamic) depth-meshes. While the static depth-meshing approach is always faster...
Table 4. Comparison between static and dynamic (view-dependent) depth-mesh simplification and rendering.

<table>
<thead>
<tr>
<th>Range</th>
<th># Triangles</th>
<th>Rendering time</th>
<th>FPS</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Static</td>
<td>Dynamic</td>
<td>Static</td>
</tr>
<tr>
<td>Far</td>
<td>209,124</td>
<td>54,925</td>
<td>6.55 ms</td>
</tr>
<tr>
<td>Medium</td>
<td>209,124</td>
<td>99,474</td>
<td>6.29 ms</td>
</tr>
<tr>
<td>Near</td>
<td>209,124</td>
<td>75,258</td>
<td>6.23 ms</td>
</tr>
</tbody>
</table>

Fig. 22. Smooth rotation around depth-mesh object exhibiting strong changes in the weighted blending factors of the different depth-meshes.

in rendering performance, the view-dependent alternative may provide better quality depth-meshes for close-ups and may also exhibit less aliasing artifacts at far distances because the number of rendered triangles is adapted to the projection size on screen. One can also observe that depth-image warping speed using our DMesh technique is not restricted by pixel-fill rate limits of the video hardware. Independent of the screen-projection size the rendering performance is dominated by the complexity of the depth-meshes, and the adaptive triangulation in the view-dependent version.

Figure 22 illustrates the smoothness of our positional weighted blending approach. Despite strong changes in the weighted blending factors of the different contributing depth-meshes, the continuous rotation of the depth-mesh object does not exhibit any disturbing popping artifacts in the blending results.

8. Discussion

In this paper, we presented DMesh, an efficient depth-image meshing and segmentation method as well as a novel real-time depth-image blending algorithm that exploits hardware graphics acceleration. The proposed depth-buffer triangulation approach significantly reduces the complexity of depth-image warping by adaptive simplification of the triangulated depth-mesh, and by rendering textured triangle strips. Our novel blending technique provides real-time per-pixel weighted blending of multiple depth-images.
Our approach improves over previous depth-image warping methods\textsuperscript{6,7,16,18,33} in particular in the following aspects: fast generation of depth-mesh based object representations, and per-pixel weighted blending of multiple reference views at interactive frame rates. The presented approach can be used as a rendering component in visualization systems such as large scale walk-through applications\textsuperscript{1,2,28,30,32}.

The main limitation of the presented approach lies within the simple (but fast) segmentation of rubber-sheet triangles. The current approach is to provide a solution for use in interactive rendering systems where initialization and segmentation cost of depth-meshes is critical, but in specific circumstances may lead to over-conservative removal of triangles. At the expense of CPU cost, more sophisticated segmentation methods could be used (see for example Ref. 12) within the same framework.

Another conceived limitation of DMesh is its inability to cover disocclusion artifacts due to scene areas not visible in any reference depth-image. As already pointed out in Sec. 2, handling this type of exposure artifacts requires additional sample data and cannot be addressed algorithmically. DMesh does not introduce any restrictions handling the given input depth-images and hence does not de facto suffer from this limitation. For example, to alleviate disocclusion problems more sample data can be added by more depth-images. However, there are more intelligent approaches to add sample data such as layered approaches\textsuperscript{9,13,17,31}. Thus extending DMesh to include multiple layers and sparsely populated depth-images is an important area of future work.

Further future work will include application of our blending algorithm to other IBR and point-rendering methods, and taking advantage of per-pixel and per-vertex programming features of modern graphics hardware accelerators.

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References


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