Scalable Rendering of variable Density Point Cloud Data

Priyadarshini Kumari†
Department of Electrical Engineering,
IIT Bombay

Sreeni K.G‡
College of Engineering,
Thiruvananthapuram

Subhasis Chaudhuri‡
Department of Electrical Engineering,
IIT Bombay

ABSTRACT

In this paper, we present a novel proxy based method of adaptive haptic rendering of a variable density 3D point cloud data at different levels of detail without pre-computing the mesh structure. We also incorporate features like rotation, translation and friction to provide a better realistic experience to the user. A proxy based rendering technique is used to avoid the pop-through problem while rendering thin parts of the object. Instead of a point proxy, a spherical proxy of variable radius is used which avoids the sinking of proxy during the haptic interaction of sparse data. The radius of the proxy is adaptively varied depending upon the local density of the point data using kernel bandwidth estimation. During the interaction, the proxy moves in small steps tangentially over the point cloud such that the new position always minimizes the distance between the proxy and the haptic interaction point (HIP). The raw point cloud data re-sampled in a regular 3D lattice of voxels are loaded to the haptic space after proper smoothing to avoid aliasing effects. The rendering technique is experimented with several subjects and it is observed that this functionality supplements the user’s experience by allowing the user to interact with an object at multiple resolutions.

Index Terms: Point cloud rendering, levels of detail, bounding surface, adaptive rendering, kernel bandwidth [1]:

1 INTRODUCTION

Haptic and visual rendering require the object geometry to be well defined. Generally, objects are represented by polygonal or triangular meshes as it makes further processing of rendering very simple. But these meshes have to be constructed and are not directly available. When an object with very fine details are represented by a triangular mesh, the number of small facets required increases drastically. The same object when displayed graphically as a whole, some facets may occupy less than one pixel, which results in a wastage of memory as well as the computation time. Hence when a highly detailed object is rendered haptically, the computational time can be reduced by reducing the number of small facets without affecting the haptic experience. The scaling feature allows us to display the object using a fewer number of facets when it is less significant, i.e. the object needs to be experienced at a gross level and by more number of facets when it is to be experienced at finer details. With a mesh based representation, creating different levels of detail requires the mesh to be re-sampled and re-meshed. Re-sampling and re-meshing during the interaction is not possible for a dense data as it may not meet the required proxy update frequency for rendering haptically. For graphic rendering the ideal frame rate should be 30 frames per second, while to perceive the object haptically, proxy update frequency should be at least 1 KHz. Many of such issues can be resolved, if the object is represented as a point cloud. In the proposed method the average proxy update time is 0.03 ms which is 30 times faster than the required proxy update frequency. Also there are devices like laser scanners, RGB-D cameras (eg. Microsoft Kinect) etc. that can directly create the point cloud data from a real object. Therefore the ease of data availability also makes rendering with point cloud preferable. A point cloud rendering algorithm uses the data directly and avoids the intermediate step of mesh construction. However the key issue is how to handle the proxy sinking problem in case of variable density point cloud. With a fixed radius, the proxy may fall through the point cloud of highly sparse region. If we use a proxy of sufficiently large radius, sinking problem can be solved but then the finer details cannot be experienced. In order to address this problem, the radius of a spherical proxy is adaptively changed by statistical bandwidth estimation of a Gaussian kernel.

In order to feel the 3D dense model at a closer level at run time we need to create hierarchies of levels of detail. With the Monge surface representation, such an hierarchy of levels of detail need to be pre-computed as given in [14]. Authors have used a point proxy which is moved iteratively over the locally computed surface to minimize the distance to the HIP during collision. Using this algorithm an user cannot rotate and translate the 3D object to render the rear part of object as depth data is generated here by using OpenGL depth buffer. This algorithm also does not incorporate the surface property like friction, which deters the realistic experience of user. In the proposed method a full 3D object is used instead of a Monge surface, where \( z = f(x, y) \) and hierarchies of levels of detail are computed during haptic interaction time itself. There are very few haptic rendering algorithms for point cloud data. However, currently available algorithms cannot deal with scalability without constructing a surface or mesh of the whole point cloud. The Proxy based rendering is difficult with a 3D point cloud data as the standard bounding surface is not defined. So we propose to use a sphere of sufficient radius as a proxy to avoid the sinking in the point cloud data. To the best of the authors’ knowledge, there exists no method for rendering a variable density point cloud at run time when it is being scaled, rotated and translated. Our proposed work facilitates the user with all these features and also incorporates surface friction.

Depending upon the user’s interest the point cloud can be transformed (scaled, rotated or translated) prior to sampling. In our method, we re-sample the point cloud data in a regular 3D lattice of voxels and load them into the haptic space. In order to avoid the aliasing we use a mean filtering before sampling the 3D lattice. Since we do not have lattice structure of data, we cannot use other filters like Gaussian filter for smoothing. In order to scale, rotate and translate the 3D model, the data points are multiplied by a suitable transformation matrix and then mapping is done to the 3D lattice. The mapped 3D points decide the actual object geometry for haptic rendering.

In real world every object gives the sensation of a resistive force while touching it. Without incorporation of the surface friction in the algorithm [12], rendering gives a very slippery experience. In the proposed work, we simulate the surface friction to provide a more realistic experience to the user. Generally friction is simulated based on the local texture of the surface. Since we are rendering
point cloud data, we do not have any texture information. Therefore to provide the feeling of friction, we simulate it by slowing down the movement of proxy.

The paper is organized as follows. Section 2 gives an overview of relevant literature. In section 3 we describe the proposed method of rendering a variable density point cloud data. This section also includes the key contributions of the paper, which describes how to handle scale change for zooming in and out, rotation and translation. Section 4 explains the force and friction computation during the haptic interaction. Results are presented in section 5. We draw the conclusions in section 6.

2 LITERATURE REVIEW

In haptic rendering literature there are very few approaches with point cloud data. The common practice is to first create the surface or mesh from point data, and then render the mesh haptically. A good description of the basic haptic rendering using such a mesh is provided in [10]. In order to render the polygon based model, mainly two operations are involved: haptic collision detection and haptic force computation. There are several algorithms to detect the collision with high density points [3], implicit functions [4], and polygonal mesh models [6]. The most commonly used method is based on the polygon meshes which requires the calculation of the penetration depth of haptic interaction point(HIP) into the surface. If \( D \) is the penetration depth in the model, the reaction force can be computed as \( f_r = -KD \), where \( K \) is the stiffness constant of the surface. Nevertheless, haptic rendering with polygon based model suffers from the problems such as dependency on the primitive level inputs like vertices and surface normal. The second problem associated with this rendering technique is ambiguity in determining the appropriate direction of the force while rendering thin objects. These problems have been solved by Zilles and Salisbury [14], and Ruspini et al. [9] by introducing the concept of god-object and the proxy point, respectively. The problem with the mesh based representation is that when the object is not fully enclosed by the bounding planes, a small hole may remain and therefore the proxy sinks during the rendering process. Mesh based rendering is not amenable to render the object at different levels of detail as with a large number of polygons it requires to pre-compute the mesh at each level.

Another alternative rendering technique is the voxel based haptic rendering [13, 6]. In these methods, properties of the object like stiffness, color, density and distance field are sampled and stored in the 3D grid. The major drawback of this rendering technique is that we need to have a pre-computed distance field and computing distance field of dense model is itself a very complex and time-consuming process. This technique also suffers from the thin object problem. As regards rendering of a point cloud data, Leeper et al. discuss a constraint based approach of rendering where the points are replaced by spheres or surface patches of approximate size [1]. Another algorithm is described in paper [2], where author uses axis aligned bounding cube to fill the voids between the points and then use god object rendering technique. Haptic rendering with point cloud data has also been proposed by Lee et al. in [5], which creates the moving least square surface from the point cloud data and then minimizes the distance between proxy and HIP. As this algorithm does not keep track of HIP penetration, it suffers from the thin object problem as well as creating a moving least square surface increases the complexity of the algorithm. We try to overcome these limitations by introducing a novel approach as explained in section 3.

3 PROPOSED METHOD

For better understanding, let us first assume that the point density of the object is uniform and this assumption will be relaxed later. In order to haptically render the point cloud data we need to find out the penetration depth of HIP into the surface and the surface normal to apply force feedback. By taking these requirements into consideration, our algorithm tries to move the proxy on the surface in small steps in the tangential direction during the interaction and the new position always minimizes the distance between HIP and proxy which is the measurement of the amount of force to be fed back.

The proposed rendering technique is in some sense similar to the technique given in [11]. For better understanding, let us consider a point data in 2D as shown in Fig. 1. The small red circle represents the HIP and the big black circle represents the proxy. In free space HIP and proxy move together and are collocated as shown in Fig. 1a. Let HIP and proxy touch the boundary at point \( P \) as shown in Fig. 1b. After this contact, HIP continues to move through the point data, and the proxy stays over the surface. In Fig. 1c HIP has penetrated the surface and is inside the point cloud data and proxy remains over the surface. Once the HIP penetrates the point cloud data, reaction force need to be fed back to the user. In order to calculate the reaction force, penetration depth and direction of normal need to be computed. We calculate the penetration depth from the nearest point on the object boundary. We find the most appropriate position of proxy through our algorithm at \( R \) as shown in Fig. 1d. At point \( R \) the vector joining the HIP and proxy is oriented opposite to the normal vector at \( R \), hence the distance between HIP and subsequent proxy position is minimum at this point. The normal at the proxy location \( P \) is shown with the vector \( \mathbf{v}_n \) and the vector \( \mathbf{v}_p \) is the vector to HIP from proxy. If \( \mathbf{C}_p \) denotes the proxy center and \( \mathbf{H}_p \) denotes the HIP point, the vector \( \mathbf{v}_p \) is given by \( \mathbf{v}_p = \mathbf{H}_p - \mathbf{C}_p \). We move the proxy along the direction tangential to the surface and lying in the plane found by the vectors \( \mathbf{v}_n \) and \( \mathbf{v}_p \) to find the most appropriate location \( R \) during the collision. The following equation is used to evaluate the tangent vector \( \mathbf{v}_t \) at every step.

\[
\mathbf{v}_t = \mathbf{v}_n - (\mathbf{v}_n \cdot \mathbf{n}) \mathbf{n},
\]

where \( \mathbf{v}_n \) and \( \mathbf{n} \) denote the surface normal and unit surface normal at the point of contact of proxy with the surface. With point cloud data we do not have other information about underlying objects such as surface normal or bounding surface. In our work we estimate the normal at the proxy location by using the points inside the spherical proxy as explained below.

In free space, there are no data points inside the spherical proxy as shown in Fig. 1a. If points are present inside the proxy, we need to restrict the further movement of proxy into the object. Once the proxy touches the point cloud data, a function, \( \mathbf{v}_n \) is computed.
dynamically which measures the sinking of proxy into the object \([11]\).

\[
v_n = \sum_{i=1}^{N} d_i. \tag{2}\]

The function \(v_n\) can be computed from the radial overshoot of each point as given in equation \([2]\). For each point \(p_i\), overshoot \(d_i\) can be evaluated using equation \([3]\) where \(r_p\) and \(C_p\) are the radius and the center of the proxy, respectively.

\[
d_i = \left( r_p - |C_p - p_i| \right) \left( C_p - p_i \right) \left| C_p - p_i \right|^{-1}. \tag{3}\]

The unit surface normal as used in equation \([1]\) at the proxy point during the interaction is given by

\[
n = v_n / |v_n|. \tag{4}\]

The vector \(n\) is estimated at each iteration and it gives the direction along which proxy should be moved to reduce sinking.

### 3.1 Proxy Update Algorithm

The proposed algorithm can be illustrated using the following steps.

1. Let us consider the situation when HIP is outside the object. The moment proxy starts sinking into the point cloud, the proxy may enclose some points and \(v_n\) becomes non-zero. To avoid the proxy from further sinking, we move the proxy by an amount of \(k_n\) in the normal direction \(v_n\). If \(\left( |v_n| > \zeta \right)\) where \(\zeta\) is a small non-zero value then the proxy position is updated according to the following equation.

\[
C_p \leftarrow C_p + k_n v_n. \tag{5}\]

In our work the value of \(\zeta\) is set 0.05 times the proxy radius.

2. Once the HIP penetrates the object, the value of \((v_n, v_h)\) becomes negative and the magnitude of \(v_t\) becomes non-zero from equation \([1]\). Then the proxy is updated along the tangential direction according to the following equation.

\[
C_p \leftarrow C_p + \delta v_t + k_n v_n. \tag{6}\]

The value of \(\delta\) decides the step size to move the proxy in the tangential direction. If the proxy is updated only along the tangential direction, there exists a possibility that the proxy may sink into the point cloud. To avoid this we incorporate the term \(k_n v_h\) in the proxy movement which moves the proxy along the normal direction by a small amount. The value of \(k_n\) is very small and it is chosen such that proxy neither leaves the surface nor remains inside the surface.

3. When HIP comes in free space, the proxy is expected to latch on to HIP as explained earlier. But sudden withdrawal in penetration of the HIP toward the free space may lead to sudden vibration. To ensure a smooth movement of the proxy during withdrawal process, the following proxy update equation is used.

\[
C_p \leftarrow C_p + k_h v_h. \tag{7}\]

To ensure all three conditions, we run a separate thread to check the magnitude of \(v_n\) and \(v_h\) continuously. The parameters \(k_h, k_n\) and \(\delta\) are chosen empirically and are less than 1. In this study we have used \(k_h=0.002, k_n=0.064\) and \(\delta=0.000008\). For a larger value of \(\delta\), proxy converges faster but it creates some undesirable vibration in the haptic device. On the other hand for a smaller value of \(\delta\), it does not cause vibration but convergence is a bit slower. It may be noted that the normal can also be computed locally from point cloud data by fitting the surface points. But it cannot measure the sinking of proxy, which is needed to keep the proxy on the point cloud data and hence such a technique cannot be used to update the proxy.

### 3.2 Kernel Density Estimation

The above discussed algorithm assumes that the point cloud density is uniform over the data. In fact the density of points and the details of the object are closely related. For example an object with a flat surface requires less number of points whereas an object with very fine details requires a large number of points. Because of these reason the point cloud density over the surface of an arbitrary object cannot be considered uniform.

In our work we use a variable proxy radius depending on the local point cloud density using kernel bandwidth estimation. The kernel density estimator from an unknown distribution of the samples \((x_1, x_2, \ldots x_n)\) is given by

\[
\hat{f}(x) = \frac{1}{nb} \sum_{i=1}^{n} K(x - x_i/b), \tag{8}\]

where the parameter \(b\) is called the bandwidth and \(K(\cdot)\) is the chosen kernel. The function \(\hat{f}\) estimate the pdf \(f\) of the distribution of the sample with a given bandwidth \(b\) as explained in \([7]\). Proxy radius is varied according to the relation \(r_p = \beta b\) where \(\beta\) is an appropriate scale factor. The proxy variation is limited by two limiting radii \(r_1\) and \(r_2\) where \(r_1\) denote the lowest possible proxy radius and \(r_2\) denote the highest possible proxy radius. Assuming a Gaussian kernel, the optimum bandwidth \(b\) for a univariate data is given by equation \([9]\)

\[
b = 1.06 \sigma^{-1/5}, \tag{9}\]

where \(\sigma\) denotes the standard deviation (SD) of the sample points. The proxy radius is varied according to the relation given in equation \([9]\) between the two limiting radii \(r_1\) and \(r_2\). We estimate \(\sigma\) locally over a neighborhood from the given point cloud.

### 3.3 Scaling, Rotation and Translation of the Object

In order to find out the normal to the surface as explained in the previous section, we need to find out the points inside the proxy. To reduce the search space for finding neighborhood point of the proxy, we partition the haptic space using a regular 3D lattice of voxels and map the point cloud data to this 3D lattice.

We first filter the point cloud data with a mean filter before resampling them to the 3D lattice, to avoid the bump during rendering. Let us consider a single voxel as shown in Fig. \([2]\). Let \(\{x_1, x_2, \ldots x_m\}\) be the unordered data points inside the lattice, and the mean of \(x_i\)'s is given by

\[
p_i = \frac{1}{m} \sum_{i=1}^{m} x_i. \tag{10}\]

The mean is calculated dynamically at run time. This prevents unnecessary memory requirements and enhance the speed of execution. We represent the object only by the mean of points inside the lattice. The lattice which has data points and subsequently the mean of points and the details of the object are closely related. For example an object with a flat surface requires less number of points whereas an object with very fine details requires a large number of points. Because of these reason the point cloud density over the surface of an arbitrary object cannot be considered uniform.
experience objects of all sizes at different resolutions to get a feel of overall structure to a finer details from the same data set. For instance, if a user wants to feel the 3D point cloud model of an entire monument, one needs to scale down the model to fit it into the 3D lattice. However, if the user wishes to feel a smaller part of the monument like a pillar at a closer level, that part of the model has to be scaled up appropriately. To incorporate all three features scaling, rotation and translation, we use an affine transform given by equation (11) where matrix $T$ can be a scaling or a rotation matrix and $T_0$ is a translation vector.

$$x_{T} = [T][x] + [T_0].$$  \hspace{1cm} (11)

To modify the resolution of the object, we transform the data points by using a scaling matrix $T$ which can be any diagonal matrix. After scaling up, the transformed points $x_{T_i}$ become dispersed and may require more number of lattices to inhabit. In this way, we represent the scaled object with more data points which increases the graphic and haptic resolution. Similarly while scaling down, number of active grids and hence the resolution decreases. As explained earlier, data smoothing is performed on the transformed points after which they are mapped back into the 3D lattice. The sampled set of points, $s(p)$ are finally rendered haptically.

Rotation allows the user to feel even the rear side of the object. We observed that, rotation feature gives the user a better perception about the overall structure. In order to rotate the model, we transform the point cloud data by using an appropriate orthonormal rotation matrix. The rotated points are then mapped back into the lattice after proper smoothing. Since our haptic workspace is limited, we cannot render a scaled up version of whole object at a time. In such cases, we use the translation vector $T_0$ to bring the desired part of object into the haptic workspace. In this way an user can feel the complete object more precisely at a closer level when needed and zoom out when context is desired.

### 4 RENDERING

To render the reaction force we need the penetration depth of HIP into the surface. If $(v_h, v_n) < 0$ then proxy has touched the object and a force needs to be fed back by the haptic device. Subsequently, the reaction force is evaluated as $f_r = -Kd$, where $K$ is the Hooke’s constant and $D$ is the penetration depth of HIP into the object. For finding the penetration depth $D$ we assume the sinking of the proxy into the point cloud negligible. Hence we approximate $D$ as

$$D = \left|v_h - r_p\right| v_n \quad \text{for} \quad \left|v_h\right| \geq r_p$$
$$D = 0 \quad \text{otherwise}, \hspace{1cm} (12)$$

where $r_p$ is the radius of the proxy which is varied according to the equation.

### 5 RESULTS

The proposed method was implemented using visual C++ in a Windows XP platform with CORE 2QUAD CPU @2.66 GHz with 2 GB RAM. A Falcon haptic device from NOVINT with 10 cm³ cube of interaction space is used to feel the reaction force. The 3D lattice of size $300 \times 300 \times 300$ is used to map the point cloud data with a lattice spacing of 0.04 m between the adjacent lattice nodes. We can increase the resolution by decreasing the space between two

![Figure 2: Illustration of data smoothing for re-sampling on a discrete grid. The small blue points represent the point cloud data and the big blue circle $p_i$ represents the mean of all the points inside that grid. The red circle denotes the center of grid.](image)

![Figure 3: Illustration of calculating resultant force along the tangential direction for the proxy movement after incorporating surface friction.](image)
Figure 4: A point cloud data of an Indian idol rendered at 3 different resolutions. (Data Courtesy: www.archibaseplanet.com)

Figure 5: Point cloud corresponding to an angel rendered at 3 different resolutions. a) At lowest level of detail b) Model rotated by 90 degree and at higher level of detail c) Model rotated by 180 degree along with scaling. (Data Courtesy: www.archibaseplanet.com)

Figure 6: a) Point cloud data corresponding to a Dragon at the lowest resolution b) Model of the dragon scaled and rotated by 180 degree and c) Model of the dragon scaled up and translated to bring the front part of it into the active cube. (Data Courtesy: www.archibaseplanet.com)

Figure 7: Illustration of estimated varying radius of proxy at different density locations. (Data courtesy: www.turbosquid.com)

adjacent lattice. We use HAPI library for haptic rendering. Fig. 4 shows the point cloud corresponding to an Indian idol with 47516 points. The proxy is shown in pink color in the figure. Fig. 4a corresponds to the lowest level of surface details. We successively zoom the object and the corresponding scaled objects are shown in Fig. 4b and Fig. 4c. It is quite clear from the figure that later figures have more details and a visual perception of closeness. In the above and subsequent two figures, haptic interaction space is shown by a black wire mesh. The green line attached to the proxy shows the normal direction computed at the shown proxy location. Fig. 5 demonstrates the rotation feature. Fig. 5a is shown at the lowest resolution and it allows rendering at a gross level, while in Fig. 5b and Fig. 5c it is rotated by 90 and 180 degree, respectively. This feature becomes desirable when the user is interested to feel the rear part of the object. Fig. 6a and 6b are other examples of scaling and rotation. The scaled up version of Fig. 6a is shown in Fig. 6c in which an user can feel even minute variations of the model. Hence they can have a more realistic experience. Whenever the scaled object goes out of the haptic space, it gets clipped off. Fig. 6c demonstrates the translation feature in which the object is translated to bring the front part of it into an active cube and feel it more precisely. Fig. 7 demonstrates two different regions of point cloud model having varying density and the computed surface normals at corresponding locations. The left side of the Fig. 7 shows the low density region where the radius of proxy is large and the right side of the Fig. 7 shows the higher density region and here the radius of proxy is smaller. The average time required to compute the normal vector is 0.0328 ms, which means we could update the proxy nearly 30 times faster than the required haptic update frequency of 1 KHz. The average time required for data smoothing and loading it into haptic cube depends on the density of input point cloud data and it was observed to be around 1.109 s for a point cloud data with 543652 points.

The fact that the natures of the two plots are quite different during the interaction demonstrate that the reaction force does not get affected due to the change in proxy radius.

In haptic rendering there is no standard way to validate the results. We demonstrate the validation of our results by using a point cloud model of sphere having a radius of 0.025 m. The illustration of variation of proxy radius during haptic interaction with the object is shown by Fig. 8(a). In free space proxy radius is 0.025 m. As the proxy touches the point cloud model of object, its radius changes according to the local density of points. We also computed the ideal reaction force by using the known implicit equation of the sphere. Force computed by using the implicit surface equation is compared with the reaction force computed by the proposed rendering technique. From Fig. 8(b) we can see that the ideal reaction force is quite close to the rendered force. In order to validate the scaling technique, we scaled up the model of sphere and recorded the vari-
variation of proxy radius and rendered force. Fig. 9(a) and Fig. 9(b) illustrate the variation of proxy radius and the comparison between the rendered and the ideal force at a scaled up level respectively. We noticed that there is no dependency between the reaction force and proxy radius. For experiments related to Fig. 8 and 9, we served that, even after scaling the sphere, error was within the limit of 0.4% of the rendered force. Further, we asked different subjects to evaluate the performance of the proposed technique on various 3D point cloud data. The subjects reported a very good experience and did not complain about any unusual slipperiness or vibration.

6 CONCLUSIONS

In this work, we proposed an algorithm to render a variable density point cloud data without constructing the surface or polygonal mesh out of them. Our primary goal was to efficiently render the 3D point cloud data with attributes like scaling, rotation, translation, and friction to augment the users’ experience. The distance between the proxy and HIP was minimized iteratively during haptic interaction. The average proxy update time was 0.0328 ms even for a highly dense point data which is far less than the required haptic update time of 1 ms. We also proposed a way to get rid of unnecessary vibration from haptic device due to mapping a large number of points into a small haptic active space. The method enhances the efficiency of searching neighborhood points around the proxy which subsequently makes the normal computation faster. In order to handle the variation in point density, we have proposed a run time kernel bandwidth estimation technique that allows the size of the proxy radius to change adaptively. We validated the proposed algorithm by using a point cloud model of a sphere and found the force obtained by the proposed rendering technique to be almost same as the ideal one. We are currently exploring the possibility of using an adaptive octree for partitioning of the haptic workspace.

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