ABSTRACT
Dynamic Tubular Grids (DT-Grids) are designed to encode grid-aligned data in level-set simulations. While they are extremely efficient for storing sparse volumetric data, they require logarithmic time for random access. We demonstrate that DT-Grids can be used to efficiently render sparse volumetric data that would otherwise not be able to fit in texture memory. For many real-world biomedical data sets, such as 3D images of microvascular and neuronal networks, this results in over 10X compression compared to representation as a 3D voxel grid. DT-Grids also have several advantages over GPU-based octrees, requiring less memory and fewer indirect lookups per voxel. Finally, storing additional information per voxel does not increase DT-Grid overhead. This is an important feature for visualizing data sets from emerging imaging methods, such as Array Tomography and infrared spectroscopy, which allow many channels to be imaged per spatial voxel.

Index Terms: I.3.6 [Computer Graphics]: Methodology and Techniques—Graphics data structures and data types; I.4.10 [Image Processing and Computer Vision]: Image Representation—Volumetric

1 INTRODUCTION
The visualization of large-scale biomedical data is important for understanding the structure and function of biological tissue. Recent advances in high-throughput microscopy [5, 7] and infrared spectroscopy [2] pose a particular challenge in visualization. These imaging methods produce large data sets representing complex yet sparsely-packed structures. Two examples of these are anatomical imaging methods producing neuronal and microvascular networks. These structures span large regions of tissue but are composed of very fine filaments that require a high-resolution representation. In addition, these data sets often have a high dynamic range (32+ bits) and contain multiple spectral components or channels, making them cumbersome to visualize since they often exceed the storage capabilities of GPUs.

We demonstrate that GPU-based Dynamic Tubular Grids [8] support efficient rendering for large sparse data sets with high-precision and multiple spectral components. We compare GPU-based DT-Grids to GPU-based octrees, known as N3-trees [4], which are often used for large-scale volume visualization [1]. We show that DT-Grids offer:

- Efficiency over N3-trees in terms of memory requirements and rendering time.
- Byte-order insertion and therefore direct streaming from out-of-core storage, making them ideal for converting large-scale volume data since the original grid does not have to be kept in main memory.

2 GPU-BASED DT-GRIDS
Dynamic Tubular Grids are an efficient run-length encoding structure for sparse grid-aligned data sets and are known to provide constant-time data look-up when voxels are accessed in lexicographic order (e.g., convolution). Like octrees, DT-Grids require a sequence of indirect look-ups for random access. Implementation details for the CPU-based structure can be found in the original article by Nielsen and Museth [8]. We describe a simplified version of this structure to accelerate three-dimensional DT-Grids for parallel random access on the GPU.

We convert a sparse uniform grid to a GPU-based DT-Grid by eliminating empty voxels and projecting all valid data onto the XY plane. This results in a stack of voxels in z-order at each position in (x, y). These voxel stacks are referred to as columns and are stored end-to-end in a single texture map (texValue). In each column, a connected component is a series of pixels that were connected in the original volume data set. Since empty voxels are eliminated, spatial context between connected components in the z-direction must be maintained. An additional 16-bit RGBA texture map (texCoord) is used to store the position and z-coordinate of each connected component in a column. Finally, a 16-bit RGB integer texture (texGrid) represents the XY plane and stores a pointer to each column.

Random access into the GPU-based DT-Grid involves a texture fetch from texGrid, which provides a list of connected components in texCoord. A binary search is then performed to find the correct connected component and position of each voxel stored in texValue. For most sparse biomedical data sets, the number of iterations required to retrieve the final voxel value is far less than the number of look-ups required to traverse an equivalently encoded octree.

3 RESULTS
We evaluate the performance of GPU-based DT-Grids for volume visualization of multi-channel 3D textures. Three data sets are represented using 4 channels with 32-bit floating point precision (Fig. 1) and a fourth spectroscopy data set is represented using a single channel with 32-bit floating point precision.

We encode each full-precision data set as both an N3-tree and DT-Grid. The memory reduction resulting from encoding is compared to the original data on a uniform grid (Fig. 2). In order to compare render time to 3D textures, we also cast the data to an 8-bit single-channel format so that it will fit on the graphics card as a 3D texture map. Since cache coherence plays a significant role in rendering time, we average the time required to render each data set along all three primary axes. The average render time (in milliseconds) is recorded over several thousand frames (Fig. 2).
Figure 1: Volumetric data sets represented as a grid of 4 32-bit floating point values. (left) 512³ (2.15GB) high-genus solid. (center) 512³ (2.15GB) data set of brain microvessels captured using Knife-Edge Scanning Microscopy. Blue and red channels map to vessel diameter (largest to smallest) and the green channel reflects depth from the cortical surface. (right) 781x687x268 (2.3GB) grid of neurons imaged using Array Tomography. Each channel represents the density of a different protein at that voxel location. Green voxels represent a subset of neuronal fibers, blue voxels represent DNA, and red voxels represent protein density at synaptic junctions.

Figure 2: (left) Reduction in data set size using \(N^3\)-trees and DT-Grids to eliminate empty space. (right) Average time required to render all voxels in three volumetric data sets (Fig. 1) and a 600x300x1641 spectroscopy image of a breast biopsy. Speedup over \(N^3\)-trees is obtained through greater cache coherence along two axes. When visualization is constrained to these two axes, rendering time becomes competitive with 3D texture maps. The spectroscopy image is an orthographic projection along one axis, therefore only the constrained time is presented.

4 Conclusion

The rendering time for DT-Grids is significantly lower than \(N^3\)-trees as a result of both (a) fewer texture fetches required to evaluate the final fragment color and (b) a greater percentage of coherent and sequential texture fetches. The rendering time for 3D texture maps is larger along the \(z\)-axis while DT-Grids take longer to render along the encoding axis. This is due to incoherent texture fetches inherent in slice-based volume rendering. Methods for smoothing the frame rate have been reported previously in the literature for uniform grids [9] and are applicable to DT-Grids. However, constraining rendering to the two most efficient axes provides performance comparable to uniform grids for data that is too large to fit on the graphics card as a 3D texture map. DT-Grids allow additional data elements to be stored per-voxel without increasing the required overhead. Additional channels can be used to store information such as lighting, time-dependent functions, and identifiers for use in selective visualization [6].

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References