Ultrasound Volume Reconstruction: Open-Source Implementation with Hole Filling Functionality

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Introduction

Motivation

Volume Reconstruction is the combination of many tracked 2D Ultrasound (US) images to create a 3D US volume. It has clinical applications, such as cross-modality registration. However, reconstruction quality can be affected by holes that result from inadequate sampling.

Figure 1: The distribution of a pixel into a volume is shown by a green box. Holes are shown in white.

Objective

We aim to create freely-available, open-source volume reconstruction software that features hole-filling capability.

Results

Qualitative Analysis

- Hole filling makes images easier to interpret
- Large holes are filled only when the kernel size is variable



Methods

Hole Filling Algorithm

- Distribute pixels into the volume using reverse tri-linear interpolation [1]
- Fill holes with a Gaussian weighted average over a cubic kernel region [2]





- **D** = Gaussian Distance Weight
- V = Voxel Intensity

Figure 2: The hole is filled with an interpolated value

• Determine kernel size based on available input



Figure 3: *Left* – There is not enough information in the kernel region for interpolation, *Right* – The hole can be filled using a larger kernel region

 Implemented as free, open-source software in the <u>Public</u> software <u>Library for Ultrasound</u> (PLUS)

Evaluation

- Generate a Ground Truth by inserting a dense set of tracked US images directly into the volume
- Introduce holes by using only every 4th slice this simulates uniformly faster probe movement



Figure 5: Images are shown for volume reconstruction without hole-filling, with hole filling using a static kernel size (3 voxel diameter), and with hole filling using a variable kernel size (7 voxel diameter maximum). All images are compared to the Ground Truth on the far left. The red arrow shows a larger hole that was not filled continuously.

Quantitative Analysis

- Intensity range: 0 255
- Hole filling reduces the Mean Absolute Error of hole voxel intensities
- Hole filling is best with a variable kernel size

	Dataset	No Hole Filling	Static Size	Variable Size
	Spine	16.98	2.52	2.30
	Prostate	64.47	17.09	7.70

Table 1: MAE of hole voxel intensities are presented for reconstructed volumes



Figure 6: The Absolute Error distribution in hole voxel intensities is shown for the Prostate volume reconstructions. The MAE are marked.

Conclusions

- A volume reconstructor was implemented as free, opensource software available at: https://www.assembla.com/spaces/plus/
- Compare the results of using a static kernel size (diameter 3 voxels) against those of using a variable kernel size (diameter of 3, 5, or 7 voxels)
- Qualitative Analysis:
 - Visual comparison, but there is potential bias
- Quantitative Analysis:
 - Calculate the Mean Absolute Error (MAE) of hole voxel
 - intensities [3]

$$MAE = \frac{\sum |V_G - V_H|}{N}$$

V_G = Ground Truth Voxel

- V_{H} = Hole Voxel
- *N* = Number of Hole Voxels



Figure 4: US data being collected on a spine phantom

- The software continues to be tested on data from the University of British Columbia and Queen's University
- Using a variable kernel size enhances the accuracy of reconstructed volumes.

References

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