### Introduction

#### Background

Tracked ultrasound (US) uses a tracking system to sample the probe’s position and orientation (pose). Unequal processing times of the imaging and tracking systems cause corresponding US images and poses to receive different timestamps. To correct for this misalignment, temporal calibration is needed to compute the time offset between the tracker and image data.

#### Objective

The goal of this work is to develop a fast and fully automatic temporal calibration methodology that is robust under a large range of imaging parameters. Furthermore, by making all code freely available, this work aims to make a practical contribution to the ultrasound research community.

### Methodology

Our temporal calibration algorithm most closely follows the work of Treece et al. [1]. Calibration begins by imaging the bottom of a water bath in a periodic, uniaxial motion. Then, temporal calibration is performed by executing the workflow that is outlined in Fig. 1 and discussed below.

#### Tracker sequence

- Compute tracker position signal
- Correlate tracker and position signals

#### Image sequence

- Compute image position signal
- Time offset

**Figure 1.** Overview of the temporal calibration algorithm

#### Computing the tracker position signal

Position information from the tracking data is projected onto the principal axis of probe motion. The signal amplitude at a given tracking frame is defined to be the distance from the mean projection.

#### Computing the image position signal

Each image is sampled along vertical scanlines. Then, for each scanline, a center-of-gravity (COG) point is computed as follows (with reference to Fig. 2): (a) the pixels with intensities below half of the maximum intensity of the scanline are discarded, (b) the contiguous region whose sum of pixel intensities is largest is sought, and (c) the center-of-gravity (COG) of that region is computed.

**Figure 2.** An example intensity profile of a scanline. All pixels below $\frac{1}{2}$ Max are discarded. Because the area—i.e. sum of pixel intensities—between $A_0$ and $A_1$ is larger than that between $B_0$ and $B_1$, the region of the signal between $A_0$ and $A_1$ is used to calculate the COG.

### Methodology Cont’d

For each US image, a line is then fitted through the detected COG points using the RANSAC algorithm (Fig. 3 A, B). The signal amplitude of a given US image is then taken to be the distance from the midpoint of that image’s COG line to the midpoint of the mean COG line.

#### Correlating the tracker and position signals

The tracker and image position signals are aligned by finding the time shift that minimizes the sum-of-squared distance (SDD) between the two signals (Fig. 3 C, D). The optimal time shift is computed using a multi-resolution search.

**Figure 3.** (A) US image of the bottom of a water tank and (B) the same image with COG points and a COG line. (C) Normalized tracker position signal (blue) and image position signal (green) before calibration, and (D) after calibration. x-axes are time, and y-axes are the normalized signal values; $\Delta t$ is the temporal offset.

### Results and Discussion

For evaluation 200 US sequences, collected under varying imaging parameters, were used for evaluation. The algorithm was found to compute temporal offsets precisely, generally with a standard deviation of <5ms between scans with the same imaging parameters. Additionally, temporal offsets were found to vary considerably as a function of imaging parameters, falling in the range of 40-90ms. All code is freely available as part of PLUS, which is an open-source software package providing library functions and applications for tracked US image acquisition, calibration, and processing [2].

### Conclusions

A temporal calibration algorithm equipped with a robust line detection scheme was presented. The algorithm’s precision was evaluated under a range of imaging parameters, and was found to be adequate for routine use in tracked ultrasound applications. The algorithm is freely available as part of PLUS [2].

### References


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