Application of hybrid MR-ultrasound imaging to multi-baseline thermometry
Pei-Hsin Wu1, Cheng-Chieh Cheng1, Frank Preiswerk1, Bruno Madore1
1Department of Radiology, Brigham and Women’s Hospital, Harvard Medical School, Boston, MA, United States.

Introduction
Proton resonance frequency (PRF)1 is the most commonly-used MR thermometry method, mostly because its temperature coefficient is nearly independent of tissue type2. However, in abdominal imaging, breathing motion creates additional phase shifts that may corrupt temperature measurements. Two main PRF approaches have been proposed to handle breathing motion: the referenceless3 and multi-baseline4 methods.

The present work fits in the same general category as multi-baseline thermometry. However, ultrasound (US) signals are further included here as biometric navigators5, leading to hybrid MR-US acquisitions6. US-based sensors as shown in Fig. 1 were employed: they consist of an MR-compatible transducer, a 3D-printed capsule, ultrasound gel for acoustic coupling and two-way tape for skin fixation. These devices are called here ‘organ configuration motion’ (OCM) sensors, and one was fixed just below the rib cage to characterize breathing motion. As compared to more traditional motion-monitoring approaches such as respiratory bellows and navigator echoes, advantages of OCM sensors include: OCM sensors contact only a small area of the torso and their signals directly capture internal motion (unlike bellows), and OCM signals are acquired in parallel with MRI and as such do not affect the overall scan efficiency (unlike most navigator echoes). Compared to conventional multi-baseline thermometry, the inclusion of OCM signals into the reconstruction chain led to improved PRF measurements in the presence of free breathing.

Methods
Multi-baseline thermometry requires reference data to be acquired, without cooling or heating, to provide a phase reference. Temperature-induced phase shifts are then calculated with respect to such reference data. For every time point in the heating/cooling process, a best-matching image must be chosen from the reference set. While the matching process would normally be based on image similarity, the present method further involved signals from an OCM sensor.

Human scans were performed after informed consent, on a 3.0T Siemens Verio system (45mT/m, 200/T/m/s), using a flexible body coil matrix, with parameters: FOV=38×38cm², matrix size=192×192, slice thickness=5mm, flip angle=30°, bandwidth=390Hz/px, TR/TE=10/4.8ms, 500 dynamic acquisitions, temporal resolution=0.6 s, partial Fourier ratio=5/8. An MR-compatible OCM sensor that included a 1MHz transducer (Imasonics, 15-mm diameter) was employed (Fig. 1). Phase-based, Doppler-like processing was performed on OCM signals to generate velocity and displacement measurements, for depth locations up to 15cm below the sensor. In one of two measurements, a cold compress was applied to the abdomen, in-between the body matrix and the OCM sensor, to create small but detectable temperature changes. The cooling phase following the application of the compress was captured, and reference data were acquired after a 15 min delay, once most of the cooling effect had dissipated. Similarity in image and OCM signals were used to identify optimal matches: while images tested similarity in morphology, OCM signals tested consistency in terms of biomechanical motion.

Results
Figure 2 shows measurements derived from OCM signals: mean displacement (black line) and velocity (green line) were averaged over all sampled tissue depths. Velocity was about zero at full inspiration and expiration while displacement was maximal at these time points. Figure 3 shows an example of a matching process, where an optimal match was sought for the image in Fig. 3a. Using only image similarity, the image in Fig. 3b was identified as the best match, as opposed to that in Fig. 3c when further including OCM-based information (c). Yellow and red markings, common to Fig. 3a, 3b and 3c, help appreciate how Fig. 3c appears to be the better match, as opposed to Fig. 2b. Using the proposed processing, temperature change is displayed as a color overlay in Fig. 4a, for three time points. Two different ROIs, at shallower and deeper locations, are defined in Fig. 4b with blue and white squares, respectively. As shown in Fig. 4c the shallower ROI, the one in closer proximity to the cold compress, was associated with a temperature change of about 10°C while temperature was nearly constant for the ROI located more deeply in the abdomen.

Discussion and Conclusion
An improved version of the multi-baseline free-breathing thermometry approach was introduced, which included the use of OCM sensors and associated hybrid MR-US imaging. The OCM sensor offered rich motion-related information, including phase-based
Doppler like measurements of the displacement and velocity of tissues due to breathing. Such information helped identify the best available matches out of a library of pre-heating data, as part of multi-baseline thermometry. Both morphological similarities and internal motion as detected by OCM were taken into account in the proposed processing, giving rise to and hybrid US-MR imaging method for PRF thermometry.

References