

Estimating Regional Myocardial Contraction Using Miniature Transducers on the Epicardium

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1 **Estimating Regional Myocardial Contraction Using Miniature Transducers on the Epicardium**

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13 **Abstract**

14 This paper describes an ultrasound system to monitor cardiac motion using miniature transducers
15 attached directly to the epicardial surface. The aim is both as a research tool for detailed studies of
16 cardiac mechanics, and to develop a continuous, real time system for perioperative evaluation of heart
17 function. The system was tested on a porcine model. Two 3 mm diameter, 10 MHz ultrasound
18 transducers were sutured to the epicardial surface. As the epicardial surface is the reference for the
19 velocity and strain estimations, this procedure compensates for the motion of the heart. The short
20 distance allows use of high frequencies and pulse repetition rates. The system was driven in pulse-echo
21 mode, using electronics developed for the application, and RF-lines were recorded at pulse repetition
22 rate 2500 s^{-1} . The endocardial border was detected using an algorithm based on fuzzy logic with filtering
23 to reduce noise and remove outliers, and the myocardium was divided into 4 layers. Inside the
24 myocardium, radial tissue velocity as function of depth was calculated from the recorded RF signals,
25 and the velocity estimates were used to estimate radial strain rate and strain, and to track the motion of
26 the myocardial layers. The scope of this paper is technical, giving a detailed description of system
27 design, hardware electronics, and algorithms, with examples of processed velocity patterns and
28 myocardial strain curves. The results from a study on a porcine model demonstrate the system's ability
29 to estimate myocardial velocity and strain patterns and to track the motion of the myocardial layers,
30 thereby obtaining detailed information of the regional function of the myocardium.

31 *Keywords:* Ultrasound, strain, tissue velocity, myocardium, pig, miniature transducer, **perioperative**
32 **monitoring**.

33 **Introduction**

34 Reliable methods to assess ventricular function during and after cardiac surgery are essential tools to
35 evaluate patient prognosis (Landesberg et al. 2001; Espinoza et al. 2011). The most common method
36 for monitoring heart status is the electrocardiogram (ECG). ECG is an invaluable tool for heart
37 monitoring, but the sensitivity for detecting an occlusion or ischemia is regarded as low (Comunale et
38 al. 1998; Crescenzi et al. 2004; Smith et al. 1985). ECG and hemodynamic monitoring address the
39 global heart function, and are used as low-threshold, continuous first-line monitoring methods
40 (Ludbrook et al. 1993). Tissue Doppler echocardiography is a more specific and quantitative tool for
41 the assessment of cardiac function, being able to measure regional tissue velocity and displacement in
42 the myocardium (Skulstad et al. 2006; Yu et al. 2007). But cardiac ultrasound systems are primarily
43 designed for intermittent imaging, not for continuous monitoring. Small single-element ultrasound
44 transducers attached directly to the heart surface provide an alternative option for continuous heart
45 monitoring. Ellis et al. (Ellis et al. 1956) used *sonocardiometry* to measure left ventricular (LV)
46 diameter continuously. In this technique, two transducers were attached to the heart surface, one as
47 transmitter, the other as receiver, and the method was used to measure local dimensional changes of the
48 myocardium (Bugge-Asperheim et al. 1969). This is a valuable research tool, but too invasive for
49 monitoring patients during cardiac surgery, and it does not provide local information of strain inside
50 the myocardium. Hartley et al. demonstrated a method using one single-element transducer on the
51 epicardium in a pulse-echo technique, to measure myocardial thickening (Hartley et al. 1983). We
52 developed this method further, attaching miniature ultrasound transducers to the epicardium, measuring
53 myocardial velocities (Hoff et al. 2008; Nguyen et al. 2011).

54 The myocardial deformation, represented by myocardial strain, reflects the work load of the
55 myocardium, and the strain is therefore a suitable quantitative parameter characterizing the myocardial
56 function (D'hooge et al. 2000; Kowalski et al. 2001; Kukulski et al. 2002; Kukulski et al. 2003;
57 Weidemann et al. 2002). Cardiac mechanics is complicated and still not completely understood. For
58 example, most studies of the distribution of radial strain across the LV wall show increasing strain
59 values from the epicardial layer to the endocardial layer (Matre et al. 2005), while others show the

60 highest strain in the mid-myocardium (D'hooge et al. 2001). Computer models using the Finite Element
61 Model to simulate the LV contraction indicate that transmural layer strain depends on the local curvature
62 of the investigated segment (Choi et al. 2010; Choi et al. 2011). Studies have also showed that there is
63 a link between ischemia and changes in radial strain (Matre et al. 2007; Skulstad et al. 2006). Hence,
64 detailed measurements of ventricular motion are of great interest to better understand the details in the
65 heart mechanics, and may contribute to improve heart disease diagnosis.

66 This paper builds on the previously described methods using miniature transducers attached to the heart
67 surface, developing these further to monitor cardiac motion, most notably strain, as function of depth
68 into the myocardium. Two transducers were attached to the epicardium and used in pulse-echo mode,
69 and results were processed to find velocity, strain, and displacement as function of depth into the
70 myocardium. Compared to transthoracic ultrasound, our approach uses transducers that move with the
71 heart surface, using the epicardium as reference for the velocity calculations, thereby compensating for
72 the heart's own movements. This can be beneficial for strain and strain rate measurements, which are
73 based on small velocity differences within the myocardium. In conventional echocardiography, parts of
74 the myocardial tissue move in and out of the imaging plane, causing problems in the interpretation.
75 Furthermore, the short depth allows higher frequency and higher pulse repetition frequency, allowing
76 improvement in spatial and temporal resolution (Nguyen et al. 2011). The proposed clinical application
77 of this system is two-fold: The long-time goal is to develop a small, dedicated system to continuously
78 monitor regional cardiac function during and after cardiac surgery. This should have faster response
79 time and better sensitivity and specificity than ECG, and operate continuously, perhaps for several days,
80 with minimal operator interaction. It should also be smaller, simpler and less expensive than
81 conventional cardiac ultrasound scanners. Secondly, as a shorter-term goal, this system's high spatial
82 and temporal resolution, inherent compensation for the heart's own movements, and continuous
83 capabilities make it suited as a research tool for fundamental studies of heart mechanics. It can give
84 continuous detailed measurements of the heart's contraction pattern and regional strain distribution over
85 longer periods, on a level not easily achieved by conventional ultrasound systems.

86 Sensors that can track myocardial motion, may be used to detect motion abnormalities and give early
87 warning of potential complications occurring during surgery such as ischemia. Motion changes appear
88 before changes detected by other methods such as ECG. This has been demonstrated in patients and in
89 animal experiments (Espinoza et al. 2011; Hyler et al. 2015). We, and others, have found an immediate
90 reduction in systolic wall thickening velocities, together with an increase in early diastolic wall
91 thickening (post-systolic wall thickening). Ischemia can result from obstruction in blood flow in the
92 bypass grafts from various reasons, such as thrombosis, kinking of graft or even wide retraction of the
93 sternum retractor (Espinoza et al. 2012). The early warning of such blood flow impediments can give
94 the surgeon time to correct the underlying cause before chest closure.

95 The data acquisition system is identical to the one used by Espinoza et al. (Espinoza et al. 2011). The
96 present paper gives more detail on the technology, which was not so thoroughly described in this purely
97 clinical paper. Further, Espinoza et al. (Espinoza et al. 2011) used a pulse Doppler estimator to find the
98 velocity at a fixed depth. This has been extended to calculate velocity at all depths, and combine these
99 data with boundary detection and tissue tracking algorithms, allowing us to track myocardial layer
100 motion and estimate strain in the myocardial layers, with example measurements on an open thorax
101 porcine model.

102 **Materials and Methods**

103 **Animal Experiment Procedure**

104 The ultrasound system was tested in open chest porcine experiment. The ultrasound recording was
105 obtained in a study previously published (Hyler et al. 2015). The use of animals in that study was
106 approved by National Animal Research Authority in Norway (No. 27/09-1747). The handling of the
107 animals was in accordance with institutional guidelines, and national and international regulations. The
108 re-use of the recordings in this study is in accordance with the Three R's, to reduce the number of
109 animals used in research (Directive 2010/63/EU). Two sensors were sutured to the epicardial surface of
110 the left ventricle in the apical region, near the intervention area, and in the basal region, far from the
111 intervention area, as shown in Figure 1. The sensors had to be placed at stable positions. ECG and LV

112 blood pressure were recorded concurrently and synchronized with the ultrasound measurement from
113 the surface of the left ventricle.

114 **Transducers, Electronics and Data Acquisition**

115 A brief description of an earlier version of this system can be found in (Hoff et al. 2008). A schematic
116 drawing is shown in Figure 2. The system consisted of a two-channel ultrasound transmit-receive
117 system built in-house from state of the art electronic components. This was connected to two single
118 element transducers sutured to the epicardium of the LV wall. The results were sampled by a high-
119 speed data acquisition board, and stored on a computer disk. The system employed custom-build single-
120 element transducers (Imasonic SAS, Besancon, France). The transducers have 3 mm active diameter,
121 center frequency 10 MHz, 60% bandwidth, and are focused geometrically to 20 mm. In each
122 experiment, two such transducers were sutured to the epicardium of the LV wall, at two different
123 positions. Ferrite ring transformers at ratio 9:4 were connected between the transducers and the rest of
124 the equipment to provide galvanic isolation for electric safety, with the additional benefit of improving
125 the electrical impedance matching from the transmit electronics to the transducer. The two transducers
126 were excited simultaneously, and the received echoes were split into two separate receive channels in
127 the transmit/receive switch. The transducers were mounted so that the distance between them was larger
128 than twice the maximum imaging depth, ensuring that the receiving was finished before the direct wave
129 from one transducer reached the other to avoid interference between the two transducers.

130 The analog electronics for the ultrasound transmit-receive system was assembled in-house using
131 evaluation boards from electronics manufacturers. The transmitter was based on a Supertex
132 MD1210DB1 evaluation board (Microchip Technology, Chandler, AZ, USA), programmed to transmit
133 short pulses of center frequency 10 MHz at repetition rate 2500 pulses/s. The transmit voltage was set
134 to $\pm 18V$. The MD1210DB1 evaluation board was modified by decreasing the output protection resistor
135 to 22Ω , increasing the transmitted power while still giving sufficient short-circuit protection, and the
136 internal oscillator was disabled. Clock signals were taken from an external oscillator, to obtain phase
137 synchronization between the transmitted ultrasound pulses and the sampling of the echoes.

138 A two-channel passive transmit-receive switch was implemented by using a diode network. The
139 transmit pulses were simultaneously sent to the two transducers, whereas the received echoes from the
140 two transducers were separated and directed to the receiving amplifier.

141 Timing and sampling were controlled by an external oscillator (IQXO-350C, IQD Ltd., Somerset,
142 England), running at four times the transmit frequency, i.e. 40 MHz. The transducers were driven by
143 two-cycle symmetric square wave pulses, generated by dividing the oscillator clock frequency by four,
144 and the pulse repetition frequency was fixed to 1/4000 of the center frequency, giving a pulse repetition
145 rate of 2500 pulses/s. Echoes received from the myocardium were amplified in a dual-channel ultra-
146 low noise amplifier, AD8332-EVALZ (Analog Devices Inc., Norwood, MA, USA) and sampled by a
147 high-speed 14 bit digitizer board (NI-PCI 5122, National Instruments Inc., Austin, TX, USA), placed
148 on the PCI-bus of a desktop personal computer. The external 40 MHz oscillator was also used to control
149 the sampling, to avoid jitter between the transmitted and received signals, giving sample rate 40 MS/s.

150 The system was controlled by software written in-house using LabVIEW (National Instruments Inc.),
151 controlling the acquisition and processing of the echoes. During operation, the results were displayed
152 in real-time on the computer screen, both as wall thickness (M-mode) images over time, and as tissue
153 velocity at a fixed depth. Unprocessed RF-lines were streamed to disk during selected time intervals for
154 detailed processing and evaluation off-line. These stored scanlines were used in the calculations
155 presented in this paper. A user-friendly graphical user interface was designed to control the operation
156 of the system.

157 To support the ultrasound data, ECG and blood pressure were registered synchronously with the
158 ultrasound measurements using separate analog input channels in the low-speed multi-function DAQ,
159 operating at rate 500 Samples/s. In the experiment described in this paper, the main purpose of these
160 signals was to provide time references for the cardiac cycle. ECG signals were registered by a Siemens
161 SC 9000XL monitor (Siemens, AG, Erlangen, Germany) and the analog output from this monitor was
162 sampled by a separate 16-bit digitizer board (NI-USB 6211, National Instruments Inc.). The delay in
163 the analog ECG unit was measured using a signal generator and an oscilloscope to 21.5 ms, and this
164 was compensated in the processing and display. Instantaneous blood pressures at up to three different

165 positions were measured invasively by three Millar MPC-500 Mikro-Tip Pressure Transducer
166 Catheters, connected to Millar TC-510 Pressure Control Units (Millar Instrument, Houston, TX, USA).
167 The locations of the three pressure catheters varied between different experiments, but in most cases,
168 they were positioned in the aorta, left ventricle and left atrium. The microvolt signals from the TC-510
169 control units were amplified 400 times and bandwidth-limited to 100 Hz by in-house developed
170 electronics designed around INA101 instrumentation amplifiers (Texas Instruments Inc, Dallas, TX,
171 USA). The outputs from these amplifiers were sampled simultaneously with the ECG signals, using the
172 same AD-board. The amplifier circuits used to amplify the pressure catheter signals showed no delay.
173 Synchronization between the low-speed digitizer for ECG and pressure and the high-speed digitizer for
174 ultrasound signals was achieved by letting an analog output in the low-speed DAQ control the analog
175 gain in the ultrasound pre-amplifier. This gain control responds fast enough to control the TGC. This
176 gain was set briefly to zero at start of the recordings, creating a brief lack of signal in the received
177 ultrasound signals. By this procedure, synchronization between the pressure and ECG-signals and the
178 ultrasound recordings was achieved at precision limited by the sample interval of the low-speed
179 digitizer, i.e. 2 ms.

180 **Data processing**

181 The received echoes were saved to disk as raw, unprocessed RF-scanlines. These scanlines were loaded
182 into Matlab (The MathWorks Inc., Natick, MA, USA) for estimating the myocardial velocity, strain
183 and strain rate relative to the transducer, as function of distance.

184 M-mode images were rebuilt from the recorded RF signals, using the Hilbert transform for envelope
185 detection. The M-mode images served mainly as background maps to define the position of the
186 endocardium and the myocardial depths for estimating strain. The endocardial border was found by a
187 boundary detection algorithm, based on fuzzy logic, and moved 1.5 mm inwards to avoid boundary
188 effects. A thin layer beneath the epicardial surface was omitted to avoid near-field artifacts. The
189 thickness of this layer was 4 mm in the apical region and 3.5 mm in the basal region. End-diastole was
190 determined as the onset of the R-wave in the ECG recordings. Segmentation into myocardial layers was
191 done at end-diastole, where the myocardial wall was divided into n equally sized layers, from the

192 epicardium to the detected endocardial border. The number of myocardial layers n , giving the layer
193 thickness, could be chosen as a balance between spatial resolution and noise, and was in this study
194 selected to $n=4$.

195 The motion of each layer was determined by tracking the velocity forward and backward over one
196 cardiac cycle. A weighted average of the forward and backward tracking results was used to
197 compensate for drift. Radial strain rate was computed as the spatial velocity gradient, estimated from
198 linear regression within each layer, and the radial strain for each layer was found by temporal integration
199 of the strain rate. The end-diastolic strain was set to zero at each heart cycle, as the heart should return
200 to initial state before a new cycle (D'hooge et al. 2000). It should be pointed out that reliable velocity
201 estimates and endocardial boundary detection are essential to obtain good radial strain estimates.

202 **Velocity Estimations**

203 Figure 3 illustrates the flow chart of velocity estimation. Local velocity was estimated from the acquired
204 RF-lines. These were first filtered using a zero-phase 4th order bandpass Butterworth filter centered
205 around the transmit frequency, to remove noise. Tissue velocities were then estimated from the RF
206 signals by using cross-correlation to estimate the time-delay between successive scanlines. Cross-
207 correlation is an established tracking method to obtain a high signal to noise ratio, however, this method
208 reduces the spatial resolution, and is computationally very heavy. The accuracy of the time delay
209 estimates, and consequently, the velocity estimates, was improved by first up-sampling the RF signals
210 a factor $R=10$ using a FIR-filter based interpolation method, before the cross-correlation was calculated
211 (Nguyen et al. 2011). Then, the estimate for the position of the peak in the cross-correlation curves was
212 improved by using sub-sample interpolation applying a parabolic-fit (Céspedes et al. 1995). Cross-
213 correlation between successive scanlines was done using a kernel size corresponding to $616 \mu\text{m}$, or 4λ ,
214 with 50% overlap. Here, $\lambda=c/f$ is the wavelength of the transmitted pulses, c is the speed of sound, and
215 f is the center frequency of the transmitted pulses. The displacement between two consecutive RF-lines
216 was restricted within the interval $[-\lambda/2 \ \lambda/2]$, giving a maximum detectable velocity $v_{max} = \frac{c f_{PR}}{2f}$

217 =192.5 mm/s. The minimum detectable velocity was $v_{min} = \frac{c f_{PR}}{R f_s} = 9.625$ mm/s where f_{PR} is the pulse
218 repetition frequency, and f_s is the sample rate.

219 **Endocardial boundary detection**

220 The deepest layer in these estimations is limited by the endocardial border. Inside the myocardium, the
221 layers are tracked based on the velocity estimates, but the endocardial border is better tracked based on
222 the large differences in echo strength between blood and myocardial tissue. Several boundary detection
223 methods for two-dimensional echocardiographic images have been presented in the literature
224 (Alshennawy and Aly 2009; Chu et al. 1988; Feng et al. 1991; Setarehdan and Soraghan 1999), mainly
225 based on the intensity of the M-mode image. The method used in this paper was described by Abdallah
226 et al. (Alshennawy and Aly 2009), using a fuzzy logic technique to determine the image edges. A block
227 diagram of the method used to detect endocardial border is shown in Figure 4. The estimate for the
228 endocardial boundary determined from these image edges was then improved by filtering in the time
229 direction and employing a snake algorithm (Kass et al. 1988) in the time direction to smoothen the
230 curve.

231 On our data, the robustness of the endocardial border detection obtained from this fuzzy logic technique
232 was found to be better than a conventional edge detection method using the Sobel operator, in line with
233 the results from (Alshennawy and Aly 2009). Instead of having only true or false values as Boolean
234 logic, the membership value in Fuzzy logic varies continuously between 0 and 1. The membership
235 function is a curve used to calculate the membership values for pixels from the gray scale M-mode
236 image as shown in Figure 5. The fuzzy system rules given in (Alshennawy and Aly 2009) based on the
237 membership values of pixels in a 3x3 mask were used to detect the boundary. The range of the M-mode
238 image was mapped to gray scale of range [0 255] as shown in Figure 6. The results were found to be
239 sensitive to the thresholds a and b used to determine whether a pixel is black or white, and these
240 parameters had to be adjusted for each M-mode image. In the following example, the thresholds were
241 set to $a=140.25$ and $b=214.2$. After fuzzy logic step, the gray scale M-mode image became an image
242 which has “white” pixels at the boundary and “black” pixels at the other positions.

243 This fuzzy logic method will detect the boundaries of the myocardial fibers. In this study, it was only
 244 applied to find the endocardial border. A depth range limiting the search for the endocardial boundary
 245 was defined from 10 mm to 19.98 mm. The processed M-mode lines were scanned using the fuzzy logic
 246 algorithm, and the boundary determined as the last ‘white’ points in the depth. The blue line in Figure
 247 7a shows the endocardial boundary detected by the Fuzzy logic processing and boundary search steps,
 248 before further processing. A maximum filter of length 7 was then applied to the detected endocardial
 249 border, as a function of time, to remove noise, resulting in the red line in Figure 7a. The maximum filter
 250 is defined as a transformation which replaces the value of the first element with the maximum value of
 251 all the elements within the running window:

$$252 \quad z_b(i) = \max(z_b(i), z_b(i + 1), \dots, z_b(i + n - 1)) \quad (1)$$

253 where $z_b(i)$ is the position of the border at discrete time i and n is the length of the filter. n is an integer.
 254 The result in Fig.7a demonstrates how this maximum filter effectively removed spikes in the original
 255 estimate.

256 After the maximum filter, the boundary estimate still contained points suspected to be outliers. These
 257 were removed by requiring the distance in the depth direction between two successive points in the
 258 boundary to be smaller or equal to the maximum velocity at that time, multiplied by the pulse repetition
 259 interval. Maximum velocity at a time is defined as the maximum velocity along the depth at that time,
 260 given by the velocity estimator described previously. As a final step, the myocardium expands
 261 monotonically during systole, and this requirement was applied to improve the results further, removing
 262 the last outliers. After applying these steps, the snake algorithm (Kass et al. 1988) was employed to
 263 smoothen the boundary along the time. The snake parameters were set to $\alpha=5000$, $\beta=0$, step size $\gamma=10$,
 264 and was run over 500 iterations. The result after applying the snake algorithm is shown in Figure 7b, as
 265 the red line. This can be compared with the result before applying the algorithm, the blue line in Figure
 266 7a. Figure 7b also compares the result with (red line) and without (blue line) requiring monotonous
 267 expansion during systole. In this example, we believe the endocardial border determined with assuming
 268 monotonous expansion during systole is slightly more accurate than without assuming monotonous

269 expansion, see around 4.68 second in Figure 7b. However, the ground truth is not known, and this must
270 be viewed as an assumption based on the shape of the curves and the M-mode image.

271 **Results**

272 The received scanlines were processed to M-mode images, and these were used to divide the
273 myocardium into four layers at end-diastole. The results are shown in Figures 8 to 11. In these figures,
274 the end-systole is marked with blue vertical lines, and the end-diastole with magenta lines.
275 Synchronously recorded ECG and LV pressure curves were used for timing, primarily to identify end-
276 diastole and end-systole. Note that the RF data analyzed in Figures 8 to 11 were acquired during a study
277 where the animal had been exposed to previous interventions, and the curves may not be representative
278 of a healthy, undisturbed myocardium. Table 1 shows the parameters used in the fuzzy logic endocardial
279 boundary detection algorithm, based on the intensity of M-mode images. The resulting estimated radial
280 velocity patterns in the myocardium are displayed as color-plots in Figure 8, for the two transducers.
281 The results in Figure 8 were calculated without any filtering of the velocity. These velocities were used
282 to calculate the motion of the four myocardial layers. Figure 9 shows the calculated motion of the
283 myocardial layers by tracking the motion in the forward direction only, while Figure 10 shows the
284 motion of the layers by using both forward and backward tracking in order to compensate for drift seen
285 in Figure 9. The detected endocardial border is displayed together with the myocardial layers in Figures
286 9 and 10.

287 The motion of the myocardium is close to periodic, and can be assumed to return to its initial state after
288 each cardiac cycle. From the results, we notice that little noise is seen in Figure 8. Likewise, very little
289 apparent drift is seen in the tracked layers in Figure 9, as the layers seem to return to their initial position
290 after each cardiac cycle. The M-mode images and forward estimated velocity images show that the data
291 near the apex contain somewhat more noise than the data near the base. Close examination of the tracked
292 layers indicates that there is some drift in the layers tracked near the apex, but considerably less in the
293 curves tracked near the base. The drifts are 0.19 mm for lower boundary and 0.1 mm for the upper
294 boundary for the 4th layer of myocardium in apical region at end-diastole, at 4.485 second, in Figure 9
295 by using only forward tracking as the thickness of the layer is 1.73 mm at end-diastole.

296 The radial strain rate was calculated as the gradient of the estimated velocities along the depth. This
297 radial strain rate was integrated temporally to obtain the radial strain, the result is shown in Figure 11.
298 One curve is shown for each of the four myocardial layers, numbered from layer 1 at sub-epicardium
299 to layer 4 at sub-endocardium. From Figure 11, it is seen that the radial strain repeats itself periodically
300 every heart cycle, as should be expected. In the recordings from the base of the heart, the strain curves
301 from the different myocardial layers have almost identical shapes. In contrast to this, the recordings
302 from the apex show substantial variation between the strain curves calculated from different myocardial
303 layers. The reason for these differences is not clear, but it should be noted that the animal model had
304 been exposed to various interventions and handling prior to this recording, mainly affecting the apical
305 region. This might explain the differences in shape between results from the base and apex, but further
306 studies involving several animals and interventions are needed to draw any conclusion about this.

307 **Discussion**

308 The aim of this study was to demonstrate a miniaturized, simple system to continuously monitor strain
309 inside the myocardium. The method offers several advantages for high signal quality compared to
310 conventional non-invasive ultrasound imaging. First, the method by design compensates for the heart's
311 own motion, and is ideally only sensitive to the myocardial contraction. In addition, the sound pulses
312 do not have to penetrate the thorax wall, giving a short distance to the region of interest, and low
313 attenuation. This allows higher pulse repetition rate and higher frequency, offering better temporal and
314 spatial resolution than conventional ultrasound imaging. This study used a pulse repetition rate of 2500
315 pulses/s and frequency 10 MHz, but these are conservative choices that may be increased.

316 The 10 MHz transducers attached directly to the myocardium gave low noise raw data of sufficient
317 resolution in space and time to allow reliable tracking the motion of the myocardial layers. The velocity
318 estimates found from these raw RF scanlines are the basis for the following calculations. Hence, robust
319 velocity estimators are essential for all further computations such as layer tracking, strain rate and strain
320 estimates. We found the cross-correlation between successive RF-lines to yield robust and reproducible
321 velocity data, although a reference to a gold standard for myocardial velocity is not available in this
322 setting. An indication of the robustness of the method is that the forward tracking in Figure 9 ended

323 very close to the first tracking point of the next cycle, even though no filter was applied to the velocity
324 estimates. This was true for both data sets, i.e. from the apex and from the base. A check on
325 reproducibility and drift in the system was done by tracking the myocardial layer motion both forwards
326 and backwards, and comparing the results. This is shown in Figure 9, where differences were found
327 negligible, demonstrating very little drift in the tracking algorithm.

328 The maximum absolute myocardial velocity detected during this study was 100 mm/s, see Figure 8.
329 This is smaller than maximum velocity of the phase shift estimator, 192.5 mm/s, defined by limiting
330 the displacement of consecutive RF lines to be within the interval $[-\lambda/2 \lambda/2]$. Hence, aliasing was not a
331 problem with the settings used in this study. In a previous study (Nguyen et al. 2011), we used the Snake
332 regularization (Kass et al. 1988) to reduce noise and remove outliers from the curves. This is a
333 computationally heavy method. In the present study, outliers were not a problem, and no filter was
334 applied to the estimated velocities. However, a low pass Butterworth filter could be used to reduce
335 noise. This required less computations than the Snake regularization, allowing faster calculations.

336 The endocardial boundary detection was based on Fuzzy logic supported by a combination of a
337 maximum filter, removal of outliers, and snake regularization. The result, in Figure 7, indicates that this
338 procedure was able to track the endocardial border well. Some further improvement could be achieved
339 by in addition requiring monotonous expansion during systole see Figure 7b, but this difference is not
340 dramatic. The apparent improvement achieved by this requirement must be weighed against the risk of
341 imposing too strict restrictions to the myocardial motion. This border detection method does not work
342 well when the border is too close to the edge of the image, as the method organizes pixels in a 3x3
343 mask. This can explain why the detected border seems to deviate from the actual endocardial border
344 around end-systole in Figure 7.

345 Strain measurements are susceptible to noise, and careful signal processing is crucial to obtain reliable
346 strain estimates. The radial strain in Figure 11 shows different behaviour of myocardium at two different
347 regions.

348 The long-term goal of this study is to develop a monitoring tool for patients during and after cardiac
349 surgery. However, it can also be useful as a research tool for fundamental studies, offering detailed

350 information on heart mechanics. The prototype sensors used in these experiments are too large to be
351 removed after chest closure. In future versions, the transducers may be thinned down to the shape of a
352 thin disc, preserving the 2 to 3 mm diameter acoustic aperture, but reducing the thickness. This could
353 allow the sensors to be removed after chest closure similar to removal of temporary pacemaker leads,
354 which are routinely used during cardiac surgery today. These are attached to the heart before chest
355 closure, but the small size allows removal through the chest wall several days into the postoperative
356 phase. A sensor encapsulated in biocompatible materials and incorporated in such temporary pacemaker
357 leads, would allow monitoring of the patients also in the interesting period of the first few days
358 following surgery. We will emphasize that this method is not an alternative to trans-thoracic ultrasound
359 imaging, but intended as a tool giving detailed information about the myocardial contraction in special
360 situations when the thorax has been opened for other reasons, i.e. during cardiac surgery.

361 **Conclusion**

362 We have developed an experimental ultrasound system using small transducers directly sutured on the
363 epicardium to measure the heart contraction pattern at high spatial and temporal resolution. We have
364 demonstrated how this can be used to track myocardial deformation and study regional myocardial
365 strain. The velocity-based layer tracking was combined with an automatic boundary detection algorithm
366 to find and track the endocardial border. The high temporal resolution allowed detecting changes in
367 phases during the myocardial motion. The high spatial resolution together with up-sampling and time
368 delay estimation increased the accuracy of the velocity estimates, showing very little drift through the
369 cardiac cycle. The presented study demonstrates the feasibility of the measurement system and the layer
370 tracking method, with emphasis of the technological solution. The main purpose of this study was to
371 develop and investigate the technology, algorithms and the method, and no conclusions about the
372 clinical usefulness are drawn from this study.

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465 protection of animals used for scientific purposes Text with EEA relevance. 276, 32010L0063
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467 **Figure captions**

468 **Figure 1:** The open chest porcine experiment. Two sensors were sutured to the epicardial surface of the
469 left ventricle in the apical region and in the basal region. In order to reduce number of animal
470 experiments the experimental protocol also included attachment and testing of accelerometer sensors
471 as part of a different study.

472 **Figure 2:** Schematic illustration of the main parts of the measurement system. Two transducers sutured
473 to the heart are connected to the analog ultrasound transmit-receive system. The signals from this are
474 digitized and transferred to a computer. ECG and pressure catheter signals are sampled simultaneously,
475 and synchronized with the ultrasound recordings using pulses from an analog output (DAC).

476 **Figure 3:** Flow-chart illustrating the velocity estimation. The RF-lines were filtered and up-sampled 10
477 times, and cross-correlation was applied to find the time-shift giving the velocity as function of depth.

478 **Figure 4:** Diagram of boundary detection method based on intensity of M-mode image. The first
479 estimate for the endocardial border was found from a fuzzy logic technique and boundary search steps.
480 This estimate was refined and smoothed by a maximum filter along the time, removing outliers based
481 on maximum velocity, and applying snake algorithm in the time direction.

482 **Figure 5:** Illustration of the fuzzy logic definitions. A pixel is associated a value 0, 'black' or 1, 'white'
483 depending on the grey level of the M-mode image relative to the thresholds a and b.

484 **Figure 6:** The range of 20 RF lines after Hilbert transform and log compression mapped to gray scale
485 of range [0 255].

486 **Figure 7:** Gray scale M-mode image with the detected endocardial border. Results before (a) and after
487 (b) removing outliers and applying the snake algorithm. (a): The endocardial border first found from
488 the fuzzy logic algorithm and boundary search (blue) and after applying the maximum filter (red) to the
489 blue curve. (b): Result after removing outliers based on maximum velocity and applying the snake
490 algorithm to the red curve in (a), with no requirements (blue), and requiring monotonous expansion
491 during systole (red).

492 **Figure 8:** Estimated radial velocity inside the myocardium near the apex (top panel) and the base
493 (bottom panel). No velocity filter was applied. The end-systole is marked with blue vertical lines, and
494 the end-diastole with magenta lines. Synchronously measured ECG (blue) and LV pressure (red) are
495 shown below. The ECG recording was of rather low quality, but sufficient for timing.

496 **Figure 9:** M-mode gray scale images with the motion of the detected endocardial border (thick curves)
497 and the four layers (thin curves) inside the myocardium as overlays near the apex (top) and the base
498 (bottom). The curves are calculated from the velocity patterns in Figure 8 by tracking the motion in the
499 forward direction.

500 **Figure 10:** M-mode gray scale images with the motion of the detected endocardial border (thick curves)
501 and the four layers (thin curves) inside the myocardium as overlays near the apex (top) and near the
502 base (bottom). The curves were calculated from the unfiltered velocity patterns in Figure 8 by
503 combining data from tracking the motion in the forward and backward directions, assuming periodic
504 motion.

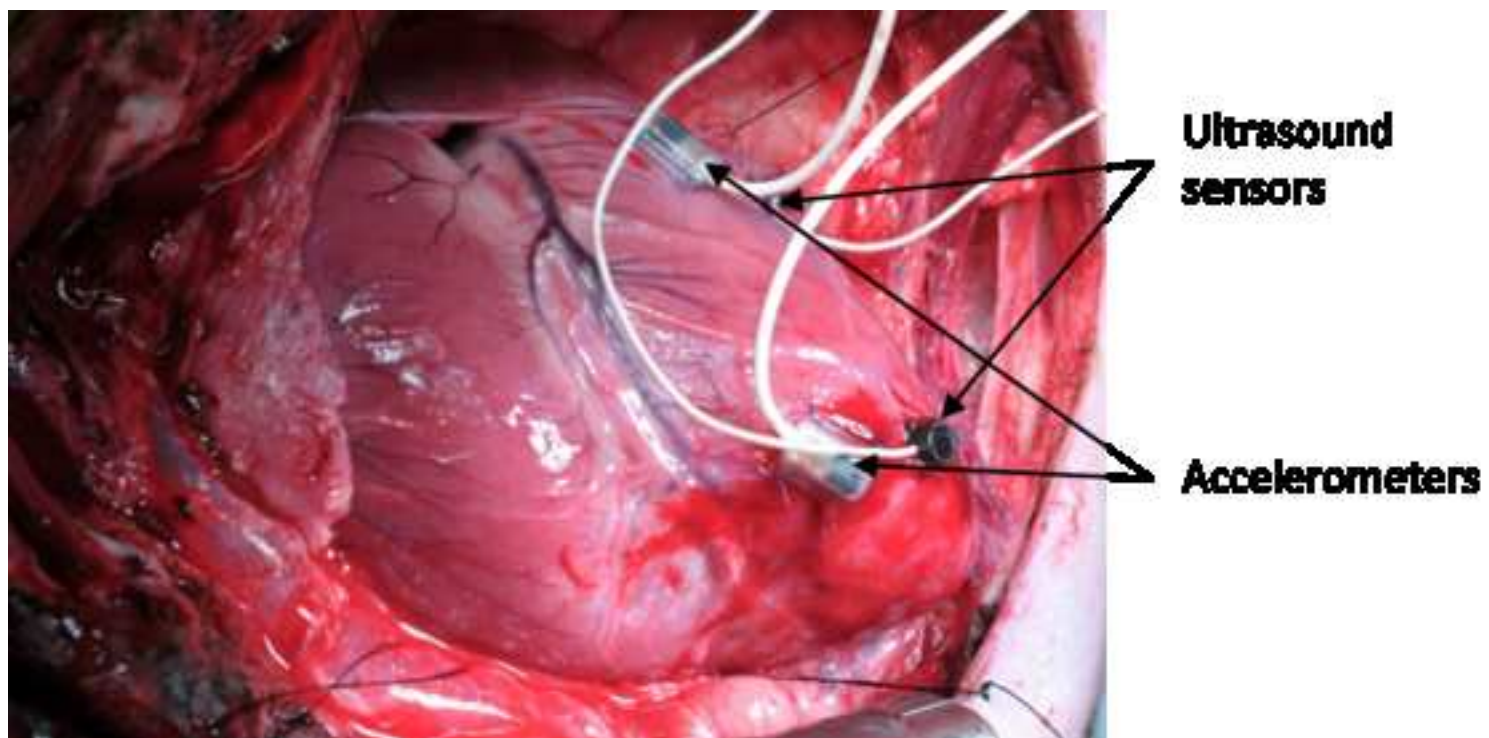
505 **Figure 11:** Estimated radial strain of four layers inside the myocardium near the apex (top) and the base
506 (bottom). Layers are numbered in increasing order from the subepicardial layer 1 to the subendocardial
507 layer 4. The strain curves were computed from the velocity patterns in Figure 8 combining with
508 myocardial layers in Figure 10.

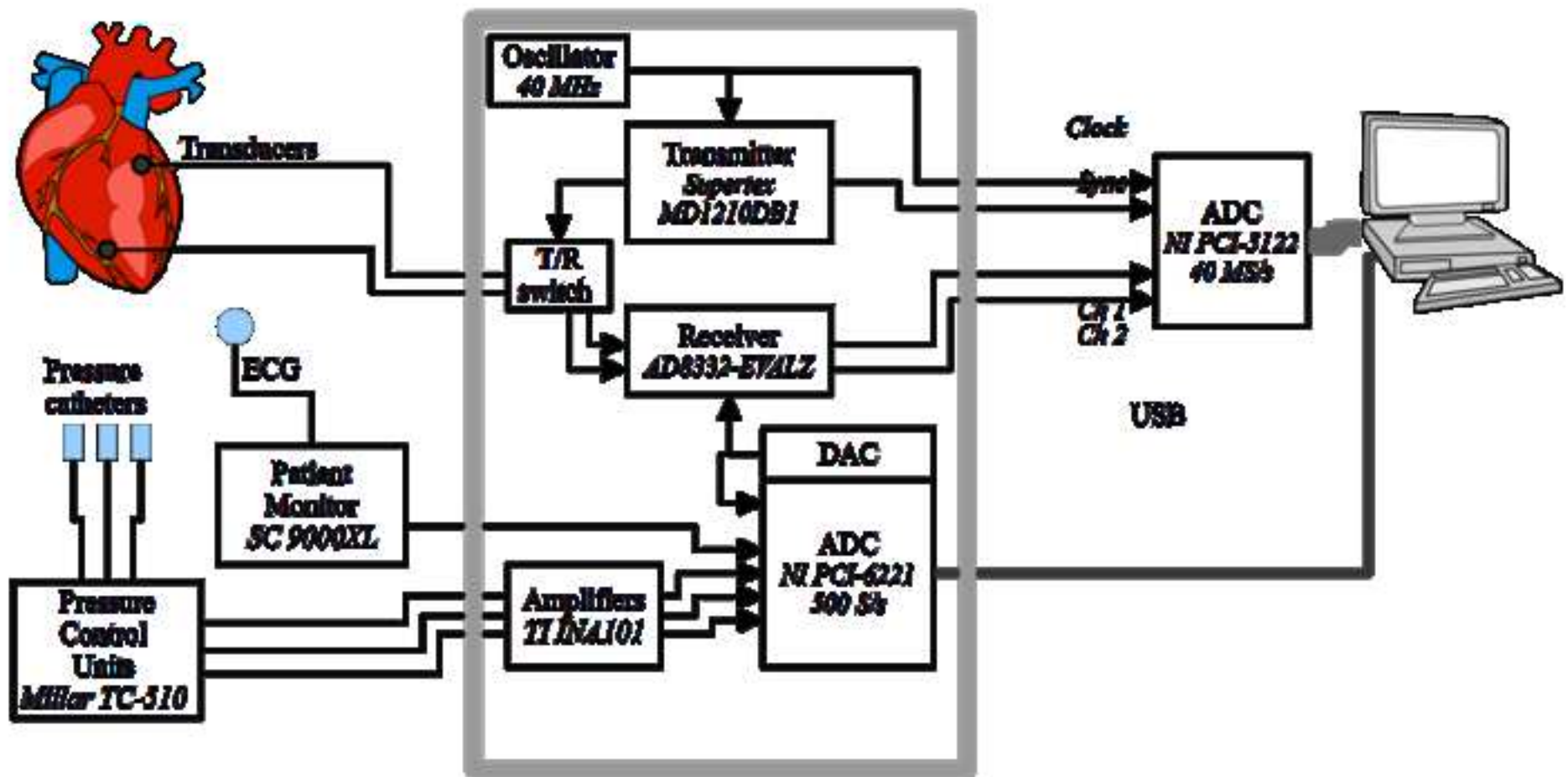
509 **Tables**

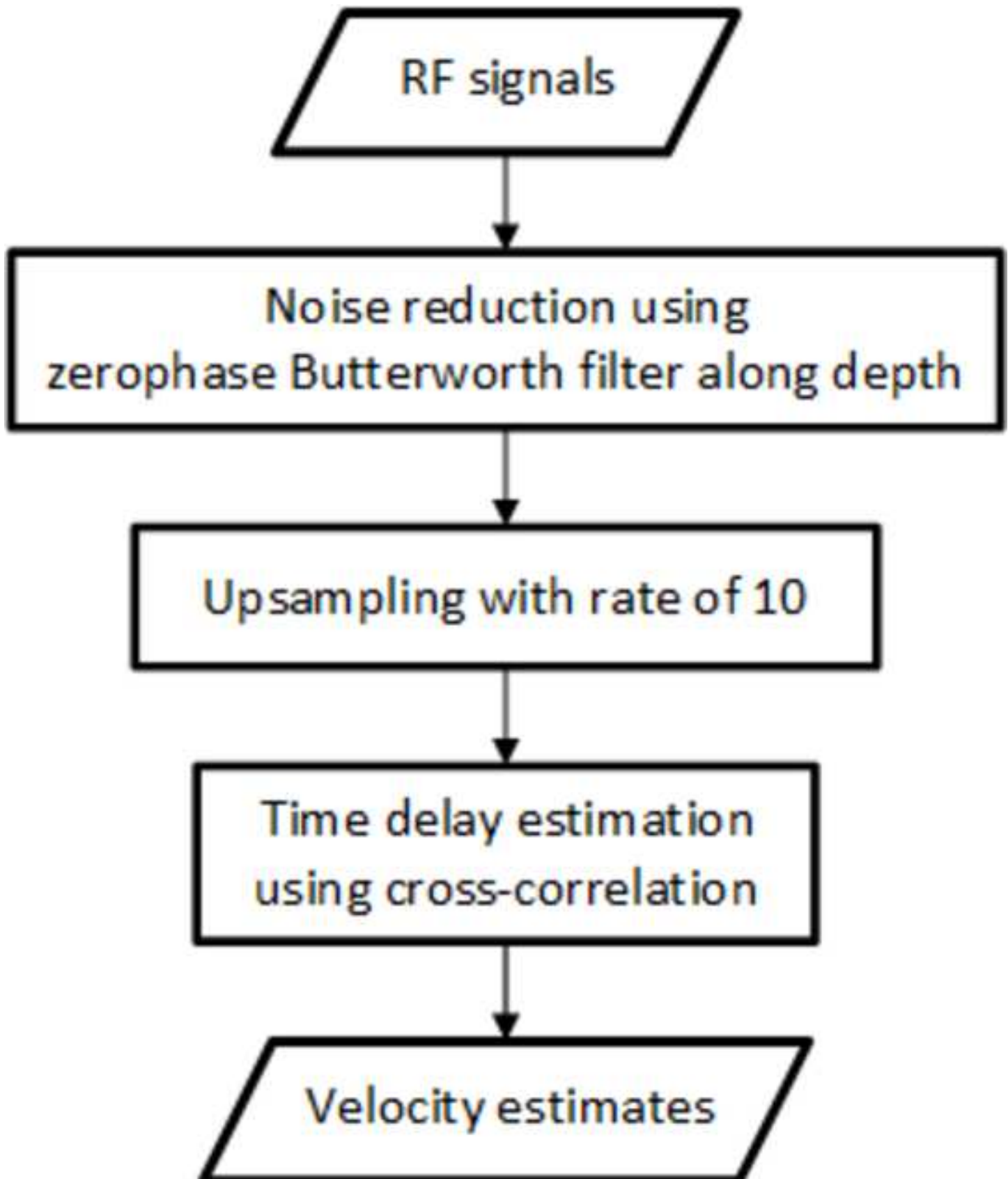
510 **Table 1:** Parameters for endocardial boundary detection for myocardium. Two transducers were used,
 511 ‘Transducer 1’ was positioned near the apex and ‘Transducer 2’ was positioned near the base of the
 512 heart.

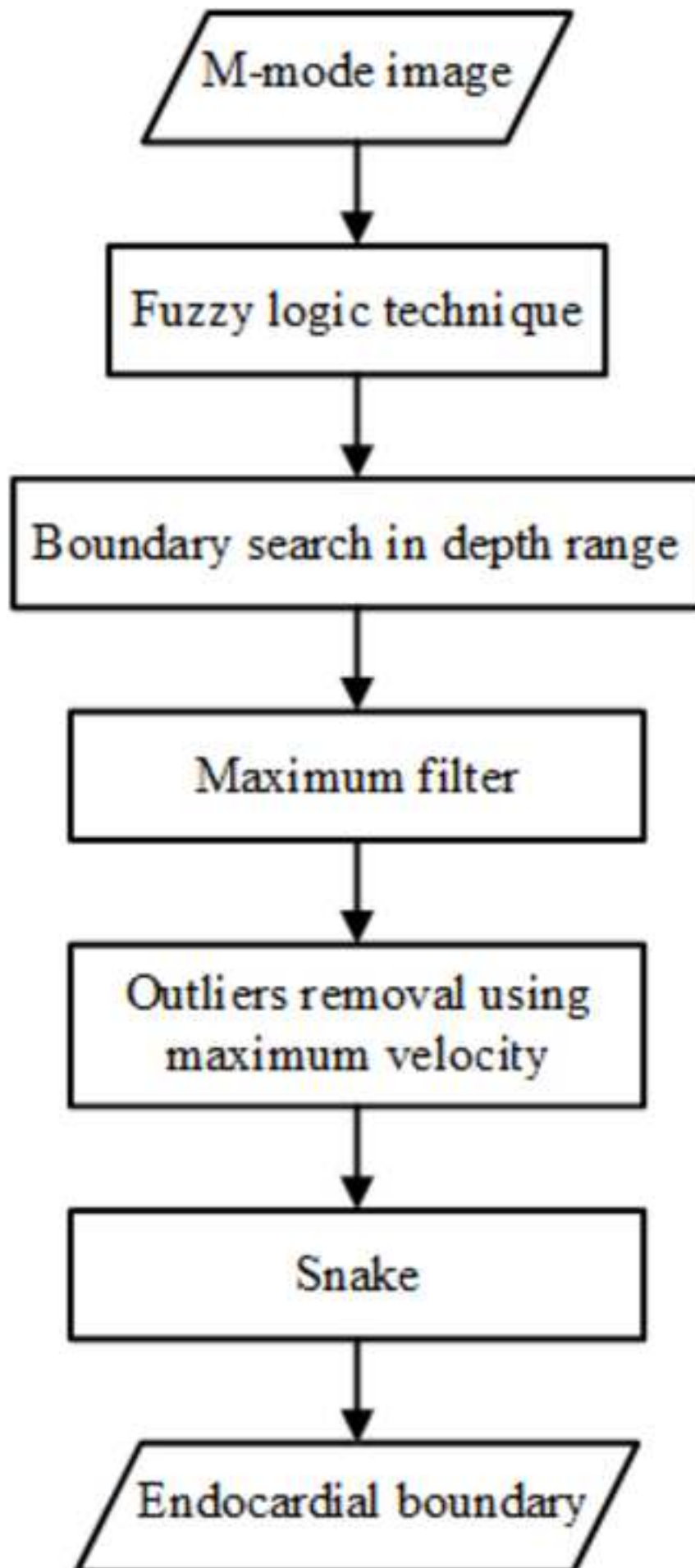
	Transducer 1 Apex	Transducer 2 Base
Thresholds $[a, b]$ for fuzzy logic processing	[158.1 196.35]	[140.25 214.2]
Depth range for searching the endocardial boundary	10 mm to 18 mm	10 mm to 19.98 mm
Snake algorithm parameters		
α	5000	5000
β	0	0
γ	10	10
Iterations	1000	500
Assume monotonous expansion during systole	No	Yes

513

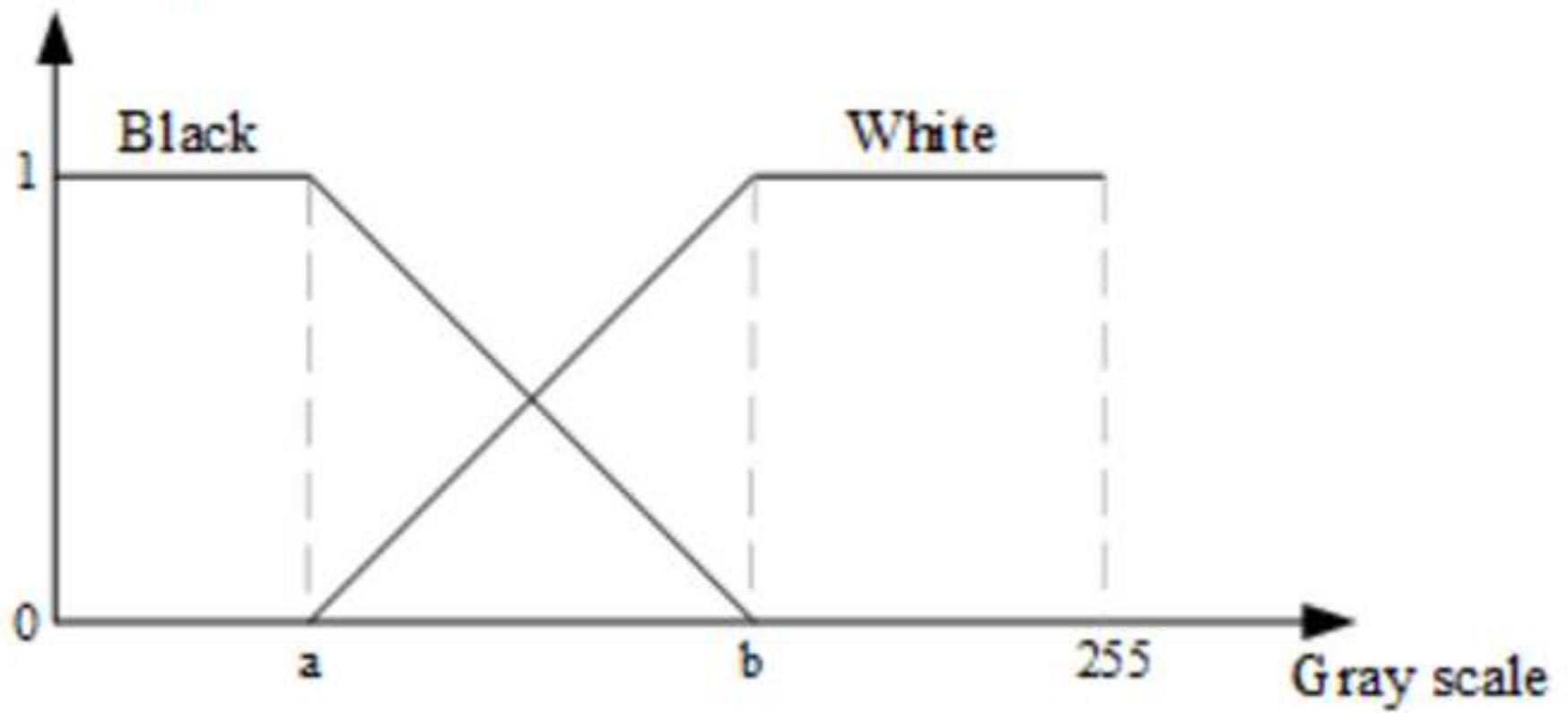


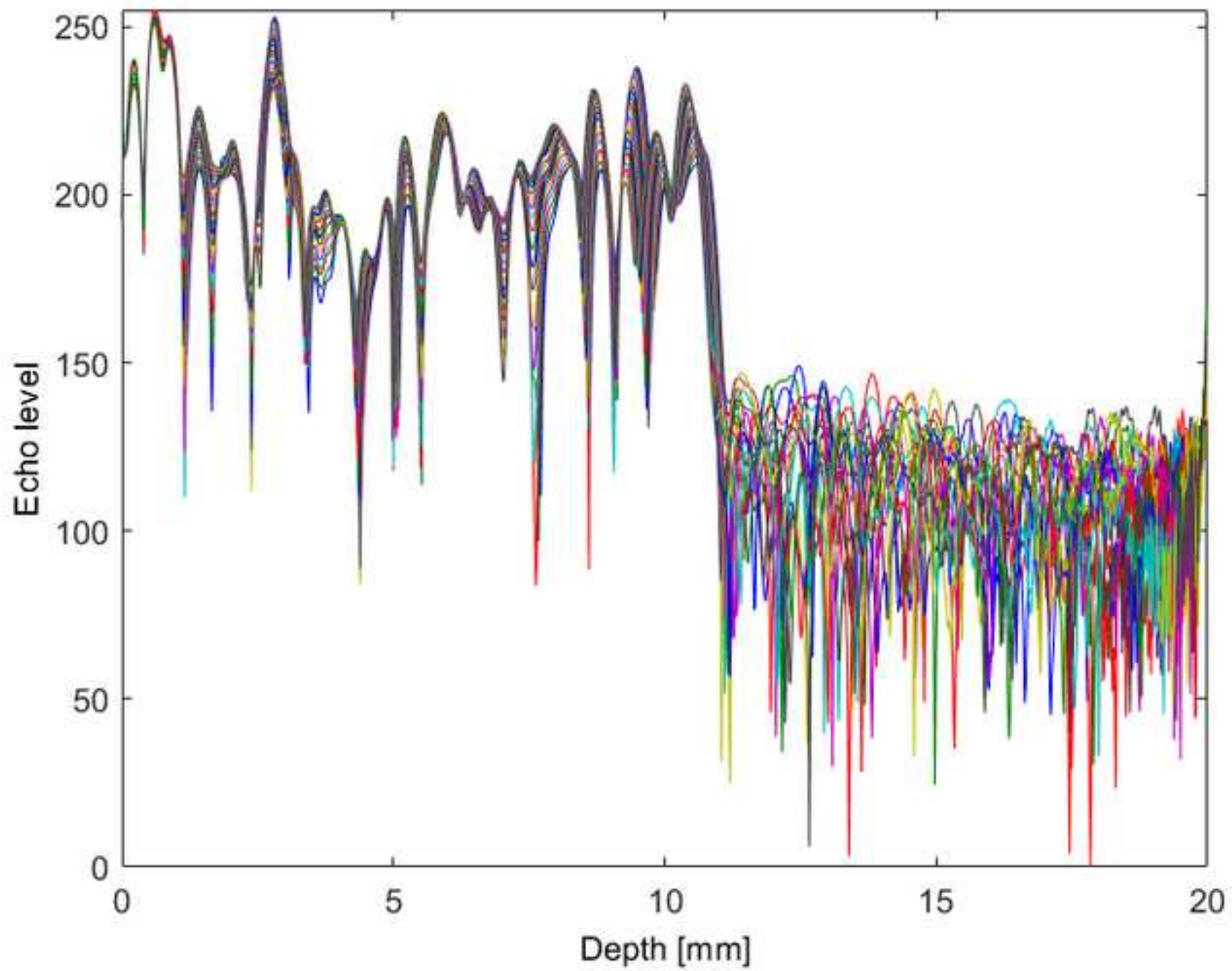


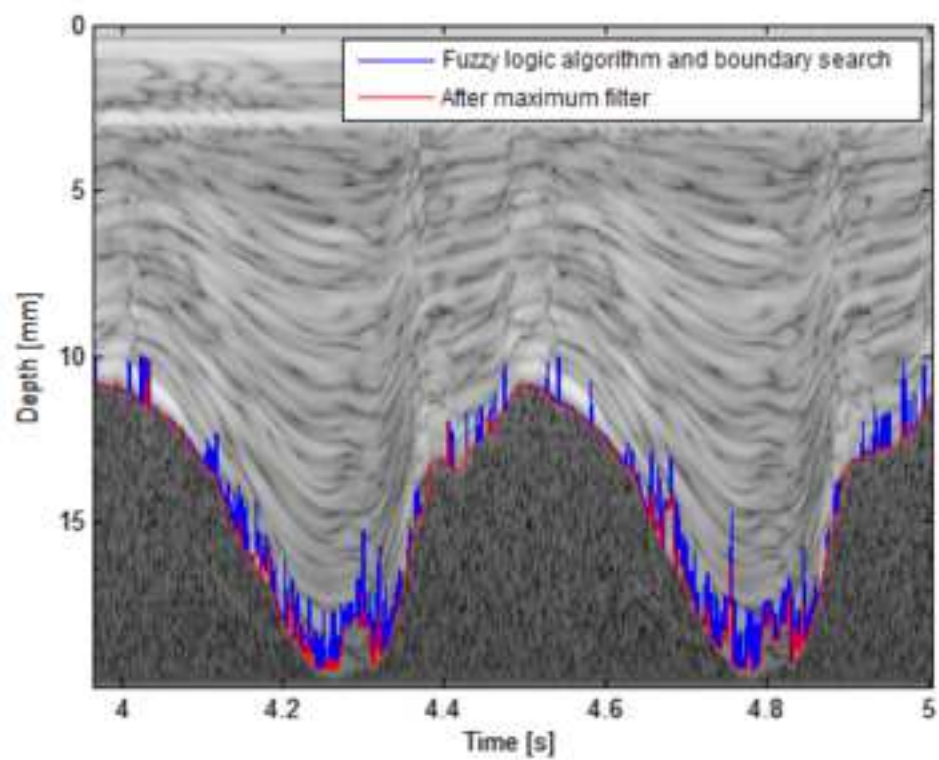




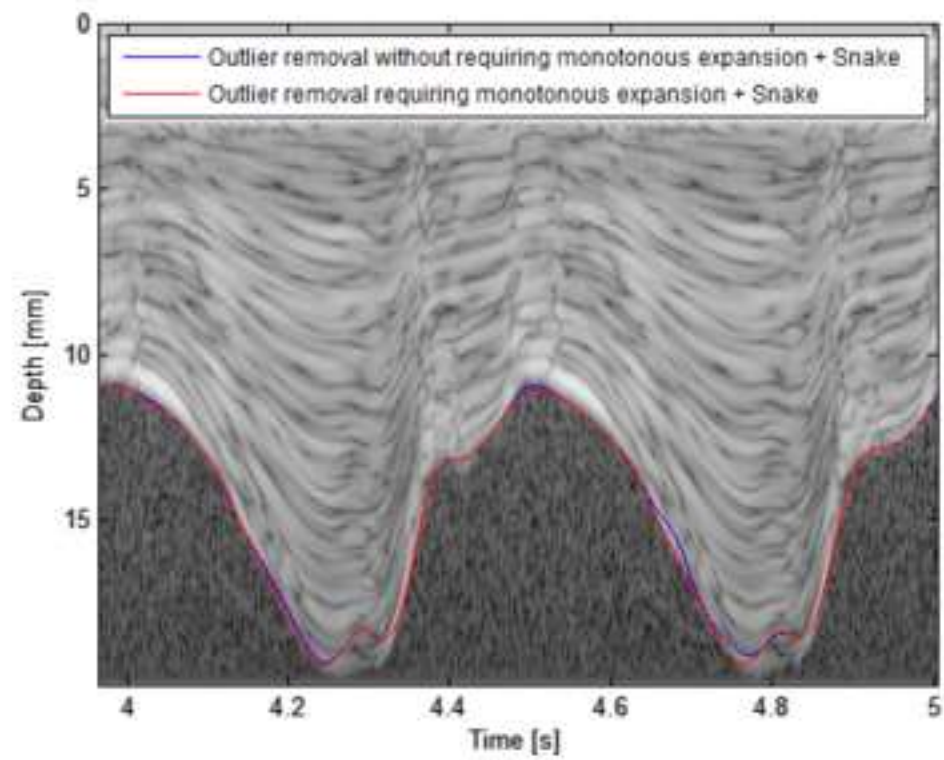
Membership







(a)



(b)

