Parameter Estimation Techniques for Ultrasound Phase Reconstruction

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Presentation Outline

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Motivation

- In human body, the musculature has a great deal to do with the wellbeing and health of an individual
 - Athletic injury
 - Muscular disease
- Quickly diagnosing and thus, preventing muscular disorders would help both patients and medical practitioners

Thesis Objectives

• Accurately measure the phase information of ultrasonic received (RF) signals

 Define an algorithm to reconstruct the less accurate phase outcomes

• Implement the algorithm on the simulated and measured ultrasonic signals captured from experimental phantoms • Among several symptoms, stiffer muscle is a noticeable sign commonly found in different types of muscle disorders

- Muscle strain, muscle cramp, repetitive stress injuries

• Elastography is an imaging technique that could measure tissues' stiffness

Elastography

- Elastography: a method in which stiffness or strain images of soft tissues are measured and used to detect or classify hard parts of the body such as tumors and injured muscles
- Strain: defined as the deformation of the tissue, normalized to its initial shape and is usually shown by "s"



Current Imaging Techniques Used in Elastography



• Ultrasound-based elastography is more commonly used in clinical elasticity imaging

Ultrasound Compression Elastography

- Relies on radio-frequency (RF) ultrasonic signals
- Based on statically compressing the tissue -Particles' movements toward or away from the probe will cause speckle echoes to experience a shift in time

$$r(\hat{\tau}(t)) = \max_{\tau} \int_{-\infty}^{+\infty} \zeta(t_0 + \tau) w(t_0 - t + \tau) . \zeta'(t_0) w(t_0 - t) dt_0$$

$$\hat{d}(t) = \frac{c\hat{\tau}(t)}{2}$$

$$s = \frac{\partial \hat{d}(t)}{\partial z} = \frac{2}{c} \frac{\partial \hat{d}(t)}{\partial t}$$
Times shift \longleftrightarrow Phase shift



Simulation



• Frame of reference: Ultrasonic probe reference frame

- All displacements are relative to the surface of the probe

- Gaussian pulse-echo model
- Pre- and post-compression states
 - Pre-compression RF signal
 - Post-compression RF signal

$$S_r(n) = \sum_{\substack{k=1\\L}}^{L} A_k(n) \cos(\omega_c n T_s - \phi_k(n))$$
$$S'_r(n) = \sum_{\substack{k=1\\k=1}}^{L} A'_k(n) \cos(\omega_c n T_s - \phi'_k(n))$$

Quadrature Phase Detection Technique

• The phase information can be recovered by quadrature detection technique and used in estimation of displacement



Quadrature Technique Performance

- This technique was compared with the phasor method (reference) in different conditions
 - The received RF signal is the sum of echoes (sinusoids) with the same frequency but different phasor parts
 - Phase of the received RF signal would be the angle of its phasor which is equal to the sum of the phasor parts of the constitutive echoes (*phasor addition theorem*)
- Tested Parameters:
 - Signal to noise ratio (SNR) of the received RF signals
 - Bandwidth (B) of the received RF signals
 - Number of Scatterers (L)

• SNR=3dB B=0.8 MHz L=111 (small)



Signal to Noise Ratio (SNR)

• Phase shift error vs. the signal to noise ratio (SNR)

-The error increases with decrease in SNR

• It is desired to find a technique/algorithm by which the less accurate phase outcomes of the quadrature technique can be reconstructed

- SNR parameter

• Inverse problem techniques are able to somehow fix this problem

Inverse Problems

• The task where the values of some model parameters (m) must be obtained from the observed data (d)

 $\mathbf{G}(\mathbf{m}) = \mathbf{d}$: given d, m is aimed to be estimated

Mathematical model (system of equations)

• In case of having overdetermined/underdetermined systems of equations, least squares solution would be estimated

$$\min_{\mathbf{m}} \|\mathbf{d} - \mathbf{G}(\mathbf{m})\|_2 = \sqrt{\sum_{i=1}^m (d_i - \mathbf{G}(\mathbf{m})_i)^2}$$

Inverse Problems Cont.

- Inverse problems are often ill-posed (ill-conditioned)
 - Regularization techniques (Tikhonov)

$$\begin{array}{ccc} \min_{\mathbf{m}} & \|\mathbf{G}(\mathbf{m}) - \mathbf{d}\|_{2}^{2} + \mu^{2} \|\mathbf{\Gamma}\mathbf{m}\|_{2}^{2} \\ & & \searrow \text{ regularization parameter} \end{array}$$

• Nonlinear least squares problem can be solved by iterative algorithms such as Gauss-Newton (GN)

$$\begin{cases} (\mathbf{J}(\mathbf{m}^k)^T \mathbf{J}(\mathbf{m}^k) + \mu^2 \mathbf{\Gamma}^T \mathbf{\Gamma}) \Delta \mathbf{m} = -\mathbf{J}(\mathbf{m}^k)^T (\mathbf{G}(\mathbf{m}^k) - \mathbf{d}) - \mu^2 \mathbf{\Gamma}^T \mathbf{\Gamma} \mathbf{m}^k \\ \\ \text{where} \quad \Delta \mathbf{m} = \mathbf{m}^{k+1} - \mathbf{m}^k \end{cases}$$

Inverse Problem Defined in This Work

• For each RF signal two mathematical models were defined

 $\begin{cases} \mathbf{G}_1(\mathbf{m}) = \mathbf{d}_1 \\ \mathbf{G}_2(\mathbf{m}) = \mathbf{d}_2 \end{cases}$

- d₁ and d₂: in-phase and quadrature parts of the complex baseband signal obtained via quadrature method
- m: vector of phase and amplitudes of the received RF signal at different depth sample numbers
- This problem is a nonlinear least squares problem which should be solved by the GN algorithm

Phase Reconstruction by GN Algorithm

- Phase shift between the first two RF signals
- SNR= 3dB
- $\mu_1=24, \mu_2=25$
- Iteration numbers:

1, 8



Reference method Quadrature method

Error Reduction by GN Algorithm

 Quadrature phase shift error (ε₁) and reconstructed phase shift error (ε₂) vs. the signal to noise ratio (SNR)



Phantoms and Results

 Accumulated phase shift after generation of 5550 RF signals (frames) in M-mode operation





Conclusion and Future Work

Conclusion

• Defined a novel approach for ultrasound phase reconstruction by means of GN algorithm

- The algorithm acted as a filter and was able to remove noise

- In designing the algorithm, optimized regularization parameters were selected
 - Extremely small or large values reduced the effectiveness of the algorithm

Future Works

• Finding a general approach for regularization parameter selection, improving the algorithm to test greater number of iterations and testing other parameters affecting the quadrature method are left as the future work



Questions ?

Time or Phase Shift Measurement

- Tissue compression causes the speckle echoes and in turn, the received RF ultrasonic signals to experience a shift in time. This time shift changes the phase information of corresponding signals resulting in a phase shift between them.
- Resulted time or phase shift can be estimated and used for displacement estimation in one of the two approaches of:
 - Estimating the time shift between small windows by cross correlation function



 Estimating the phase shift between RF signals by means of a phase measurement technique such as quadrature phase detection technique

$$\Delta d_z(n,m) = \frac{c\Delta\varphi_z(n,m)}{4\pi f_c}$$

 Initial setting: SNR=40dB B=0.8 MHz L=111 (small)



Consecutive Phase Shift between Frames 1 and 2

 SNR=40dB B=0.8 MHz L=1055 (large)



- Phase shift error vs. the number of scatterers (L)
 - -By changing the number of scatterers, the error remains fix in a range between 0 to 15 rad.

SNR=40dB ۲ B=2.25 MHz L=111 (small)



2.5

3.5

3 RF Signal Bandwidth 4.5

Consecutive Phase Shift between Frames 1 and 2

Phase shift error vs. the RF • signal's bandwidth (B)

-As the bandwidth of echoes increases, the error increases

18L

Regularization Parameter Selection

- First RF signal
- SNR= 3dB
- Reconstructed phase error (ε₂) vs. the regularization parameter and iteration number.



Regularization Parameter Effect

• Phase shift between the first two RF signals

Reference methodQuadrature methodGN algorithm

• SNR= 3 dB



Phantoms and Results

• Phase shift between the first two consecutive RF signals (frames) in M-mode operation





Pulse-Echo Mathematical Representation

$$P(t) = \cos(\omega_c t)$$

$$S_k(\tau_k : t) = \alpha_k e^{-\beta(t-\tau_k)^2} \cdot P(t - \tau_k = \alpha_k e^{-\beta(t-\tau_k)^2}) \cdot \cos(\omega_c t - \omega_c \tau_k)$$
 Echo Signal

$$S_r(n) = \sum_{k=1}^L A_k(n) \cos(\omega_c n T_s - \phi_k(n))$$

Received Signal

In-phase and Quadrature Signals

$$I(n) = LPF\{S_{r}(n) . \cos(\omega_{dem}nT_{s})\}$$

$$= LPF\{\left[\sum_{k=1}^{L} A_{k}(n) \cos(\omega_{c}nT_{s} - \phi_{k}(n))\right] \cos(\omega_{dem}nT_{s})\}$$

$$= \sum_{k=1}^{L} \frac{A_{k}(n)}{2} \cos[\Delta \omega nT_{s} - \phi_{k}(n)] \longrightarrow \text{In-phase Signal}$$

$$Q(n) = LPF\{S_{r}(n) . (-\sin(\omega_{dem}nT_{s}))\}$$

$$= LPF\{\left[\sum_{k=1}^{L} A_{k}(n) \cos(\omega_{c}nT_{s} - \phi_{k}(n))\right](-\sin(\omega_{dem}nT_{s}))\}$$

$$= \sum_{k=1}^{L} \frac{A_{k}(n)}{2} \sin[\Delta \omega nT_{s} - \phi_{k}(n)] \longrightarrow \text{Quadrature Signal}$$

$$u(n) = I(n) + jQ(n), \qquad \angle u(n) = \arctan\left(\frac{Q(n)}{I(n)}\right) = \arctan\left(\frac{\sum_{k=1}^{L} \frac{A_{k}(n)}{2} \sin[\Delta \omega nT_{s} - \phi_{k}(n)]}{\sum_{k=1}^{L} \frac{A_{k}(n)}{2} \cos[\Delta \omega nT_{s} - \phi_{k}(n)]}\right)$$
Complex Baseband Signal Phase Information

Complex Baseband Signal Phase Information

RF Signal Waveform Shape & Phase Signal

