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Empirical Optimization Of Frequency Parameters In Chirp Inversion Imaging

Amira Zaylaa, Sébastien Ménigot, Jean-Marc Girault and Jamal Charara

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Contrast ultrasound imaging has become a significant imaging technique for the last ten years. Physiological and pathological information are accessible thanks to microbubble injection and nonlinearity detection. To improve resolution and penetration, frequency modulated signals called chirps combined with compression filter have been utilized by radar technology. Recent studies applied the use of chirps in harmonic imaging. And since the setup parameters of the chirp are important to enhance the microbubble nonlinearities and thus the contrast; yet their full potential is not tapped. The present study aimed to empirically optimize the harmonic energy backscattered by microbubbles through a polynomial frequency modulation law of order three. Two imaging techniques were considered: harmonic chirp imaging and chirp inversion imaging. In simulation, the frequency parameters of chirps were sought for microbubbles of different sizes (1 µm and 2.5 µm). It was beneficial to increase the order of the polynomial chirp to improve the contrast. Chirp Inversion Imaging was preferable over harmonic chirp imaging using a polynomial chirp for contrast enhancement. Moreover, optimal frequency parameters were selected manually. Optimization of polynomial chirps was empirically capable of enhancing the contrast in images however, this study is paramountly conceived by an automatic technique.

1 Introduction

Contrast Ultrasound Imaging (CUSI) is a significant imaging technique requiring the injection of a gas-containing contrast agent; it is possible because microbubbles resonate when impinged upon by ultrasound and emit ultrasound at harmonic frequencies. Ultrasound harmonic images are produced by anatomic structures that contain these agents. Tissues that do not contain gaseous contrast agents presumably produce negligible harmonic signals. The contrast between the medium infused with microbubbles and the surrounding tissues is given by the contrast-to-tissue ratio (CTR) [1, 2].

Ultrasonic Contrast agents (UCA) are used in ultrasound imaging to increase the sensitivity of the imaging technique by altering the image contrast between different structures. A dramatic increase in scattering which can be achieved with this microbubble has made it useful both in difficult-to-image patients and in areas for which the ultrasound signal is reduced. Injection of the contrast microbubbles and observation of their real-time passage in an ultrasound image can significantly improve the sensitivity and diagnostic power of ultrasonic imaging. Basically, Harmonic Imaging can be achieved by various acquisition techniques such as second harmonic acquisition and pulse inversion.

Most applications of harmonic imaging implicates that the spectrum of the backscatter contains not only the fundamental but also the harmonics. Therefore, by second harmonic imaging a bandpass filter selects the second harmonic, however, the bandpass filter cannot avoid totally the spectral overlap between the fundamental and the harmonics. Conversely, by pulse inversion imaging the first lobe is suppressed totally and we are left with the harmonics and/or subharmonics [3]. Regardless of the imaging technique, whether second harmonic or pulse inversion imaging, one of the major problems that ultrasonic images suffer from is bad contrast, low resolution and small penetration. To improve both resolution and penetration, frequency modulated signals known as “Chirps” often utilized in Radar technology were recommended in CUSI [1, 4]. Although the use of these chirps is capable of providing clinicians with images having good resolution and penetration in the future, the question remains what about the CTR?

Chirp imaging constitutes of sending frequency modulated signal, chirp excitation, into the medium under exploration, receiving the backscattered signal and filtering it in order to produce the image (Figure 1). The chirp could be linear or nonlinear; it depends on the frequency modulation profile or law. Traditional harmonic chirp imaging (HCI) has been utilized to reconstruct images at the second harmonic after transmitting chirps to the medium in order to improve the resolution and contrast.

Recent studies have utilized pulse inversion imaging to improve the CTR by using conventional pulses [5], however, conventional pulses have led a tradeoff between resolution and penetration. Others have used chirp inversion imaging (CII) based on a linear chirp to improve the signal-to-noise ratio (SNR) and solve the tradeoff between resolution and penetration [1, 6]. Exponential chirp was utilized to investigate nonlinear systems in audio signaling [7] and hasn’t yet been investigated in CUSI. Nonlinear chirp of order two has been optimized by HCI [5] merely. For that, optimization of the chirp’s frequency law became profoundly essential generally due to the relative safety of UCA and particularly due to the robustness of chirp imaging in enhancing penetration, resolution and the CTR of medical images. The optimization of such chirps in diagnostic and therapeutic ultrasound imaging is still flourishing and not until now did anyone investigated their full potential of chirps.

In this paper we aimed to empirically optimize the frequency modulation parameters of a third order polynomial chirp using Chirp Inversion Imaging (CII) in contrast media, i.e. instead of pulses in [8] we used chirps, with a small
overview on harmonic chirp imaging (HCI) in order to enhance the contrast of contrast ultrasound images. We proposed an algorithm for CII and implemented HCI, in addition, we employed the compression filter in the two imaging techniques. Furthermore, since the energy or the power of the signal is related to the CTR, we computed the energy of the signal backscattered by 1 µm and 2.5 µm micro-bubbles during their nonlinear response and focused our study on improving these energy outcomes i.e. improving indirectly the CTR. Under this condition we anticipated that we not only get an image with good resolution and penetration but also with enhanced contrast that is capable of providing amended clinical diagnosis in the future. We tended to optimize a polynomial chirp of order higher than currently optimized chirps. The novelty of our approach was based on utilizing a polynomial chirp of order three and applying chirp inversion imaging in CUSI to obtain primary ultrasonic medical images of high contrast and consequently discriminate between regions that were not easily differentiated.

2 Methods And Materials

We simulated chirp imaging with Matlab (Mathworks, Natick, MA, USA). First, frequency modulated (FM) signals having the form of Eq. (1) were transmitted into the medium infused with contrast bubbles by means of a transducer using HCI and/or CII.

\[
C(t) = G(t) \cos(2\pi f_1(t)t),
\]

where \( f_1 \) is the instantaneous frequency, \( t \) the time and \( G(t) \) the gaussian envelope defined by:

\[
G(t) = A \exp\left(-\frac{t^2}{\sigma^2}\right).
\]

where \( A \) is the normalized amplitude of the chirp, \( \alpha \) the shift of the envelope from the center and \( \sigma \) the width of the gaussian’s envelope.

The instantaneous frequency we used in Eq. (3) also known as FM law was governed by a polynomial of order three:

\[
f_1(t) = f_o + \beta_1 t^2 + \beta_2 t^3
\]

where \( f_o \) is the starting frequency of the chirp in (MHz) and \( (\beta_1, \beta_2) \) were the modulation indices in (THz/s) and (THz/s^2) respectively. In particular, when \( (\beta_2 = 0) \) the chirp is said to be a Quadratic FM, otherwise it is a Polynomial FM [9, 10, 11].

Second, chirps made their way towards the medium. This resulted in backscattered signals which were further detected by the same transducer and fed into a compression filter to be finally interpreted and transformed into an energy image. Third, for different microbubble sizes and chirp types several simulations were carried out to optimize the frequency parameters of chirps. We implemented the compression filter as proposed in [12]. As compared to a regular compression filter, this harmonic compression filter has twice the instantaneous frequency at every time point. The actual chirp \( C(t) \) was given by Eq. (1), the filters’ impulse response was defined by:

\[
h(t) = [G(t)]^4 \cos(2\pi f_1(t)t)
\]

with \( f_1(t) \) given by Eq. (3) and the output of the filter was given by \( y(t) \):

\[
y(t) = C(t) \ast h(t)
\]

which described the compressed echo with good axial resolution and a central frequency situated at the second harmonic of the transmitted chirp. Also we looked up for the highest energy values in order to enhance the contrast since the energy is directly proportional to the CTR and is more accessible.

2.1 Harmonic Chirp Imaging

It is a classical technique that permits the selection of the backscattered signal whose central frequency lies within the bandwidth of the transducer and is equal to the second harmonic value. This is equivalent to bandpass filtering. The backscattered signal is bandpass filtered such that the second harmonic is selected and the image is ready to be reconstructed at \( 2f_o \) [2].

In HCI algorithm we:

- Specified the bandwidth of the transducer
- Fired a quadratic chirp \( C(t) \) given by Eq. (1)
- Simulated the backscatter \( e(t) = Echo(C(t)) \) by solving the Modified Rayleigh-Plesset differential equation
- Fed \( e(t) \) into the compression filter and computed the energy image of the compression filter’s output \( y(t) \) for different frequency values \( f_o \) as a function of \( \beta_1 \)
- Selected manually the optimum \( f_o \) and its corresponding \( \beta_1 \)

We further extended the algorithm to send a polynomial chirp \( (\beta_2 \neq 0) \) receive the backscattered signal, feed it into the compression filter and compute the energy image of the backscattered signal, but here, for different \( \beta_2 \) values as a function of \( \beta_1 \). We finally selected the best \( \beta_2 \) and its corresponding \( \beta_1 \) manually from the image.

2.2 Chirp Inversion Imaging

It is the process of transmitting two pulses 180° out of phase sequentially into the medium infused with microbubbles. The received echoes are summed up such that echoes at the fundamental frequency will cancel out yet those at the harmonic frequencies will show up [2].

In CII algorithm we proposed that:

- Two quadratic chirps \( C_1(t) \) and \( C_2(t) \) having the form of Eq. (1) with \( C_2 \) satisfying equation (6) are sent in such a way that a chirp was transmitted into the medium of micro-bubbles and, after a suitable delay \( t_0 \), a second pulse was transmitted that is an inverted copy of the first pulse.

\[
C_2(t) = -C_1(t - t_o)
\]
where \( C_1(t) \) and \( C_2(t) \) represents the acoustic pressure of the first and second pulses as a function of time, \( t \).

- These signals were backscattered separately by micro-bubbles. The echo of each firing was then simulated:
  \[
  \begin{align*}
  e_1(t) &= \text{Echo}\{C_1(t)\} \\
  e_2(t) &= \text{Echo}\{C_2(t)\}
  \end{align*}
  \]  
  (7)

- Then the following pulse inversion detection signal was manipulated:
  \[
  s(t) = e_1(t - t_0) + e_2(t)
  \]  
  (8)

- \( s(t) \) was fed into the compression filter and the energy of its output \( y(t) \) was computed for various central frequency values as a function of \( \beta_1 \)

\[\text{Figure 2: The transmitted polynomial Chirp.}\]

- The optimum \( f_0 \) is selected manually

\[\text{Figure 3: The energy of the compressed polynomial FM backscatter in HCI for 2.5 µm bubble.}\]

Moreover, the previous quadratic chirp was extended to a polynomial one having \( \beta_2 \neq 0 \) and a fixed \( f_0 \) value, as previously described, the resultant backscattered signal was computed and led to a specific energy image and optimum \( \beta_2 \) and \( \beta_1 \) values after filtering.

## 3 Results and Discussion

In this part we will show the simulation results of the emitted quadratic and polynomial FM, particularly, the energy of the backscattered signal during the nonlinear response of micro-bubbles by traditional HCI and CII. We will show also the effect of the chirp’s frequency parameters as well as discuss the results of the energy backscattered from 1 µm and 2.5 µm bubbles in HCI and CII.

### 3.1 Simulated Emitted Chirp

The chirp excitation given by Eq. (1) was depicted in Figure 2, the upper plot manifested the temporal representation of the third order polynomial chirp insonating the medium of micro-bubbles, the image exhibited its spectrogram with the energy indicated by the colorbar in an ascending order and the left plot represented merely its spectrum. The spectrogram topologically represented a small region having significant energy values. The highest energy value occurred at \( t=1 \mu s \) when the chirp attained its maximum amplitude.

### 3.2 The Influence of Chirp’s Frequency Parameters

In concern of the effect of the frequency parameters we simulated Figure 3 and Figure 4. Figure 3 represents the energy of the signal backscattered from 2.5 µm bubble by HCI and Figure 4 represents the energy of the resultant signal backscattered by the same bubble by CII. For 2.5 µm bubble when insonifying the medium by a polynomial chirp of order 3 with \( (\beta_1=0) \) it was vivid that the backscattered energy relied on both \( \beta_2 \) and \( f_0 \) in HCI and CII.

\[\text{Figure 4: The energy of the compressed and backscattered polynomial FM in CII for 2.5 µm bubble.}\]

<table>
<thead>
<tr>
<th>Frequency Parameters</th>
<th>( \beta_1=0 )</th>
<th>( \forall \beta_1 )</th>
<th>( \beta_1=0 )</th>
<th>( \forall \beta_1=0 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \beta_2=0 )</td>
<td>-31.6</td>
<td>-31.5</td>
<td>-30</td>
<td>-29.2</td>
</tr>
<tr>
<td>( \beta_2=0 )</td>
<td>-31.6</td>
<td>-31.5</td>
<td>-30</td>
<td>-29.2</td>
</tr>
</tbody>
</table>

In the range tested, optimal values of \( \beta_2 \) and \( f_0 \) existed in the regions where the best backscattered energy occurred (dark red color). These optimal values were different in HCI and CII. The highest backscattered energy was attained in CII. The same remark was valid when transmitting a quadratic chirp, and studying the variation of \( (f_0) \) as a function of \( \beta_1 \) in HCI and CII techniques. Moreover, we noticed that lines \( \beta_1=0 \) and \( \beta_2=0 \) in (Figures 3, 4) were not the axis of symmetry this implied that the frequency parameters were asymmetric.

Many believed that nonlinear chirp signals were designed to minimize the peak range sidelobe level by optimizing its
frequency sweep rate and time duration [6]. The major finding of our study manifested the strong impact of the FM law which could be quadratic or polynomial on the energy of the backscattered signal. Our data (see Table 1) exhibited that the higher the degree of the polynomial in FM law, the higher the energy was carried by the backscattered signal. In other words, as the FM law was more curved the energy received was dramatically higher.

From Table 1 we inferred that by increasing the degree of the FM law from: 1 to 2 an improvement of 0.1 dB for CII and 0.6 dB for HCI was observed, from: 1-3 an improvement of 1.6 dB for CII yet 0.0 dB for HCI was obtained and from: 2-3 \((\beta_1 \neq 0, \beta_2 \neq 0)\) 2.3 dB for CII and 0.1 dB for HCI was attained. The best results were those governing a third order polynomial governed by equation (2) using CII. However, in HCI the improvement of the Energy-to-complexity ratio was low as compared to that obtained for CII.

Table 2: The variation of the maximum energy and the modulation indices as a function of the radius (2.5 \(\mu\)m,1 \(\mu\)m) of bubbles and the imaging method (HCI,CII) in the frequency range \(f=\) [0.5 MHz, 6 MHz].

<table>
<thead>
<tr>
<th>Radius ((\mu)m)</th>
<th>2.5</th>
<th>1</th>
</tr>
</thead>
<tbody>
<tr>
<td>ImagingMethod</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Max(E) (dB)</td>
<td>-29.2</td>
<td>-36.3</td>
</tr>
<tr>
<td>(f_0^*)</td>
<td>2.75</td>
<td>3.3</td>
</tr>
<tr>
<td>((\beta_1)^*)</td>
<td>1.2</td>
<td>34.3</td>
</tr>
<tr>
<td>((\beta_2)^*)</td>
<td>-20410</td>
<td>-0.8</td>
</tr>
</tbody>
</table>

3.3 The Influence of Microbubble’s Size

We impinged two bubbles of radius 2.5 \(\mu\)m and 1 \(\mu\)m respectively by a third order polynomial chirp and summarized the results in Table 2. Table 2 represented the optimum frequency parameters \(f_0^*\), \((\beta_1)^*\) and \((\beta_2)^*\) in the two imaging modalities utilized and for 2.5 \(\mu\)m and 1 \(\mu\)m bubbles. In summary, the maximum energy values, when the signal was backscattered by 1 \(\mu\)m bubble, were much imperative than those when the signal was backscattered by 2.5 \(\mu\)m bubble. In addition, it was advantageous to use CII over CHI, since the backscattered energy obtained with CII is always superior of 7 dB as compared to CHI technique.

The optimal values depicted in Table 2 led to the best backscattered energy that were not only dependent on the degree of the FM law but also on the imaging technique. Consequently, searching for the best imaging technique and the best FM law are crucial. Note however that the optimal value of the frequency parameters varied with both the transmits’ frequency \(f_0\) and the size of the microbubble. Nevertheless, it was not empirically demonstrated in this study, it was obvious that the optimal value of each parameter depended not only on the transmits’ frequency \(f_0\) and the size of the microbubble but also on other parameters such as the bandwidth of the transducer and the initial pressure level.

4 Conclusion

Our study manifested that accounting for both the chirp type and the size of the micro-bubble it was necessary to investigate the frequency parameters that govern the behaviour of time-frequency transmit signal to optimize the energy reflected by microbubbles and indirectly the CTR. The third order Polynomial chirp insonation using CII have provided the optimal and most promising results as compared to CHI during the nonlinear response of microbubbles. Using the robust CII algorithm and a third order polynomial several optimum frequency parameters that tended to enhance the CTR were selected manually. Indeed the improvement of the CTR opened the door for high quality images and consequently amended diagnosis. Future step could be the validation of our findings experimentally for improving clinical diagnosis. And since the selection of the optimum parameters in our empirical study was done manually, we ought to consider an advanced automatic technique.

Acknowledgements

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References


