Sidelobe Suppression Featuring the Phase Coherence Factor in 3-D Through-the-Wall Radar Imaging

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Abstract. Through-the-Wall Radar Imaging (TWRI) could be performed by beamforming of the signals from an array of ultrawideband (UWB) antennas. Preferably the imaging is done in 3-D so that both the position and the height of the target are revealed. This is possible with planar array antenna geometry. However, implementing this technique that fulfills Nyquist criterion leads to a large number of antennas while sparse array suffers from an increase of the sidelobe level. In this work, the Phased Coherence Factor (PCF) is applied to conventional delay and sum (DAS) beamformer to suppress the sidelobe of a sparse planar antenna array. The performance of the proposed technique is experimentally evaluated in terms of the target-to-clutter ratio (TCR), and the separation resolution. It is discovered that PCF is effective in reducing the sidelobe's artifacts, resulting in TCR of greater than 10 dB and a separation resolution of 20 cm at 2.5 m range.

Keywords
Through-wall, radar, sidelobe, suppression, sparse array, phase coherence factor

1. Introduction

TWRI is crucial in many practical applications like the search-and-rescue operation, urban sensing, surveillance and tactical missions [1]. Radar technology is usually adapted for this purpose due to the ability of electromagnetic wave to penetrate visually obscure walls. Conventionally, the image is reconstructed using signal processing technique such as beamforming in which the signals received by an antenna array are coherently summed [2].

In achieving higher resolution, an array with large aperture size is usually preferred. However, this requires many sensors in order to fulfill the Nyquist sampling criterion. In sensing, this means the element spacing must be equal or less than $\lambda/2$. Therefore, a large number of antennas is needed in order to meet this criterion especially when the resolution in both $x$ and $y$ directions need to be maintained at relatively the same degree of accuracy. On the other hand, designing an array with fewer sensors with spacing exceeding $\lambda/2$ produces spatial ambiguities in the form of grating lobes [2].

However, for a wideband system, the grating lobes spread over a large region and their amplitudes are significantly lower than the main lobe [3]. Grating lobes also do not build up in a UWB pulse-based system due to the compactness and low number of oscillations of the pulses i.e. the pulses only interfere near the focal point [4]. Unfortunately, sidelobes still exist due to the finite aperture length and finite number of elements [3], [4]. An increase in the sidelobe ratio increases the apparent background level thus decreasing the image contrast [5]. When the beam is steered, a point target will be detected in several consecutive beams, leading to the broadening of the target extent in the crossrange. Meanwhile, the main lobe-to-sidelobe ratio determines the image’s dynamic range [3].

Human sensing behind the wall was investigated by means of the Moving Target Indicator (MTI) utilizing the Change Detection (CD) technique [6], [7]. These procedures eliminate the need to use the scene’s background information, and the image is reconstructed from two consecutive or non-consecutive coherently subtracted radar profiles. However, the residual clutter still exists due to the grating/side lobe artifact [8]. This is in addition to other inherent artifacts like wall reflection [7], and the multipath scattering [9]. Therefore suppressing the grating and side lobes will bring significant enhancement to the performance of a CD based MTI.

In recent years, a lot of research have been performed to minimize grating or sidelobe of antenna arrays either through synthetic operation [10], virtual [11], [12] or fixed aperture arrays [13], [14]. However managing a high number of elements requires a complex system and design. Similarly, a lot of work have also been performed to enhance TWRI’s image via image processing like the use of statistical methods [15], fuzzy logic [16], filtering [17], autofocusing [18], and multi-location image fusion [19]. However, these techniques are less effective since they do not directly address the artifacts caused by the grating and sidelobe interferences. Moreover, most of the previous work employed a scanner based measurement technique,
synthesizing a large number of densely spaced antenna elements, which is impractical for a real implementation due to large data volume, long sampling time and high cost.

A conventional method in reducing the sidelobes is through tapering by introducing weighting factors to each sensor channel in expense to the widening of main lobe’s width and consequently poorer spatial resolution [2]. Popular technique includes the adaptive weighting which is effective in sidelobe suppression but requires intensive computation [20]. In addition, a method based on correlation of adjacent signals like the delay, multiple and sum (DMAS) algorithm, has also been applied [21]. Though effective to a certain extent, however, this algorithm is less efficient due to the large number of multiplication operations involved. Moreover, methods based on multiplication of the beamformer’s output with a weighting factor like the Coherence Factor (CF) [22], the Phase Coherence Factor (PCF), and the Sign Coherence Factor (SCF) [3] have also been reported. Image enhancement resulting from the application of CF is discussed in [23], [24]. Meanwhile, PCF was also applied for reducing the sidelobe’s artifact in the frequency domain [8]. However, the analysis was only performed in 2-D and no resolution analysis was reported.

In this work, we investigate the application of PCF to suppress the sidelobes of a time domain Delay-And-Sum (DAS) beamforming in combination with a sparse planar array. The investigation is extended to TWRI in 3-D that would reveal not only positions but also the target’s height. The ability to sense in the third dimension is vital for enhancing target identification. The system is intended for short range and high accuracy applications. We analyze the performance of the technique in terms of clutter reduction and resolution. Firstly, the through-the-wall signal model and the concept of instantaneous phase are presented. Secondly, the working principle of PCF is addressed. Finally, experimental results are presented and discussed from which the conclusion are drawn.

2. Signal Model

Firstly, we study the travel path of a TWRI signal for one antenna pair in multistatic configuration. For clarity reason, we first consider 2-D plane as shown in geometry in Fig. 1(a). The 3-D representations could easily be derived by including the orthogonal plane of this figure. This figure shows a signal from the m-th transmitter TXm propagating through the wall with thickness Dw and homogenous permittivity of εw. Upon exiting the wall, it is reflected at a point scatterer, F, passing through the wall again before being received by the n-th receiver RXn.

The backscattered signal at point F, Ym,n(F), can be modelled as follows:

\[ Y_{m,n}(t) = A(t - T_{total}^{m,n}) \]  (1)

where A(t) and T_{total}^{m,n} are the signal’s envelope and total time delay respectively. In this case, T_{total}^{m,n} is given by:

\[ T_{total}^{m,n} = \tau_{stand}^{m,n} + \tau_{wall}^{m,n} + \tau_{air}^{m,n} \]

where \( \tau_{stand} \) is the time delay due to the stand-off between the antenna and the wall, \( \tau_{wall} \) and \( \tau_{air} \) are time delays in air and wall respectively. Meanwhile, \( L_{stand} \), \( L_{wall} \) and \( L_{air} \) are distances corresponding to the propagation of the signal in stand-off region, wall and air respectively. In this case, \( L_{stand} \) is assumed known, while \( L_{air} \) is calculated based on the distance to the points in the region of interest. Assuming known wall’s permittivity and thickness, therefore, \( L_{wall} \) can be calculated based on the position of the target with respect to the antenna using Snell’s law [25]. While \( c \) represents the speed of light.

Equation (1) represents an ideal signal for a sidelobe free and single scatterer case. In the presence of sidelobes, however, the reflection from other scatterers particularly those located in proximity to the focal point F will also be sensed by the receivers. This effect is illustrated in Fig. 1(b) which shows a scatterer P located close to point F. In this figure, only three reflected signals due to P are shown as illustration. They are recorded at three receivers, i.e: RXn, RX_{n+1}, and RX_{n+2}. Apparently, the reflections from P follow different paths as compared to F. This leads to a difference in the propagation time delay, \( \Delta \tau_{total}^{m,n} \), for each transceiver pair. The out of focus signal from P can be represented as follows:

\[ Y_{m,n}^{P}(t) = A(t - T_{total}^{m,n} + \Delta \tau_{total}^{m,n}) \]  (3)

\( Y_{m,n}^{P} \) can be re-focused to the intended focal point F by applying the focusing delay yielding:

\[ Y_{m,n}^{F}(t) = A(t + \Delta \tau_{total}^{m,n}) \]  (4)

From (3) and (4), we could deduce that if a point P lies in the main lobe, then \( \Delta \tau_{total}^{m,n} = 0 \). In this case, \( Y_{m,n}^{P} = Y_{m,n}^{F} \).
work the instantaneous phase, domain technique, this information is unavailable. In this
focal point \( F \) has a poor focusing quality due to the side-
lobe interference and hence blurry reconstruction. The
amount in which a signal is out-of-focus can be assessed by
lobe interference and hence blurry reconstruction. The

Mathematically:

\[
\phi_{m,n}(t) = \tan^{-1}\left( \frac{H(Y_{m,n}(t))}{Y_{m,n}'(t)} \right)
\]

where \( H(\cdot) \) denotes the Hilbert transform.

The simplest way to detect whether or not the scat-
ter lies in the main lobe is by examining the variation of \( \phi_{m,n} \) for example by using standard deviation. However, arctangent produces \( \phi_{m,n} \) in wrap around values that vary from \(-\pi\) to \(+\pi\). This causes the variation to be slightly larger than the actual value especially near \( \pm\pi \). In order to overcome the ambiguity of the phase function, it is recom-
ended to deal with the complex function, \( \exp(i\phi_{m,n}(t)) \). By omitting the element and time indices, the standard deviation of the complex function, \( \exp(i\phi) \), can be written as the square root of the sum of the variances of the real and imaginary parts [26]:

\[
\text{std}(\exp(j\phi)) = \sqrt{\text{var}[\cos(\phi)] + \text{var}[\sin(\phi)]}. \tag{6}
\]

Then, PCF is calculated from the standard deviation as follows:

\[
PCF = \left[ 1 - \text{std}(\exp(j\phi)) \right]^2. \tag{7}
\]

From (7) it can be inferred that aperture signals with constant phase produce zero standard deviation and there-
fore maximum PCF value. In contrast, aperture signals with high phase dispersion yield low PCF value and hence suppressed.

3. Proposed Method

In order to suppress the signals at the sidelobes, we
manipulate the phase information across the aperture. Sig-
als reflected by a scatterer outside the focus point produce
a substantial phase shift across the aperture. Hence, multi-
plying PCF with DAS beamformer’s output would theo-
retically reduce cluttering due to sidelobes. Consequently,
a much more accurate image reconstruction is produced.

3.1 Phase Coherence Factor

The instantaneous phase is available from direct in-
phase, \( I \) and quadrature, \( Q \) sampling. However for a time
domain technique, this information is unavailable. In this
work the instantaneous phase, \( \phi_{m,n} \) is estimated by firstly,
shifting the measured backscattered signal, \( Y_{m,n}(t) \), by \( t_{m,n} \) total and secondly, applying the Hilbert transform [3].

Mathematically:

\[
\phi_{m,n}(t) = \tan^{-1}\left( \frac{H(Y_{m,n}(t))}{Y_{m,n}'(t)} \right) \tag{5}
\]

where \( H(\cdot) \) denotes the Hilbert transform.

The desired output is achieved by multiplying the
beamformer’s output as given in (8). Here, \( M \) and \( N \) denote the total number of transmitter and receiver respectively. The corrected output is then squared and integrated over a time window, \( W \), giving the energy value at the focus point, \( En(F) \). The length of \( W \) is set to match the pulse width in order to eliminate late arriving clutter.

\[
En(F) = \int_0^W \left[ \sum_{m=1}^M \sum_{n=1}^N \Delta Y_{m,n}(t) \right] \cdot PCF \, dt. \tag{8}
\]

The CD operation can be described as follows:

\[
\Delta Y_{m,n}(t) = Y^2_{m,n}(t) - Y^1_{m,n}(t) \tag{9}
\]

where \( Y^1_{m,n} \) and \( Y^2_{m,n} \) are the first and second snapshot re-
spectively. These snapshots can be captured consecutively or at different times. Meanwhile, \( \Delta Y_{m,n} \) is the time shifted version of \( Y_{m,n} \). By taking the difference between two snapshots, therefore, the stationary signatures are elimi-
nated while simultaneously highlighting the moving

targets.

A CD image is generated by applying (8) to all voxels in the region of interest. The implementation of the algorithm is summarized graphically in Fig. 2.

The sidelobe suppression capability of PCF can be visualized by the following Point Spread Function (PSF) in Fig. 3. The PSF is calculated from a simulated data using

\[
\text{std}(\exp(j\phi)) = \sqrt{\text{var}[\cos(\phi)] + \text{var}[\sin(\phi)]}. \tag{6}
\]

Then, PCF is calculated from the standard deviation as follows:

\[
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\]

From (7) it can be inferred that aperture signals with constant phase produce zero standard deviation and there-
fore maximum PCF value. In contrast, aperture signals with high phase dispersion yield low PCF value and hence suppressed.

3.2 PCF Weighted DAS

The desired output is achieved by multiplying the
weighting factor, \( PCF \), with the beamformer’s output as given in (8). Here, \( M \) and \( N \) denote the total number of transmitter and receiver respectively. The corrected output is then squared and integrated over a time window, \( W \), giving the energy value at the focus point, \( En(F) \). The length of \( W \) is set to match the pulse width in order to eliminate late arriving clutter.

\[
En(F) = \int_0^W \left[ \sum_{m=1}^M \sum_{n=1}^N \Delta Y_{m,n}(t) \right] \cdot PCF \, dt. \tag{8}
\]

The CD operation can be described as follows:

\[
\Delta Y_{m,n}(t) = Y^2_{m,n}(t) - Y^1_{m,n}(t) \tag{9}
\]

where \( Y^1_{m,n} \) and \( Y^2_{m,n} \) are the first and second snapshot re-
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targets.

A CD image is generated by applying (8) to all voxels in the region of interest. The implementation of the algorithm is summarized graphically in Fig. 2.

The sidelobe suppression capability of PCF can be visualized by the following Point Spread Function (PSF) in Fig. 3. The PSF is calculated from a simulated data using
Finite Difference Time Domain (FDTD) technique with a metallic square shaped target of size 30 cm \( \times \) 30 cm and located at the broadside (scan angle = 0º) of the array. The average sidelobe level for DAS is –18 dB. After multiplication with PCF, the average sidelobe level reduces to –43 dB. Clearly, the angular resolution is not only preserved but also improved as evident in Fig. 3.

4. Experimental Validation and Setup

The data acquisition system is composed of a 16-elements antenna array connected to a 2-port Vector Network Analyzer (VNA) via a multiplexer. A laptop with Labview program was used to synchronize the multiplexing of the transmitting and receiving channels, providing 240 multi-static signals per snapshot. The VNA is set to sweep from 1 GHz to 5 GHz in 5 MHz discrete step. It takes approximately 0.5 s to completely record the signal from a single transceiver pair. This is equivalent to an imaging rate of about 2 min per frame. Hence, the system is suitable for visualizing slow moving targets which are common in many TWRI applications. Operation at higher speeds will require major developments in data acquisition electronics. The time domain signal is obtained through Inverse Fast Fourier Transform function with 800 data points over 70 ns duration each. The electrical delays due to cables and amplifiers have been measured and accounted for in the measurements. Two identical UWB amplifiers with 29 dB gain were used to amplify the transmitted and received signals. The maximum transmitted power of 13.5 dBm (22.4 mW) is obtained after considering the coaxial cable loss. This is almost equivalent to half of the power of a commercial Wifi system.

The configuration of the sparse antenna array is shown in Fig. 4. It has a mean spacing, \( d \) of 30 cm with non-uniform randomization of each element over \( \pm d/4 \) in order to reduce the sidelobes. Further reduction is possible by using larger randomization distance like \( \pm d/2 \). However, this would cause some elements being impractically too close, resulting in the decrease of signal-to-noise ratio due to the increase of capacitive coupling effect [27]. The antennas are the biconical type which have attractive features like small size and wideband characteristic [28]. The antenna measures 10 mm in height and 45° in cone angle, and is fabricated from copper. Each antenna is sequentially operated as transmitter and receiver. The array is mounted directly onto the wall’s surface in order to maximize antenna-wall coupling.

Meanwhile, the wall is made up of typical residential grade bricks with 10 cm thickness, 2 m tall and 2.2 m long. A homogeneous permittivity value of 3.8 was assumed for the wall material. The wall is located at a minimum distance of 4 meters from the surrounding sidewalls and ceiling. A layer of 12 inches pyramidal absorbers was placed on the floor in the vicinity of the wall to mitigate floor reflections.

4.1 Sidelobe Suppression

To evaluate the performance of the algorithm in suppressing the sidelobes, we consider an experiment with two targets moving side by side away from the wall with a displacement of 50 cm as illustrated in Fig. 5.

Two rectangular metallic sheets with width 30 cm and height 170 cm were used as targets. A few objects such as a wood plank, metal bucket, and chair were randomly placed in the scene in order to study the effect of cluttering on the image reconstruction algorithm. These objects remain stationary throughout the measurement. The experimental set-up is shown in Fig. 6.

The targets are assumed to be very slow as compared to the acquisition time so that they appear stationary during the snapshot [6], [29], [30]. This was done by taking the snapshots with the targets stationary at each position. CD image from the snapshots is reconstructed using the proposed technique. For comparison purpose, the same image is also reconstructed using original and tapered DAS beamformer with Chebyshev weighting [2], [31].
4.2 Image Spatial Resolution

In addition to sidelobe suppression’s performance, investigations were also performed to study the effect of the algorithm on image resolution. In any imaging modality, resolution is an ambiguous concept which takes different definitions and metrics. In this work, the resolution is defined as the ability of the system to resolve two targets separated by $s$ and located at a specific distance from the wall as illustrated in Fig. 7. The objective is to find $s_{\text{min}}$, which is the minimum distance before the targets become completely indistinguishable. Only the crossrange resolution is considered in this case. The measurements were performed at three positions - P1, P2 and P3.

The limit of the separation resolution is determined by the resolution ratio, $r$, from the 1-D cross section of the reconstructed image taken on a line passing through the center of the targets:

$$r = \frac{U_{\text{min}}}{U_{\text{max}}}, \quad (10)$$

where $U_{\text{min}}$ is the minimum pixel value between the two targets, and $U_{\text{max}}$ is the maximum pixel value on the 1-D cross-sectional image. Figure 8 illustrates an example of the determination of $r$ from an arbitrary generated image.

In this case, the highest threshold limit for $r$ generally equals to 0.81 [27]. If $r > 0.81$, then the targets are completely indistinguishable. The value of $s$ where $r$ is closest to 0.81 is chosen as the minimum separation distance, $s_{\text{min}}$. This value is compared with the value calculated analytically. Although the separation resolution has never been scientifically proven to be equal to the theoretical resolution, however, this metric provides a good estimation of the system’s resolution and has been used by other researchers [32].

The radiation pattern of a focused beam in the near field is same as the radiation pattern in the far field [33]. Therefore, the 3-dB beamwidth, $\Delta \phi_{3\text{dB}}$, for a UWB non-sinusoidal pulse-based beamformer with an array length, $L$, and in scan direction, $\theta$, from the array normal can be estimated by the following far field formula:

$$\Delta \phi_{3\text{dB}} = \frac{b \cdot T \cdot c}{L \cos \theta} \quad (11)$$

where $T$ and $b$ are the effective pulse duration and a constant respectively [34]. In general, $b$ is fixed to 0.886 in the calculation of the beamwidth of a conventional and untapered DAS beamformer near the broadside where the beamwidth is narrowest [35]. The formula for the theoretical crossrange resolution, $\delta_{\text{cr}}$, is then given by the arc length of the beamwidth over the target’s range, $R_p$.

Mathematically:

$$\delta_{\text{cr}} = R_p \cdot \Delta \phi_{3\text{dB}} \quad (12)$$

The geometry for calculation of $\delta_{\text{cr}}$ is visualized in Fig. 9.

5. Results and Discussion

Fig. 10(a) shows the measured reflection coefficient of the biconical antenna indicating a bandwidth from 2.25 GHz to 5 GHz (maximum sweeping frequency) as defined by the frequency range where $S_{11} < -10$ dB. This figure confirms the wideband characteristic of this antenna which exceeds 3 GHz. This bandwidth is sufficient for TWRI for most applications.

Meanwhile, Fig. 10(b) shows the transmission coefficients $S_{21}$ measured from one pair of the antennas using VNA. The solid curve represents the response without targets. The responses in the presence of targets are superimposed as dashed lines in the figure, however, they can only be clearly seen at points where the reflection signals from the targets arrived at positions marked as P1, P2 and P3 respectively. A close examination of Fig. 10(b) shows
These early arriving clutters are common for TWRI and harmless if mapping of the region very close to the wall could be avoided. However, the clutter increases again almost at the same time as signal P3, this time due to the sidewalls of the laboratory. As can be seen, the magnitude of the targets reflection signals decreases from 9 dB at P1, to 6 dB at P2 and only 2 dB at P3. Evidently, the magnitude of the signal decreases as the distance from the wall increases. Additionally, the clutter interference causes a smaller increase in the amplitude in the presence of the targets, and consequently a further degradation of the signal as evident from Fig. 10(b) at P3.

Figure 10(c) shows an example of the instantaneous phase calculated using (5) for \( m = 9 \) and \( n = 11 \). Meanwhile, Figure 10(d) shows the alternative representation of Fig. 10(c) as unit circles. It can be seen from this figure there is a smooth transition of \( \phi_{m,n} \) when it is represented as \( \exp(j\phi_{m,n}) \).

5.1 Sidelobe Suppression

The CD images of two targets with separation distance, \( s = 40 \) cm located at P1 moving away from the wall are shown in Fig. 11. The original DAS image in Fig. 11(a) shows strong ring artifacts, shadowing the two targets making them unresolvable. This artifact appears as though the target images being replicated at multiple points along the arc due to the sidelobes. The intensity of replication, however, starts to decrease as the beamformer is steered away from the targets. Clearly in this case, the CD failed to minimize the sidelobe artifacts. Nevertheless, it is efficient in removing the influence from the foreign objects placed in the scene due to their invariant positions during the two snapshot’s interval.

Figures 11b(i-iii) show CD images with tapered beamformer using Chebyshev weighting technique [2], [31]. For this purpose, we reconstruct the image with different values of attenuation factors ranging from 5 dB to 50 dB. The best clutter suppression is achieved by using weighting with an attenuation factor of 25 dB as displayed in Fig. 11(b)(i). Nevertheless, it is not possible to correctly distinguish the two targets. A higher or lower attenuation factor leads to an increase in the ring artifact again as can be seen in Fig. 11(b)(ii) and Fig. 11(b)(iii). For a tapered beamformer, the main lobe’s width is strongly coupled with the value of the attenuation factor. As expected, the randomization of the elements in the array and tapering with Chebyshev are insufficient to reduce the sidelobe. Hence, the image distinguishability remains relatively low as these results indicated.

In this case, the PCF weighting produces much accurate reconstruction compared to conventional DAS and Chebyshev weighted as clearly visualized in Fig. 11(c). The two targets at two different positions, appear more distinct compared to Fig. 11(a) and Fig. 11(b). It can also be seen that the image intensity at P2 is substantially much weaker than P1 due to the increase in the downrange distance.
To quantitatively assess the suppression of the side-lobes, the target-to-clutter ratio (TCR) is used. TCR is the ratio between the maximum pixel value of the target to the average pixel value in the clutter region. Mathematically:

\[
TCR = 10 \log_{10} \left( \frac{\max_p |E_{\text{Target}}(p)|}{\frac{1}{N} \sum_{n=1}^{N} |E_{\text{Clutter}}(p)|} \right)
\]

where \(E_{\text{Target}}(p)\) is the pixel value in the target area, \(E_{\text{Clutter}}(p)\) the pixel value in the cluttered area and \(N\) is the number of pixels in the cluttered area. The cluttered area is the region where no target is present. This area was chosen just next to the target area where the ring artifact is the strongest. Examples of the target and cluttered areas are shown in Fig. 11(a). In this case, the areas enclosed by solid and dotted boxes represent the target and cluttered areas respectively.

Applying (13) to the images in Fig. 11 gives TCR values which are summarized in Tab. 1. It can be observed from this table that the TCR calculated from the proposed DAS-PCF is much higher compared to other image reconstruction techniques. In this case, DAS-PCF resulted in TCR of approximately 11 dB compared to 5.5 dB of the original DAS (Ori. DAS), 6.6 dB, 5.3 dB and 3.6 dB of Chebyshev weighted (Cheb. Wt.) corresponding to attenuation factors of 25 dB, 40 dB and 10 dB respectively. The use of PCF has resulted in the increase in the TCR value by more than two folds compared to other methods. This explains the significant improvement in the overall quality of the reconstructed image that this method produced.

<table>
<thead>
<tr>
<th>Algorithm</th>
<th>Ori. DAS</th>
<th>Cheb. Wt. 25 dB</th>
<th>Cheb. Wt. 40 dB</th>
<th>Cheb. Wt. 10 dB</th>
<th>DAS-PCF</th>
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</thead>
<tbody>
<tr>
<td>TCR (dB)</td>
<td>5.5</td>
<td>6.6</td>
<td>5.3</td>
<td>3.6</td>
<td>10.9</td>
</tr>
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</table>

Tab. 1. TCR values calculated from images in Fig. 11, comparing PCF and standard approaches.
In general, the effectiveness of PCF in suppressing the sidelobes has been demonstrated by the improvement in visual quality and quantitative TCR. The effect of the algorithm on the spatial resolution is discussed in the following paragraph.

5.2 Spatial Resolution

The resolutions were evaluated at positions P1, P2 and P3. In this case, two metallic sheets of the same size were used as targets. For this purpose, we use background subtracted images, where the snapshots of the scene without target are subtracted from the scene with targets. Background subtraction is possible under surveillance mode where the information of the empty scene is available. Furthermore, the background subtraction is based on the same principle as CD, in which the snapshots captured at different time are subtracted coherently.

Examples of the reconstructed images are shown in Fig. 12. The images are reconstructed at P2 with decreasing separation distances. As obvious from this figure, the distinguishability increases as the separation distance increases. It can also be seen from the result in Fig. 12(c) that the two targets appear as one blob when the separation distance is reduced to 20 cm. This is the limit of the resolving power of the current set-up.

The resulting resolution ratios, $r$, for all imaging positions and separation distances are displayed in Fig. 13. As expected, the shorter the range, the lower the value of resolution ratio, $r$, and hence the better separation resolution. From the graph in Fig. 13, it can be seen that at P1, the minimum separation distance, $s_{\text{min}}$, where $r$ approaches 0.81 is approximately 20 cm. Similarly, the $s_{\text{min}}$ values at P2 and P3 are 30 cm and 50 cm, respectively.

![Fig. 12. PCF weighted images of two metallic targets located at P2 with separation distances of (a) $s = 60$ cm, (b) $s = 40$ cm, and (c) $s = 20$ cm.](image)

![Fig. 13. Resolution ratio $r$ for PCF weighted images.](image)

<table>
<thead>
<tr>
<th>Parameter</th>
<th>P1</th>
<th>P2</th>
<th>P3</th>
</tr>
</thead>
<tbody>
<tr>
<td>Range, $R_p$</td>
<td>250 cm</td>
<td>350 cm</td>
<td>450 cm</td>
</tr>
<tr>
<td>Minimum separation, $s_{\text{min}}$</td>
<td>20 cm</td>
<td>30 cm</td>
<td>50 cm</td>
</tr>
<tr>
<td>Theoretical resolution, $\delta_{cr}$</td>
<td>19.6 cm</td>
<td>27.5 cm</td>
<td>35.3 cm</td>
</tr>
<tr>
<td>$s_{\text{min}} - \delta_{cr}$</td>
<td>0.4 cm</td>
<td>2.5 cm</td>
<td>14.7 cm</td>
</tr>
<tr>
<td>$(s_{\text{min}} - \delta_{cr})/\delta_{cr}$</td>
<td>2%</td>
<td>9%</td>
<td>42%</td>
</tr>
</tbody>
</table>

Tab. 2. Comparison between measured and computed separation resolutions.

In Tab. 2, the calculated theoretical resolutions and the measured minimum separation distances at P1-P3 are compared. The theoretical crossrange resolution, $\delta_{cr}$, is calculated using (12) with $T = 0.59$ ns, which is the measured duration of the synthesized pulse at 50% amplitude. The value of the aperture length, $L$, is taken as 2 m. A thickness of 10 cm and a permittivity value of 3.8 are assumed for the brick wall.

Referring to Tab. 2, the $s_{\text{min}}$ at P1 is comparable with the theoretical resolution with an error of 2%. However, the error increases as the distances increases. In this case, the error calculated at P2 is 9% compared to 42% at P3. Even though the measured separation resolutions at P2 and P3 indicate increasing deviations from the theoretical resolutions, however, the latter do not consider other external factors such as the effect of SNR values.

In the real application, the SNR values could deteriorate, for example due to an increase in the target’s distance or due to interference with the surrounding clutters as previously seen in Fig. 10(b). This is consistent with an earlier finding which concluded that the decrease in SNR leads to a degradation in the separation resolution for both pulsed based and stepped frequency radar imaging [28]. In addition to clutters, other inherent artifacts like wall’s dispersion, creeping wave [31] and jitters in device [36] would degrade the SNR and consequently the resolution too. Considering all these factors and based on the achieved resolution at P1, we can conclude that PCF technique performs well in both sidelobe suppression and preserving the spatial resolution.
5.3 3-D TWRI Capability

Finally, to demonstrate the capability of the proposed technique in 3-D, the images corresponding to two metal sheets of different heights located at 2.5 m from the wall are reconstructed. The heights of the first and second metal sheets are 150 cm and 170 cm respectively. This experiment is repeated using a real human target with height 165 cm standing at 3 m from the wall. The results are shown in Fig. 14. In this case, the images are reconstructed in x-y, y-z and iso-surface planes. The iso-surface is generated from the voxels with energy value 70% from the maximum energy. All are reconstructed using PCF weighted DAS. Again, the targets have been accurately localized as evident from Fig. 14(a)(i) and Fig. 14(b)(i). However, the heights are less accurately reconstructed, with absolute errors ranging from 20 to 30 cm as can be seen in Fig. 14(a)(ii). This is mainly due to the shorter aperture length in the vertical direction, and hence, limited resolution along the z-axis. Nevertheless, a close examination of Fig. 14(a)(ii) reveals that it is possible to distinguish the two targets of different heights. A slight improvement in distinguishing the object based on height is visualized through the iso-surface image in Fig. 14(a)(iii). Meanwhile, the height profile of the human target produces an absolute error of 40 cm as can be seen in Fig. 14(b)(ii) and Fig. 14(b)(iii). The image appears noisy and cluttered. This is due to the fact that average adult human poses a smaller radar cross section as compared to the metallic sheet. Also, both targets appear as though that they are in the air 50 cm above the floor. This artifact is due the characteristic of the biconical antennas in which the power in elevation plane decreases as the elevation angle increases. Therefore, the antenna is less sensitive to the reflection coming from the direction of the floor resulting in a reduced energy intensity at the lower edges.

6. Conclusions

From the experimental results, it is evident that the proposed technique is effective in suppressing the clutter due to the sidelobes arising from a sparse antenna array. The TCR value for the CD image has been significantly improved using PCF corrected beamformer outputs, thereby improving the localization of two closely located targets. Compared to conventional beamtapering techniques, PCF produces a better sidelobe suppression while preserving the spatial resolution. This is proven by the achieved separation resolution at a range of 2.5 m, which is comparable with the theoretical value. With the proposed technique, it is possible to perform 3D TWRI utilizing a sparse planar antenna array with reduced effect of the sidelobes.

Acknowledgments

This work is funded by USM research grant 6071222. The first author acknowledges the Education Ministry of Malaysia and Universiti Teknikal Malaysia Melaka for scholarship.

Fig. 14. PCF reconstructions of (a) metallic, and (b) human targets (i) reconstructed images x-y plane, (ii) reconstructed images y-z plane, (iii) reconstructed images in iso-surface, and (iv) real images.
References


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