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Numerical study of defect localization in additive manufactured short fiber reinforced composites with diffuse ultrasonic wave inspection

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The structures of short fiber reinforced composites become designable from macroscopic to microscopic due to the advancement in additive manufacturing technologies. The diffuse ultrasonic wave inspection benefits from information from multiple scattering processes, which is suitable for the quality assurance of complex structures. This study established a two-dimensional wave propagation model assuming the decoupling of the fiber volume into the fiber distribution matrix in the plane and the local fiber fraction along the thickness axis. The k-space pseudospectral method was applied to calculate the diffuse wave fields. The defect inspection process was studied numerically based on the Locadiff technique for additive-manufactured short-fiber reinforced composites. The stretching method provided the same average distance but a smaller relative deviation to the defect than the doublet method. The localization resolution improved significantly for the initial increment of the number of transmitters; limited improvement can be achieved further. Localization results fluctuated when the transmitter combination groups were distant from the defect. This method worked well with the isotropic and quasi-isotropic plates, while an oversimplification was found for the unidirectional fiber structure.

KEYWORDS

diffuse ultrasonic wave, defect localization, short fiber reinforced composite, additive manufacturing, structural health monitoring

1 Introduction

Composite structures are widely used in the aviation, machinery, and other domains owing to their high specific strength, specific modulus, and other advantages. Efficient and rapid operation and maintenance (O&M) of composite based mechanical structures is crucial for ensuring high levels of performance and reliability of mechanical equipment. Current health monitoring methods for composite structures include vibration mode monitoring [1], structural strain monitoring [2,3,4], acoustic emission technology [5]; [6], and intelligent coating technology [7]. Among them, the ultrasonic structural health monitoring method, which often distributes piezoelectric sensors [8] inside a structure, is economical and easy to implement. However, the existing centralized and distributed general mechanical equipment structure health monitoring systems are wired-state monitoring systems. An online monitoring system based on a wireless sensor network (WSN) [9] can reduce wiring and lower deployment cost; furthermore, it offers increased flexibility, maintainability, and scalability. Therefore, the application of WSN technology to mechanical equipment online monitoring systems is highly desired. WSNs have been used in a few applications for structural health monitoring. However, many challenges, such as short lifetimes, unreliable communication, and poor real-time performance, should be resolved. Therefore, research on efficient and accurate real-time monitoring methods is crucial for structural health monitoring.

Various ultrasonic techniques can improve imaging quality and efficiency. Thick welded joints can be inspected using laser ultrasound B-scans based on the synthetic aperture focusing technique (SAFT) [10]. Phased array [11] inspection using multiple transmitter-receiver pairs can facilitate full matrix capture (FMC) and full-field imaging with the total focusing method (TFM) to increase inspection efficiency. Wavenumber algorithm [12] is considered to further accelerate the imaging speed in real-time based on the wave equation for the defect inside composites.

Additionally, classical beamforming methods, such as multiple signal classification (MUSIC), can be used to decompose the covariance matrix of an array output data into eigenvalue to obtain the signal subspace corresponding to the signal component and the noise subspace orthogonal to the signal component, which can be used to effectively denoise and extract the characteristic signal [13]. The main features of this signal can be extracted to locate the signal source. Yuan et al. proposed that the near-field two-dimensional (2D)-MUSIC method [14] can simultaneously locate the damage angle and distance in composite structures. Furthermore, Yang et al. proposed the Am MUSIC damage-imaging method [15]. The complete spatial spectrum of a test sample can be generated by quantizing the orthogonal attribute between the inherent signal and noise subspaces of a matrix, and the process is not limited by the number of damages. Bao et al. proposed an anisotropic compensation MUSIC algorithm [16] that can jointly compensate for different types of sensor phase errors to reduce the positioning error. This algorithm was verified through a reinforced composite plate structure to improve the accuracy and reliability of damage positioning. Xu et al. proposed a focused MUSIC algorithm [17] for baseline-free Lamb-wave-based damage location in isotropic materials. The virtual time-reversal technique was used to compensate for the dispersion effect, and the focused signal was truncated to avoid estimating the number of scattering sources and baseline subtraction. Fan et al. used phase-coherent MUSIC [18] for EDM line inspection, which could evaluate its length even when it was oblique to the linear array.

These methods facilitate the localization and imaging of regional damages. However, scattering occurs when applied to complex configurations, such as walls, ribs, corners, and beams in a structure and it interferes with acoustic wave transmission. In this study, we performed inversion imaging of internal defects in various composite structures using multi-channel scattering information to achieve wireless autonomous recognition in combination with WSNs.

2 Theory

2.1 Forward model

A diffuse wave field can be established after multiple scattering in a complex medium like additive manufacted composites. The defect information can be amplified in the help of this field. Suppose the waveforms $\varphi_A(S, R, t)$ and $\varphi_B(S, R, t)$ received at position R can be obtained before and after a defect appearance under an impulse excitation source at position S. The correlation between these two can then be calculated through Eq. 1 ([19,20]):

$$\langle \varphi_A(S,R,t)\varphi_B(S,R,t)\rangle = I(S,R,t) - \frac{c\sigma}{2} \int_0^t I(S,x,u)I(x,R,t-u)du$$
(1)

where *c* denotes the wave velocity, *x* corresponds to the defect location, σ represents the scattering cross-section, and $I(r_1, r_2, t)$ is the intensity propagator from position r_1 to r_2 along time *t*. Eq. 1 is an expression of Green's function recovery method without defect presentation[10] when $\sigma = 0$. It can be normalized into Eq. 2:

$$\frac{\langle \varphi_A(S,R,t)\varphi_B(S,R,t)\rangle}{\sqrt{\langle \varphi_A^2(S,R,t)\rangle\langle \varphi_B^2(S,R,t)\rangle}} = CC(S,R,x,t)$$
$$= 1 - DC(S,R,x,t)$$
(2)

TABLE 1 Material parameters.

Items	Values	Items	Values
Young's modulus of short fiber, E_f	200 GPa	Local fiber fraction, P_z	40%
Young's modulus of the matrix, E_m	3 GPa	Effective Young's modulus, $E_{eff} = E_f \cdot P_z + E_m (1 - P_z)$	81.2 GPa
Density of the matrix, ρ_m	1,150 kg/m ³	Effective density, $\rho_{eff} = \rho_f \cdot P_z + \rho_m (1 - P_z)$	1,410 kg/m ³
Density of short fiber, ρ_f	1800 kg/m ³	Frequency, <i>f</i>	2 MHz
Acoustic velocity of short fiber, V_f	8,804.7 m/s	Absorption [32], α	10 <i>db/cm</i>
Acoustic velocity of the matrix, V_m	1874 m/s	Ratio of non-zero elements in P _{surface}	50%





where *CC* represents the correlation coefficient and *DC* denotes the decorrelation coefficient. DC = 0 when the two waveforms are identical and DC = 1 when the two waveforms are absolutely different. Moreover, the decorrelation coefficient is related to the sensitivity kernel function through Eqs 3, 4 as follows:

$$DC(S, R, r, t) = \frac{c\sigma}{2}K(S, R, r, t)$$
(3)

and

$$K(S, R, r, t) = \frac{\int_0^t I(S, r, u) I(r, R, t - u) du}{I(S, R, t)}$$
(4)

If the composite sample has an isotropic scattering property in an infinite d-dimensional, the average scattering intensity distribution approximates to the diffusion equation solution:

$$I(r_1, r_2, t) = \frac{1}{(4\pi Dt)^{\frac{d}{2}}} exp\left(-kt - \frac{(r_2 - r_1)^2}{4Dt}\right)$$
(5)

where D denotes the diffusivity and k is the dissipation parameter. A more accurate solution can be obtained from the radiative transfer equation. The first born approximation can be used[21] to study the kernel in weak scattering media if the scattering is not sufficiently strong for the diffusion approximation. More generally, the kernel does not have an analytical form and should be calculated numerically for heterogeneous scattering.

2.2 Inversion model

After obtaining DC(S, R, x, t) and K(S, R, r, t), the defect location can be predicted from different inversion algorithms, e.g., linear least square inversion method[22] and Monte Carlo Markov chain method[23]. A classical grid search method[17] was used by searching for the most likely defect position through the cost function:



$$e(x) = \sum \frac{DC^{2}(t)}{\varepsilon^{2}} - \frac{\left(\sum DC \bullet K\right)^{2}}{\varepsilon^{2} \sum K^{2}}$$
(6)

The probability density of the defect appearance at x is defined in Eq. 7:

$$p(x) = \frac{1}{C} exp\left(-\frac{e(x)}{2\varepsilon^2}\right)$$
(7)

where ε corresponds to a fluctuation parameter for decorrelations and *C* denotes a normalization constant.

3 Numerical simulations

The internal structure of composites can be classified into microscopic, mesoscopic, and macroscopic scale [24]. The homogenization method is often applied to determine the effective properties of heterogeneous media [25] and to simplify the wave propagation in these structures [26] by keeping the same strain energy between the heterogeneous and homogeneous materials under arbitrary loads. It is an efficient way to balance the computational cost and the calculation accuracy, especially for lamb wave inspection concerning phase velocity or group velocity information in hundreds of kHz [27]. However, diffuse ultrasonic wave inspection simulation demands mesoscopic or even microscopic details to extract diffuse ultrasonic wave fields from the wave-structure interaction. The k-space pseudospectral method [28] was applied for acoustic wave propagation simulation in Matlab software. The material properties of polyamide 12 and short carbon fiber are listed in Table 1. A perfect matched layer was added to eliminate the edge effect.

In order to simplify the numerical model, the fiber volume fraction $\mathbf{P}_{f_{_total}}$ was decoupled into the fiber distribution matrix $\mathbf{P}_{surface}$ in the x-y plane and the local fiber fraction P_z along Z according to Eq. 8 as shown in Figure 1. Eq. 7 defines the probability density of the defect appearance at x as:

$$\mathbf{P}_{f_{total}} = P_z \bullet \mathbf{P}_{surface} \tag{8}$$

Since a higher fiber volume fraction will lead to an increasing degree of misalignment from additive manufacturing [29], the local fiber fraction is set to be isotropic. Meanwhile, the high specific strength plate is always thin ($d \ll L$) that a constant local fiber fraction of $P_z = 40\%$ was applied with 150×150 mesh size in the x-y plane. The ratio of non-zero elements in $P_{surface}$ is assumed to be 50%. The whole defect localization scheme can be shown in Figure 2 below.

In order to balance between the high attenuation in polymers by keeping relative high sensitivity, the inspection frequency was chosen to be 2 *MHz* here. A Gussian-shaped toneburst is chosen as the excitation source during simulation. Both its time and frequency domain wavefroms are shown in Figure 3. The plate dimension was 0.45×0.45 m² with the center of a 0.015×0.015 m² rectangular defect positioned at [0.3 m, 0.15 m]. The material properties of the defect such as



missing extrudates[30], were assumed to be 80% of those in the initial state. The pitch-catch configuration and the typical waveforms for a random short fiber reinforced composite plate from the pair 3–11 are presented in Figure 4. The diffusivity and the dissipation parameter can be decided from the waveform envelope through the Hilbert transform combing Equation 5, i.e., D = 34 m/s and $k = 4.5 \times 10^3/\text{s}$. The transport mean free path $L^* = 2D/c \approx 2D/((V_f + V_m)/2) =$ 0.012 m was significantly less than the smallest transmitterreceiver pair distance between the pair 1-1 $L_{1-1} = 0.3$ m, ensuring multiple-scattered acoustic wave.





4 Results

4.1 Influence of the kernel determination method

According to Eq. 2, the correlation or decorrelation coefficient can be determined from the waveforms with and without the defect. The doublet method (or cross-spectral moving-window method) and the stretching methods are often used for coefficient calculation. Both methods are influenced by the correlation window width (W). By applying the doublet method, the correlation coefficient of the



pair 3–11 (Figure 5) at t = 0.6 m varied between 0.56 and 0.69 when W changes ($W = 30 \mu s$, 60 μs and 90 μs).

The localization precision would be further influenced by such variation if only the transmitter positions 1–3 were adopted with receiver positions 1–12 fixed. The average distance (AD) to the defect and relative deviation (RD)were 0.029 m and 8.4, respectively. Such ambiguity can be prevented by adding more transmitters (Figure 6), i.e., AD = 0.016 and RD = 3.5% with all the transmitter positions. On the contrary, the stretching method gives a stable and smooth correlation coefficient between 0.64 and 0.68 at t = 0.6 m. As shown in Figure 6, the corresponding AD values were nearly identical to those obtained from the doublet method. However, RD from the stretching method decreased to 2.8% with transmitter positions 1–3 and approached zero with more transmitter positions. The stretching method with the superior coefficient calculation stability was further applied in this study.

4.2 Influence of the transmitters

Diffuse ultrasonic waves can magnify the defects that are influenced by multiple scattering. The defect can be influenced by the diffuse field after a long time at a random position, enabling defect localization. The transmitter-receiver number and distribution optimization always remain problematic for reasonable inspection efficiency. Only the influence of transmitters was studied, keeping the receivers fixed due to the repeatability and reciprocity of the diffuse ultrasonic wave inspection [31]. The different combinations can be adopted for a defined number of transmitters. Increasing the number of transmitters provided a better defect localization resolution with a



decreasing standard deviation, as shown in Figure 7. The average distance converged to 0.016 m, and six random transmitters were sufficient for this plate. Limited localization precision can be improved with more transmitters added.

The localization deviation not only depends on the number of transmitters, but also on the transmitter distribution. When 3 transmitters are chosen among the total 12 ones, 220 combinations can be expected with different root-meansquare-distances (*RMSD*) to the defect. As shown in Figure 8A, the distance prediction error increases when the *RMSD* of the transmitter group increases. A consistent value can be found for the transmitter group near the defect (*RMSD* < 0.2 m). There is



an increasing error when the transmitter group becomes far from the defect (RMSD > 0.2 m). When 8 transmitters are adopted with 495 combinations, a much more consistent value can be observed in Figure 8B. However, the distance value still becomes scatter when the transmitters are far from the defect location. For a real panel, the defect locations are always unknown. Such fluctuation information can also help to verify the proper number and distribution through combination changes in experiments.

4.3 Influence of the fiber structure

Additive manufacturing enables structure design with short fibers in composites. The mesoscopic structure can vary

significantly regarding to the physical property requirement during the design stage under the same ratio of non-zero elements in the P surface. Different multiple scattering behavior during diffuse ultrasonic wave inspection should be considered.

Three typical additive manufactured composite plates were investigated; they are shown in Figures 9A–C with a random pattern, plain weave pattern, and unidirectional pattern, respectively. They have isotropic, quasi-isotropic, and anisotropic macro mechanical properties, respectively. Their corresponding scattering behaviors can be found in Figures 9D–F. The defect localization results can be found in Figures 9G–I under the full transmitter-receiver implementation with the material properties listed in Table 1. The result shows that the localized defect coordinate is at [0.294 m, 0.134 m] for the random pattern with a small error of 3.8% ($\frac{|lr_measure-recel|}{2}$). An

acceptable result can be found for the plain weave pattern with the error of 8.4% due to its quasi-isotropic nature. For unidirectional pattern, a large error of 29.7% was observed. The isotropic sensitivity kernel calculated from Eq. 5 is no more compatible. Numerical intensity solution concerning anisotropic behavior should be used for the kernel computation.

5 Conclusion

Additive manufactured fiber-reinforced polymer composites can possess complex structures from macroscopic to microscopic scales, making their quality assurance difficult. The diffuse ultrasonic wave inspection relies on the multiple scattering domain that contains rich structure information. The Locadiff technique is processed based on a simplified 2D mechanical model by homogenization along the thickness direction. The stretching method was more stable than the doublet method for correlation/decorrelation coefficient calculation and enabled reliable defect localization. An increasing number of transmitters improved the localization precision. However, the localization results fluctuated from the transmitter combination groups that were distant from the defect. The numerical investigation indicated that this method works well with isotropic and quasi-isotropic composite plates with the errors of 3.8% and 8.4% respectively. Further exploration for the defect localization in additive manufactured parts concerning anisotropic scattering and signal processing is expected.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

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Author contributions

YP, JZ, and QZ contributed to conception and design of the study. YP organized the database and performed the formal analysis. QZ wrote the first draft of the manuscript. HX, JS, JZ, and YP wrote sections of the manuscript. All authors contributed to manuscript revision, read, and approved the submitted version.

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Conflict of interest

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