Research Article

Dual Mode Sensing with Low-Profile Piezoelectric Thin Wafer Sensors for Steel Bridge Crack Detection and Diagnosis

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Received 30 May 2011; Accepted 6 September 2011

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Monitoring of fatigue cracking in steel bridges is of high interest to many bridge owners and agencies. Due to the variety of deterioration sources and locations of bridge defects, there is currently no single method that can detect and address the potential sources globally. In this paper, we presented a dual mode sensing methodology integrating acoustic emission and ultrasonic wave inspection based on the use of low-profile piezoelectric wafer active sensors (PWAS). After introducing the research background and piezoelectric sensing principles, PWAS crack detection in passive acoustic emission mode is first presented. Their acoustic emission detection capability has been validated through both static and compact tension fatigue tests. With the use of coaxial cable wiring, PWAS AE signal quality has been improved. The active ultrasonic inspection is conducted by the damage index and wave imaging approach. The results in the paper show that such an integration of passive acoustic emission detection with active ultrasonic sensing is a technological leap forward from the current practice of periodic and subjective visual inspection and bridge management based primarily on history of past performance.

1. Introduction

According to the Federal Highway Administration (FHWA) National Bridge Inventory (NBI) of 2007, the number of structurally deficient and functionally obsolete bridges is 72,524 and 79,792, respectively [1]. While there are about 10,000 bridges being constructed, replaced, or rehabilitated annually in the United States at a cost of over $5 billion, the total annual costs including maintenance and routine operation are significantly higher [1]. As the inventory continues to age, routine inspection practices will not be sufficient for the timely identification of areas of concern and to provide enough information to bridge owners to make informed decisions for safety and maintenance prioritization. Continuous monitoring is therefore desirable for long-term evaluation; monitoring areas of concern, such as retrofits or previous repairs or monitoring an area with known flaws, before a scheduled inspection. Continuous monitoring can also be used in cases where there is a concern about vandalism, terrorism, and/or bridge element integrity. Therefore, monitoring of fatigue cracking in steel bridges is of interest to many bridge owners and agencies.

To address this urgent need, the authors are conducting research on novel and promising sensing approaches together with energy harvesting devices to reduce the dramatic uncertainty inherent into steel bridge inspection and maintenance plan [2]. One of the challenges in this research is focused on the development of dual use piezoelectric wafer active sensors for fatigue crack detection. The combined schematic uses acoustic emission to detect the presence of fatigue cracks in their early stage while active sensing allows for the imaging and quantification of cracks using minimum number of sensors.

1.1. Crack Detection on Steel Bridge Structures. The monitoring of fatigue cracking in bridges has been approached with acoustic emission using either resonant or broadband
sensors. Acoustic emission monitoring is capable of detecting crack growth behavior [3–7] and assessing the integrity of structures such as bridges and aircraft [8, 9]. Acoustic emission data has recently been directly related to crack growth rate in representative steel compact tension (CT) specimens. This development holds promise for the prognosis of in-service bridges [10]. The method has the notable advantage that the precise location of cracking does not need to be known for evaluation purposes. Rather, the sensors together with appropriate algorithms are capable of locating and quantifying active crack activity. It has been reported that acoustic emission techniques are so sensitive that fatigue cracks can be detected successfully even though the crack length may be less than 10 μm [5, 7]. However, one of the challenges in passive monitoring is that the acoustic emission relies on an active crack growing process. Also, though acoustic emission sensing can detect a crack at its very early stage, it generally cannot provide information about the crack size or crack growth rate unless an initial size is provided.

Historically, AE signals have been captured with specially designed and fabricated AE sensors. Conventional AE sensors are made of piezoelectric crystals as the sensing elements which are encapsulated for protection and coupled together with a wear plate for good acoustic coupling. The frequency content and sensitivity of the sensor are controlled by the geometry and properties of the piezoelectric crystal as well as the housing for the crystal.

1.2. Piezoelectric Wafer Active Sensors. Piezoelectric wafer active sensors (PWAS) can function as an active sensing or passive device or network using piezoelectric principles and provide a correlation between mechanical and electrical variables. They can be permanently attached to the structure to monitor condition at will and can operate in active guided wave interrogation or passive AE sensing modes. The transmission of actuation and sensing between the PWAS and the host structure is achieved through the bonding adhesive layer. The adhesive layer (Figure 1) acts as a shear layer, in which the mechanical effects are transmitted through shear effects [11].

An important characteristic of PWAS, which distinguishes them from conventional ultrasonic transducers, is their capability of exciting multiple guided Lamb wave modes at a single frequency. There are at least two Lamb modes, A0 and S0, existing simultaneously when the product of the wave frequency and structure thickness \( f \cdot d \) product falls in the range of 0–1 MHz-mm. At larger \( f \cdot d \) product values, more modes are present. In addition, due to the intrinsic dispersion property, the Lamb wave modes propagate at different speeds and the speeds change with frequency, which complicates the signal interpretation for damage detection. A single mode that is sensitive to the damage is desired for most of the SHM applications. This may be attained through wave tuning [12]. The process of wave tuning attempts to modify the excitation parameters to excite a certain mode for detection of a specified type or instance of damage [12]. By carefully selecting PWAS length at either an odd or even multiple of the half wavelength, a complex pattern of strain maxima and minima emerges (Figure 2). Since several Lamb modes, each with its own different wavelength, coexist at the same time, a selected Lamb mode can be tuned by choosing the appropriate frequency and PWAS dimensions.

An example of PWAS tuning is presented in Figure 2 for a 7-mm square PWAS installed on a 1-mm aluminum alloy 2024-T3 plate. The experimental amplitude plot in Figure 2(a) shows that for the plate being studied, a S0 tuning frequency around 200 kHz can be identified, where the amplitude of the A0 mode is minimized while that of the S0 mode is still strong. Therefore, by choosing the excitation frequency, a single mode can be obtained for damage detection [13]. Theoretical prediction given in Figure 2(b) is consistent with the experimental results.

1.3. PWAS Dual Mode Sensing toward Field Application. Dupont et al. [14] have demonstrated the possibility of using embedded piezoelectric thin wafers to detect AE signals in composite materials. In the subject project, we are developing a dual mode sensing approach using the low-profile wireless PWAS network with energy harvested from wind and/or ambient vibration energy [15]. To minimize the energy consumption, on one hand, it is envisioned that as few as four PWAS will be employed to monitor the crack growth. On the other hand, the dual mode sensing allows PWAS to operate in passive AE mode throughout the entire monitoring process unless significant AE events are detected, indicating major crack presence in the structure. When an AE event is identified, the PWAS can be switched to active mode to interrogate the bridge structure with propagating guided waves to assess the crack size and location. Eventually, an entire steel bridge may be mapped with the guided wave interrogation with visualized crack damage.

2. PWAS Passive Mode Acoustic Emission Detection

In the subject project, we adapted PWAS as AE sensors to detect stress waves with frequency components concentrated at 150 kHz where the acoustic signals propagating with minimal attenuation and background noise due to the rubbing of structural components.

2.1. PWAS as AE Sensor. Laboratory tests have been conducted to investigate the PWAS application as an AE sensor.
A typical commercial R151 (http://www.pacndt.com/downloads/Sensors/Integral%20Preamp/R151-AST.pdf) AE sensor was used to calibrate the measurements. Two specimens were used in the tests. One was a 1.6 mm thick 2024 aluminum plate and the other was a 19 mm thick A572 grade 50 structural steel panel. Both specimens were installed with 7-mm diameter 0.2 mm thick round PWAS Material APC 850 by Americanpiezo (http://www.americanpiezo.com/apc-materials/choosing.html) using M-200 bonding adhesive following the standard installation used by strain gauge. The R151 sensor was mounted using hot glue.

The test setup is illustrated in Figure 3. PAC DiSP (http://www.pacndt.com/index.aspx?go=products&focus=/multi-channel/disp.htm) system was used to perform data acquisition. PWAS was connected to a preamp then to channel 1. The preamp had a 100–1200 kHz built-in filter and could accommodate a signal amplification of 40 dB. The R151 sensor had a built-in internal preamp with a gain also at 40 dB and was connected to channel 2. AE events were introduced by pencil lead break (PLB). Rubber gloves were used to avoid causing electrical disturbance by touching the plate.

### 2.1.1. Aluminum Plate Tests.

In the first part of the work, a 1.6-mm thick aluminum plate, approximately 300-mm by 300-mm, was used for testing PWAS AE detection. PWAS and R151 were placed adjacent on the plate, about 165 mm away from the plate edge where the PLB was applied. In total, five PLB of various lead sizes were applied. The PWAS transducer detected all of them with comparable amplitudes to those captured by R151, as summarized in Table 1.

![Figure 2](image2.png)

**Figure 2:** Lamb wave mode tuning on a 1-mm thick aluminum alloy 2024-T3 using 7-mm PWAS. (a) Experimental wave amplitude within 0–700 kHz; (b) predicted strain curves [13].

<table>
<thead>
<tr>
<th>PLB size</th>
<th>PWAS AE amplitude (dB)</th>
<th>R151 AE amplitude (dB)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.7 mm</td>
<td>83</td>
<td>82</td>
</tr>
<tr>
<td>0.5 mm</td>
<td>89</td>
<td>89</td>
</tr>
<tr>
<td>0.5 mm</td>
<td>78</td>
<td>78</td>
</tr>
<tr>
<td>0.3 mm</td>
<td>88</td>
<td>86</td>
</tr>
<tr>
<td>0.3 mm</td>
<td>84</td>
<td>78</td>
</tr>
</tbody>
</table>

2200 mV against a noise of 200 mV, resulting in a signal-to-noise ratio (SNR) of approximately 11 or 21 in dB. Compared to R151 waveform, the response of PWAS was conspicuously crisper in the earlier part corresponding to the S0 mode arrival which was followed by the arrival of slower A0 mode.

Looking at the AE waveforms by PWAS and R151 present in Figure 4, it is noticed though the two waveforms have comparable signal peak amplitudes, PWAS shows higher floor noise (circled part). For AE detection, it is needed that the sensor shows good signal-to-noise ratio. In this case, floor noise for PWAS is too high and needs to be decreased.

An expanded view of the early PWAS response is shown in Figure 5(a). The fast S0 mode and slow but highly dispersive A0 mode are clearly present. The out-of-plane displacement waveform generated by PlotRLQ (PlotRLQ is a computer program that computes theoretical waveforms by summing the contributions of Lamb wave modes excited by specified moment tensor sources at specified depths within the plate based on classic Lamb wave theory) calculated for this situation is shown in Figure 5(b). The agreement between PWAS waveform and theoretical prediction is discernable, especially for the S0 mode, the earliest part of the response. The small discrepancy comes from the ragged appearance at the beginning of the A0 mode and from the relatively abrupt falloff after the A0 reaches its peak amplitude. This may be credited to the PWAS sensor aperture effect as described in [12].
Figure 3: AE testing setup. (a) Test setup schematic; (b) laboratory test setup.

Figure 4: 0.5 mm PLB detection on 1.6-mm aluminum plate. (a) PWAS AE waveform and its frequency spectrum; (b) R15I AE waveform and its frequency spectrum.

Figure 5: PWAS response compared to theoretical prediction. (a) PWAS waveform; (b) theoretical out-of-plane displacement by PlotRLQ.
2.1.2. Steel Plate Tests. The second part of the work was conducted on a 19-mm thick steel plate. 0.5 mm HB PLB was applied on the surface of the plate about 72 mm away from PWAS. The R15I transducer was glued on the plate with a distance of 98 mm from the PLB.

The PLB was detected by PWAS with an AE amplitude of 73 dB in contrast to the 87 dB detected by R15I. The waveforms and their frequency spectra are provided in Figure 6. The PWAS signal showed a strong negative-going spike accounting for the 73 dB amplitude, fitting in a 600 mV peak. The background noise was about 150 mV, resulting in a SNR of approximately 4 or about 12 in dB. The SNR of PWAS in steel plate was much lower than that in aluminum plate.

By examining the frequency spectra, it can be noted that PWAS has major frequency components beyond 200 kHz, showing a wider frequency response compared to resonant type R15I AE sensor.

2.2. PWAS AE Detection in Compact Tension Testing. Compact tension (CT) specimens made of the same material as the steel plate used in Section 2.1 were used. The geometry of the specimens is displayed in Figure 7. Custom fixtures...
were designed and fabricated to mount the CT specimens. The cyclic tension loads of minimum 1 KN and maximum 50 KN were applied to the specimen using an MTS810 servohydraulic mechanical testing machine. Fatigue tests were conducted under load-controlled mode with frequency of 1 HZ. A clip gage was employed to measure the crack mouth opening displacement (CMOD) to clarify crack opening and closure and to determine the magnitude of the CMOD. The surface cracks were also monitored optically with a high-resolution recording microscope. Two separate sets of AE sensors, namely, R15I and PWAS monitored the process, as well. They connected to the Sensor Highway II data acquisition system through preamps. The data from these two sets of sensors were analyzed separately with AEwin software.

The results of crack localization from PWAS sensors (http://www.pacndt.com/products/Remote%20Monitoring/AE SENSOR Highway.pdf) and R15I during CT testing are shown in Figure 8. R15I sensors detected 1,171 AE events prior to failure while PWAS detected only 54 events. Figure 8(a) gives the cumulative acoustic energy of R15I and PWAS together with the crack opening displacement. While PWAS detected a fewer number of acoustic activities, they detected the crack growth when...
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2.3. PWAS Adaptation toward Field Application. In the validation tests, it has been noticed that PWAS on steel specimens exhibited higher floor noise compared to the standard R15I AE sensors, therefore, providing a poor SNR ratio. To enhance the signal quality toward field application and decrease the background floor noise, we improved PWAS installation by using a coaxial cable similar to that used for R15I sensors. The shield of the coax was connected to the steel plate very close to the PWAS while the center conductor was connected to the positive electrode of the PWAS, as shown in Figure 9.

Detection of a 0.3 mm PLB about 20 mm away from a coaxial cable wired PWAS was evaluated (Figure 10). The rest of the setup remained the same as for the tests presented in Section 2.1.2. The resulting waveform had an amplitude of 71 dB, approximately 400 mV. The background floor noise was discernibly decreased and approximated at 10 mV. Therefore, SNR was measured at 40 or 32 dB, significantly improved compared to the 12 dB presented in Section 2.1.2.

3. PWAS Active Mode Crack Sensing

In the dual mode sensing schematic, after significant cracking has been identified by passive mode AE detection, active mode sensing using pitch-catch interrogation is evoked to quantify crack growth through damage index and array imaging. A PWAS network consisting of several sensors spatially distributed on the plate can be used to interrogate the plate with one sensor generating the guided wave and the others receiving the structural response. When an elastic Lamb wave is transmitted and travels through the structure, wave scattering occurs in all directions where there is a change in the material properties due to damage. The scatter signal is defined as the difference between the measurement during the development of damage and the baseline signal at the initial stage. One advantage of using scatter signals is to minimize the influence caused by boundaries or other structural features which would otherwise complicate the Lamb wave analysis.

3.1. Damage Index Evaluation. We assume that cracking is the sole source of changes in the detected Lamb waves. Also being assumed is that the waves travel in straight paths in the plate structures. Hence, the objective of our Lamb wave signal analysis is to extract damage-related characteristics from the measured sensory data. In this research damage index (DI) is defined as [16]

$$DI = 1 - \frac{C_{XY}}{\sigma_X \sigma_Y}.$$  

(1)

$C_{XY}$ is the covariance of $X$ and $Y$ given by

$$C_{XY} = \sum_{j=1}^{N} (X_j - \mu_X) (Y_j - \mu_Y).$$  

(2)

where $\mu$ is the mean value and $N$ is the length of the dataset. $\sigma_X$ and $\sigma_Y$ are the standard deviations of $X$ and $Y$, respectively, with their product given as

$$\sigma_X \sigma_Y = \sqrt{\sum_{j=1}^{N} (X_j - \mu_X)^2 (Y_j - \mu_Y)^2}.$$  

(3)

For the active sensing implemented during the CT test presented in Section 2.2, four PWAS were used to perform pitch-catch wave propagation interrogation, as marked and numbered in Figure 11(a). Using the definition of damage index defined above, the DI curves were plotted at different crack lengths as shown in Figure 11(b) for all the pitch-catch paths. The DI increases when the crack grows. The detection along sensors P0 to P1 is most sensitive, followed by the one
Figure 11: AE detection on a 1/2" CT specimen. (a) Geometry of the specimen and arrangement of transducers; (b) a snapshot of the actual specimen. AE PWAS are circled (rest are for active sensing). The R15I was installed on the other side of the specimen.

Figure 12: Array imaging for crack growth detection. (a) Test specimen and sensor network; (b) array imaging of a hairline crack of 18 mm length. Sensor locations are marked as green face circles.

along sensors P1 to P3 since the paths are perpendicular to the crack development path. The increment of the DI curves is also well correlated to the crack growth under fatigue loading.

3.2. Array Imaging. The array imaging methods have the incredible capability to map the structure and existing damage in it, providing a means to qualitatively assess the structural integrity. The sparse array uses scatter signals from a network of sensors to construct a diagnosis image. The image construction is based on triangulation principle and conducted by shifting back the scatter signals at time quantities defined by the transmitter-receiver locations used in the pitch-catch mode. Assuming a single damage scatter is located at point $Z(x, y)$ in the structure, the scatter signal from transmitter $T_i$ to receiver $R_j$ contains a single wave packet caused by the damage. The total time of traveling $\tau_Z$ is determined by the locations of the transmitter $T_i$ at $(x_i, y_i)$, the receiver $R_j$ at $(x_j, y_j)$, and $Z(x, y)$, as

$$
\tau_Z = \frac{\sqrt{(x_i - x)^2 + (y_i - y)^2} + \sqrt{(x - x_j)^2 + (y - y_j)^2}}{c_g},
$$

(4)

where $c_g$ is the group velocity of the traveling Lamb wave, assuming constant. When a wave packet is shifted back by the quantity defined by the transducers and the exact position of the damage, that is, $\tau_Z$, ideally the peak will be shifted right back to the time origin. If the wave packet is shifted by a quantity defined with otherwise cases (such as $\tau_i$ and
\( \tau_0 \), the peak will not be shifted right at the time origin. For an unknown damage, for a certain scatter signal with \( \tau_z \), the possible locations of the damage constitute an orbit of ellipse with the transmitter and receiver as the foci. To locate the damage, ellipses from other scatters (or transmitter-receiver pairs in the network) are needed (triangulation). For a given network of \( M \) transducers, a total of \( M^2 \) scatter signals will be used without considering reciprocity. The pixel value at the location \( Z(x, y) \) is defined as

\[
P_Z(t_0) = \prod_{i=1}^{M} \prod_{j=1}^{M} s_{ij}(\tau_z), \quad i \neq j,
\]

where \( s_{ij} \) is the scatter signal obtained from \( i \)th transmitter and \( j \)th receiver. Details of principles and applications of ultrasonic array imaging can be found in [17].

The crack detection demonstration was conducted on a 1-mm thick aluminum plate. The imaging was performed by a four-PWAS network to assess a simulated hairline crack centered inside, shown in Figure 12(a). A hairline crack of 18 mm was made on the plate, and then crack length was increased to 23 mm, resulting in a growth of 5 mm. The Lamb wave used to conduct the imaging was the S0 mode at 310 kHz with a wavelength about 17 mm.

The image result of the crack at 18 mm is shown in Figure 12(b), giving a clear detection with two highlighted spots representing the two crack tips; therefore, they could be used to estimate the crack length and monitor the crack growth. The first crack of 18 mm was estimated at 17.09 mm and the second one of 23 mm was estimated at 22.47 mm. A crack growth of 5.38 mm was hence measured with the array imaging method with an error of about 7.6% compared to the actual growth of 5 mm (from 18 mm to 23 mm).

4. Conclusions

Piezoelectric wafer active sensors (PWAS) have made tremendous progress in structural health monitoring during the past decade, but their exposure to civil infrastructure has been little discussed so far. The work presented here intends to explore PWAS applications for in-field monitoring of infrastructure (e.g., civil steel bridges) using both acoustic emission and active wave propagation sensing. Laboratory demonstration on both thin aluminum and thick steel plates, the PWAS have been proved as AE sensors. The use of coaxial wiring cables has greatly improved the PWAS waveform signal-to-noise ratio, making it more suitable for field application. The PWAS AE sensing of fatigue cracking on a CT specimen showed that it can provide concentrated detection around the crack tip with a relatively fewer numbers of acoustic emission events than the R151 AE transducers. The PWAS active mode sensing using propagating guided waves can interrogate the structures at will and provide a clear indication and quantitative estimation of the crack growth through damage index or array imaging. The dual mode sensing of the presented PWAS methodology has shown its promising application to insitu health monitoring and diagnosis of steel bridges.

Acknowledgments

This paper was performed under the support of the US Department of Commerce, National Institute of Standards and Technology, Technology Innovation Program, Cooperative Agreement number 70NANB9H9007. The authors would also like to thank Dr. Adrian Pollock from Mistras Group for his insightful comments on the presented passive sensing.

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