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Open cracks depth sizing by multi-frequency laser stimulated lock-in thermography combined with image processing

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Highlights

the multi-frequency lock-in infrared thermography method is used to evaluate perpendicular crack depths.

the experimental curves of the spatial second derivative (Laplacian) of the amplitude and of the phase of the infrared signal are analyzed as a function of the thermal diffusion length.

direct depth indicators depending on both the Laplacian of amplitude and phase images are introduced. experimental measurements validate the method for crack depths ranging from about 0.5 to 3 mm in metallic material.

Abstract

The multi-frequency lock-in thermography is coupled with efficient image processing to analyse infrared open crack footprint. It has been shown that the evolution of the Laplacian of modulated surface temperature amplitude image as a function of the diffusion length allows to estimate the depth of surface open defects but requires calibration abacus obtained by finite element method simulations. In this work, a method is proposed to avoid the tedious use of abacus by introducing an indicator of the crack depth using a simple expression depending on the Laplacian of both amplitude and phase images. This multi-frequency method is presented through numerical simulations. Besides, the analysis of experimental results obtained on artificial and real open vertical cracks in metallic samples show that the depth of the defects can be directly estimated.

Keywords: lock-in; thermography; crack depth; modulation frequency; infrared images; non destructive control.

1. Introduction

Structures submitted to cyclic mechanical loading can develop cracks which can be highly damaging. The depth of a crack is an important information to access its dangerousness. But this information is not easy to obtain. Liquid penetrant, the most conventional method to detect open cracks [1], cannot give worth quantitative information on crack depths. Quantitative estimation can be provided by using

methods like ultrasonically stimulated thermography, eddy current stimulated thermography or laser thermographic imaging [2-5]. But these techniques, which are based on the analysis of the generated local rise of the crack temperature, need calibration procedures, a reference response and/or blackened sample to evaluate the crack depth [6-16].

Lock-in thermography uses lateral thermal flows in order to detect vertical cracks [12-15]. The amplitude and phase of infrared images obtained by the diffusion of lateral thermal flows provide information concerning defects with a good contrast for vertical cracks. Reference [17] proposes an analytical solution of the temperature distribution to deduce the crack width for infinite vertical cracks.

In this work a lock-in thermography procedure is used for depth evaluation of surface-breaking cracks in metals without surface preparation and with no calibration procedure. The implementation of the approach is described in [18-21]: taking advantage of the thermal diffusion length to probe different regions of the sample, the analyse of the amplitude images at the modulation frequency allows to provide crack depth determination but this requires abacus or back simulations to obtain quantitative information. The improvement provided by this work is to process the phase images in addition to the amplitude images to directly allow the determination of depth indicators, avoiding thus the calibration abacus used up to now.

In the first part of this article, the study is led from results of 3D FEM numerical simulations to introduce two depth indicators (μ_A and μ_ϕ) which rely on the evolution of the second spatial derivative of both amplitude and phase images as a function of the thermal diffusion length.

In the second part of this article, experimental results obtained on constant controlled "cracks" in steel and aluminum blocks are analyzed and a global depth indicator I_h calculated from μ_A and μ_ϕ is proposed. Finally, this global indicator I_h is applied to Inconel samples exhibiting real fatigue cracks.

2. Multi-frequency lock-in infrared thermography method

2.1. Principle of the method

The presented method uses a modulated heat excitation associated with lock-in thermography. When a modulated circular laser spot excitation heats the sample surface, the 3D heat transfer equation in solids in the absence of internal heat generation can be used to evaluate the temperature distribution in the whole solid [22]. The lock-in processing of the signals allows a retrieval of both amplitude and phase of the sinusoidal infrared radiation component from the surface sample at the heating fundamental frequency. The unwanted continuous heating (DC) component is thus filtered out. The second spatial derivatives (Laplacian) of both amplitude and phase images highlight the thermal disturbance caused by the presence of a vertical crack. Crack pixels are extracted from the Laplacian of the amplitude image according to the numerical procedure presented in [23].

In a modulated heating process, the thermal diffusion length, which can be written as

$$\mu = \sqrt{\frac{\alpha}{\pi f}} \quad (1)$$

where α is the thermal diffusivity of the specimen, can be adjusted by the modulation frequency f of the thermal excitation source and used as a 3D probe of the crack depth h . The crack signature is analyzed via the changes of the Laplacian of the amplitude and phase images as a function of the thermal diffusion length μ .

The absolute value of the Laplacian of the amplitude image normalized by its asymptotic value is denoted A_N and the value of the Laplacien of the phase image is denoted A_φ . In [18,19], it is shown that $A_N(\mu)$ at the crack pixels reach their maximal value at a thermal diffusion length increasing with h and that the evolution of $A_N(\mu)$ is independent of the crack width. So the evolution of A_N with μ is then an indicator of the crack depth. In [19], simulated $A_N(\mu)$ for various excitation frequencies are used as an abacus to deduce the depth of real cracks. However, these simulations are time consuming (several hours) and must be done for every kind of sample. Besides obtaining the results

more quickly, the results do not depend on modeling accuracies. It is thus a great interest to have a direct indicator, independent of the material and its surface condition, to estimate the crack depth.

2.2. Simulated experiments

The objective of this work is to combine the analysis of $\Lambda_N(\mu)$ and $\Lambda_\phi(\mu)$ of crack responses to easily deduce quantitative indicators of the crack depth without any simulation abacus. The heat source is focused onto the surface in a circular and uniform spot of radius r at a distance d_0 of a linear crack perpendicular to the surface of width w , length L and constant depth h . The heat source is intensity modulated at a frequency f . Amplitude and phase images of the modulated surface temperature at the frequency f are calculated for different h values by 3D FEM numerical simulations. Figure 1 shows an example of amplitude (Figure 1a) and phase (Figure 1b) images of 400*400 pixels (1 pixel sees 50 μm *50 μm) obtained at $f = 0.8$ Hz. The linear crack footprint is clearly visible in the figures above the heat source. A set of pixels $P(x,y)$ located on the crack which faces the heating zone are extracted from the Laplacian of the amplitude image with a basic numerical processing described in [23] (Figure 2a). Figure 2b shows the Laplacian of the phase for these $P(x,y)$ pixels at $f=0.8$ Hz. The distance between a pixel $P(x,y)$ and the center of the heating spot is denoted as $d(x,y)$.

2.2.1 One point analysis

The evolution of the Laplacian of the amplitude Λ_N as a function of the thermal diffusion length at the closest pixel P to the heating spot ($d(x,y)=d_0$) is reported in Figure 3. Simulation results are presented for four depths h as a function of $\mu_i=\sqrt{\alpha/\pi f_i}$ for a frequency ranging from 0.05 Hz to 1.5 Hz.

The polynomial fits $f_h(\mu)$ of $\Lambda_N(\mu_i)$ show a shift along the μ axis which depends on the simulated depth h . Afterward in this work, the value of μ_A corresponding to $f_h(\mu_A)=0.7$ is then considered as a first indicator for the crack depth evaluation.

The evolution of the absolute value of the Laplacian of the phase images at the same pixel P can also be analyzed. One notices in Figure 4 that the polynomial fits of simulated $\Lambda_\phi(\mu)$ reach a maximum for

$\mu=\mu_\varphi$ which is of the order of the crack depth. μ_φ is thus retained as a second indicator for the crack depth evaluation.

Figure 5 shows that the indicator μ_φ deduced from the simulations seems to be more dependant on the crack depth h than the indicator μ_A .

2.2.2. Synthesis of crack point analysis

Let's consider the responses of 50 selected pixels on the crack at $f = 1.5$ Hz. Figure 6 displays μ_A and μ_φ as a function of the distance d between a selected pixel and the center of the heating spot for the four simulated depths h .

One notices in Figure 6 that μ_A is more d dependant than μ_φ . Therefore a combination of μ_A and μ_φ is proposed to minimize the influence of d and thus directly estimate the local depth for each selected pixels.

3. Experimental results for the evaluation of the depth

3.1 Experimental set-up and methodology

The experimental set up (Figure 7) includes a heat source (a diode-pumped Ytterbium laser with 830 nm wavelength and tuneable power), a signal generator which modulates the intensity of the heat source at frequency f , and an infrared camera (Jade III, CEDIP, FLIR). The camera has an array (240*320 pixels) of InSb detectors sensitive in the 3-5 μm wavelength range.

The laser beam is focused onto the surface in the vicinity of the crack in an approximately uniform and circular spot of 1 mm diameter. No surface preparation is required for the studied samples. The laser power is adjusted to the thermal response of the samples.

The generated infrared signals measured by the camera and the reference frequency f are sent to a lock-in detection module (FLIR R9902). The amplitude and phase images of the f -component of the IR

emission induced at the surface of the inspected sample is then extracted with a fairly good contrast (Figure 8a and Figure 8b).

The Laplacian of amplitude and phase images (Figure 8c and 8d) are proceeded according to the procedure described in the former section to calculate μ_A and μ_φ from the polynomial fits of $A_N(\mu)$ and $A_\varphi(\mu)$ (Figure 9). Notice that local diffusivity α of each sample can be directly extracted from measurements [24], in a homogeneous area sufficiently far away from the crack.

3.2 Experimental measurements with controlled artificial cracks

Artificial cracks with controlled dimensions are good samples to validate experimentally the simulated results presented in section 2.

Two 40*40*20 mm³ steel and aluminum alloy plates separated by a brass sheet of 30 µm thickness are assembled with screws to simulate a sample with a vertical open crack (Figure 10). In the presented results, the brass sheet was cut in z direction so as to leave an air gap of controlled constant depth between two steel plates (Samples F, thermal diffusivity equal to $3.7 \pm 0.2 \text{ } 10^{-6} \text{ m}^2/\text{s}$) or two aluminum alloy blocks (sample G, thermal diffusivity equal to $40 \pm 2 \text{ } 10^{-6} \text{ m}^2/\text{s}$) as described in Table 1. The laser power is 4 W .

Figure 11 shows μ_A and μ_φ as a function of the distance d obtained for the four studied samples. One can observe that μ_A and μ_φ follow the same trends as simulation results (Figure 6) : μ_A and μ_φ are both dependant on h and μ_φ is less d dependant than μ_A but μ_φ is noisier than μ_A in the presented experimental results.

3.3 Global depth indicator

In order to minimize the dependence on d of μ_A and μ_φ and to give an enhanced robustness to the depth assessment at each selected pixel, a new depth indicator I_h depending on μ_A , μ_φ and d is defined as:

$$I_h(x, y) = \frac{\mu_A(x, y) * \mu_\varphi(x, y)}{a * d(x, y)} \quad (2)$$

where a is an adjustment parameter. For depth range from 0.5 mm to 3 mm and for the experimental conditions, $a=2$ is well adapted as shown hereafter.

Figure 12 shows the depth indicator I_h obtained with the presented multi-frequency lock-in thermography method using (2) with $a=2$ for the four samples. The results are presented in x,y plane with the same color scale to give a comparative and visual indication of the depth for each selected pixels of the 4 cracks. The experimental results seem coherent with the expected h values.

Figure 13 displays the average value over the selected pixels $P(x,y)$ of $\mu_A(x,y)$, $\mu_\varphi(x,y)$ and $I_h(x,y)$, respectively denoted as $\tilde{\mu}_A$, $\tilde{\mu}_\varphi$ and \tilde{I}_h obtained for the four samples as a function of expected depths. Linear regressions through the origin are calculated for the three indicators. The R-squared coefficient R^2 which represents the correlation between the predictive value and the average of the indicators show that \tilde{I}_h indicator is the better indicator (R^2 near 1). The negative R^2 coefficient obtained for $\tilde{\mu}_A$ is nonsense and indicates that the linear fitting through the origin is not appropriate. As μ_A and μ_φ are correlated measurements, the error bars of the \tilde{I}_h indicators are smaller than the error bars of $\tilde{\mu}_A$ and $\tilde{\mu}_\varphi$ indicators. One can see that the indicator \tilde{I}_h is comparable to the expected depth h , though a bias exists for small depths.

The results validate the method to evaluate constant crack depths within the tested depth range from 0.7 to 2.5 mm.

3.4 Inconel samples with real fatigue cracks

Five open vertical fatigue cracks of 35 to 65 μm widths in a highly reflecting $130*27*9 \text{ mm}^3$ Inconel plates (samples A, B, C, D, E) are opened over the entire width of the samples. Depths are measured on the 2 lateral sides of the sample using optical microscopy [18] (Table 2). The measured thermal diffusivity of the Inconel sample is equal to $3.3 \pm 0.1 \text{ } 10^{-6}\text{m}^2/\text{s}$. The laser power is 2 W.

Figure 14 shows I_h using (2) with $a = 2$ in (x,y) planes for Inconel samples.

Figure 15 presents the averaged indicators $\tilde{\mu}_A$, $\tilde{\mu}_\varphi$ and \tilde{I}_h calculated from selected pixels of the crack as a function of expected averaged depth h . Linear regressions through the origin are calculated for the three indicators. The R-squared coefficient R^2 is nearer to 1 for $\tilde{\mu}_\varphi$ than for \tilde{I}_h and has no sense for $\tilde{\mu}_A$. If the result obtained for sample E ($h>3\text{mm}$) is ignored, the linear regression for \tilde{I}_h gives comparable results to those obtained for steel and aluminum samples (figure 13): $R^2=0.92$ and $y=0.92 x$.

Consequently, \tilde{I}_h seems to be a good direct indicator of the depth h at least for h lesser than 3 mm.

Although it seems that the sensitivity of the measurement reaches its limit for Sample E, the proposed multi-frequency lock-in thermography approach appear to be promising to locally probe local crack depth without sample modelling .

Conclusion

In this work, the multi-frequency lock-in infrared thermography method is used to evaluate perpendicular crack depths. A specific empirical data analysis is presented using the experimental curves of the spatial second derivative (Laplacian) of both amplitude and phase infrared images as a function of the thermal diffusion length. Two direct depth indicators μ_A and μ_φ are introduced and their evolution as a function of the crack depth h and the distance d of the crack to the heating spot are analyzed. A combination of μ_A , μ_φ and d is used to make a depth indictor I_h independent of distance d . It can be seen as a compromise which enhances the robustness of the depth evaluation.

Measurements were carried out on Inconel alloy samples with calibrated 0.5 mm to 3.5 mm crack depths. The measurement duration depends on the number of used frequencies. Typically for metals, one can consider 5 different frequencies between 0.05 Hz and 1.5 Hz which lead to 10 minutes measurement duration. The distance of heat source to the inspected crack and the frequencies selected to probe the volume of the sample must be thoroughly chosen so that the sensibility is optimal. This distance depends on the thermal diffusivity of the sample. The depth range for the studied material is from about 0.5 to 3 mm.

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Z. Qin

No photo for S. Beuve

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JP Roger

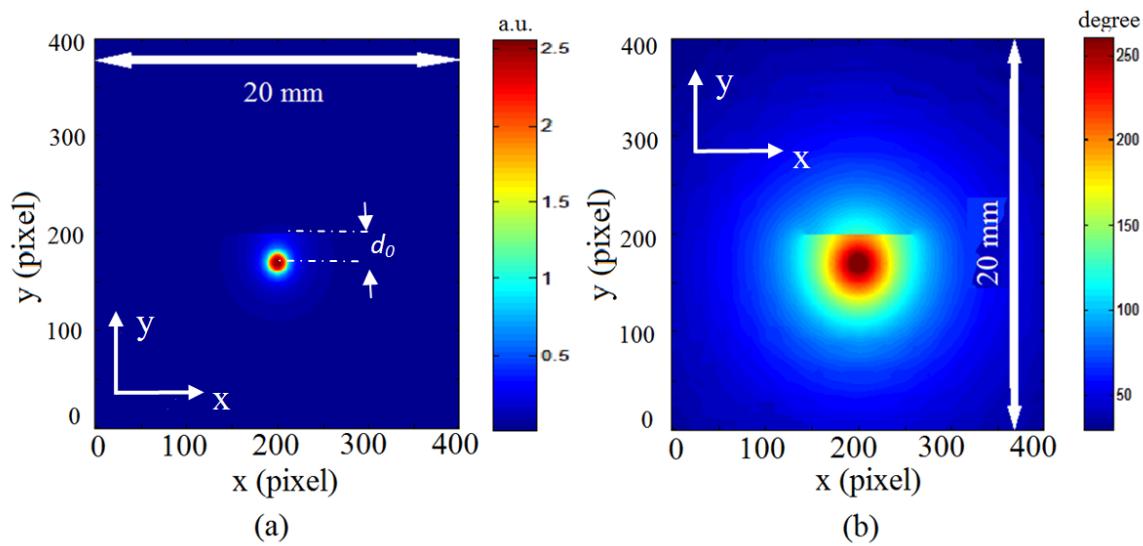


Figure 1: (a) amplitude image, (b) phase image.

Simulation parameters are $w = 30 \mu\text{m}$, $h = 1 \text{ mm}$, $L = 6 \text{ mm}$, $r = 0.5 \text{ mm}$, $d_0 = 1.5 \text{ mm}$, $\alpha = 3.3 \cdot 10^{-6} \text{ m}^2/\text{s}$, $f = 0.8 \text{ Hz}$. a.u.: arbitrary unit.

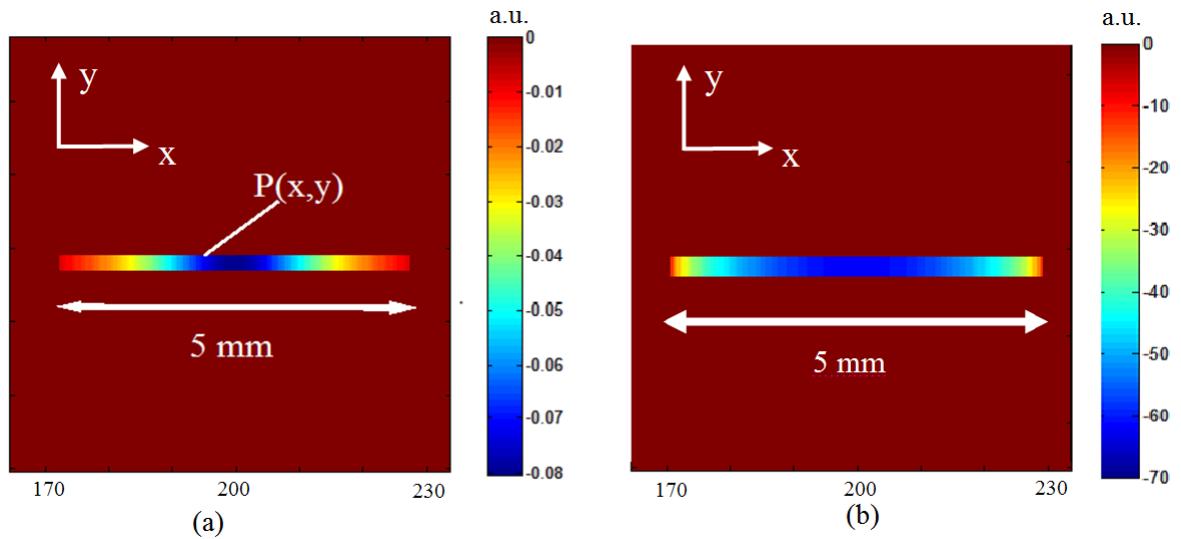


Figure 2: Pixels $P(x,y)$ extracted from the Laplacian of Figure 1 images ($f = 0.8$ Hz),
(a) Laplacian of the amplitude. (b) Laplacian of the phase. a.u.: arbitrary unit.

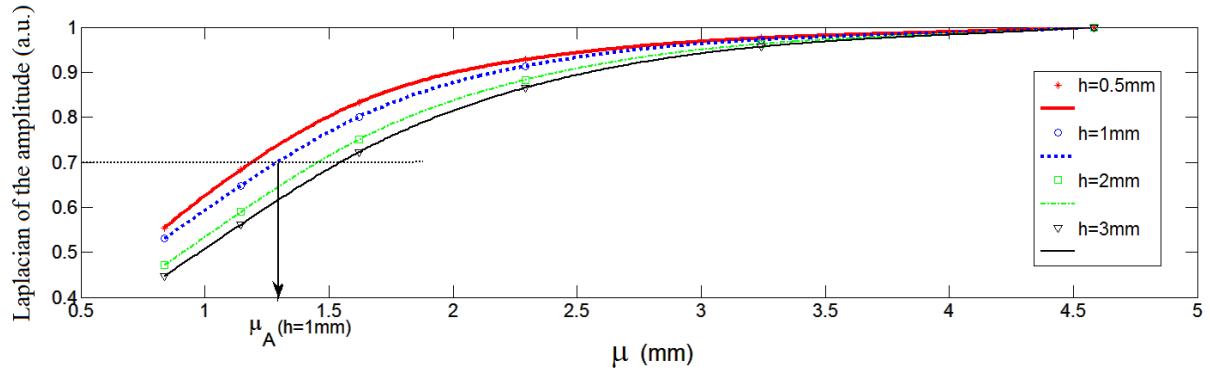


Figure 3: Simulated $\mathcal{A}_N(\mu_i)$ and polynomial fit $f_h(\mu)$ at the pixel P ($d(x,y)=d_0$) for different depth h . Simulation parameters are $r = 0.5$ mm, $\alpha = 3.3 \cdot 10^{-6}$ m²/s, $d_0 = 1.5$ mm, $w = 30$ μ m, $L = 6$ mm.

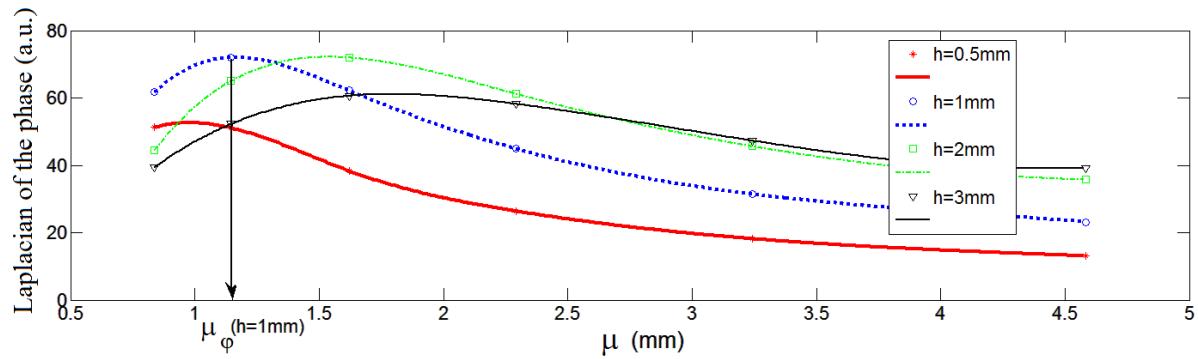


Figure 4: simulated $A_\varphi(\mu_i)$ and polynomial fits at the pixel P ($d(x,y)=d_0$) for different depths h .
Simulation parameters are $r = 1 \text{ mm}$, $\alpha = 3.310^{-6} \text{ m}^2/\text{s}$, $d_0 = 1.5 \text{ mm}$, $w = 30 \mu\text{m}$, $L = 6 \text{ mm}$.

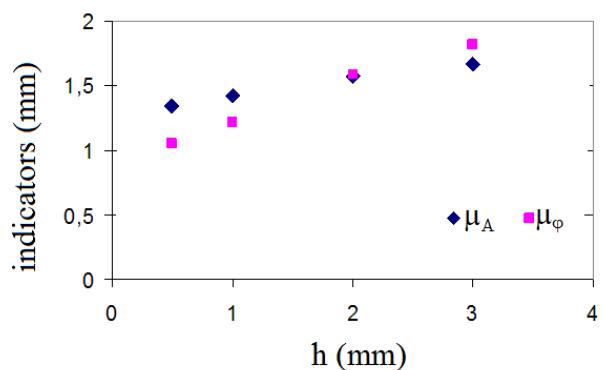


Figure 5: μ_A and μ_φ as a function of simulated depth h at the pixel P ($d(x,y)=d_0$).

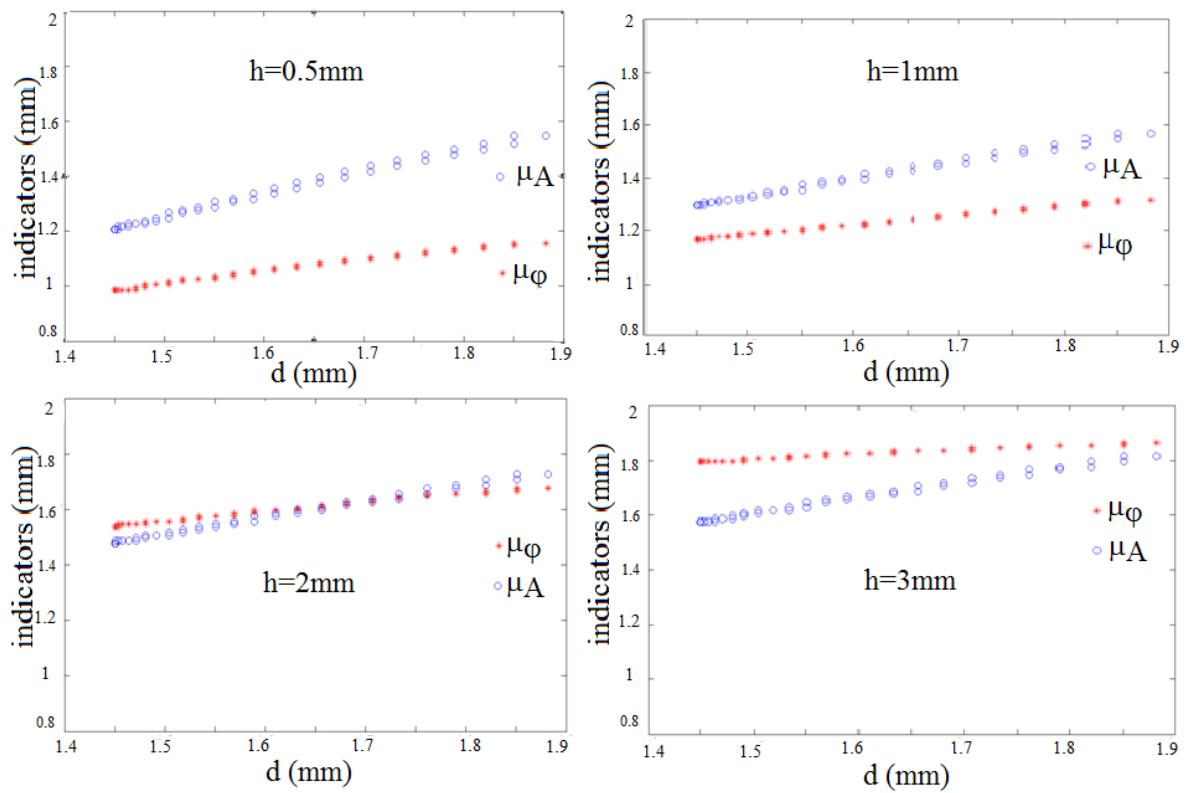


Figure 6 : μ_A and μ_φ as a function of the distance d for the 4 simulated depths h

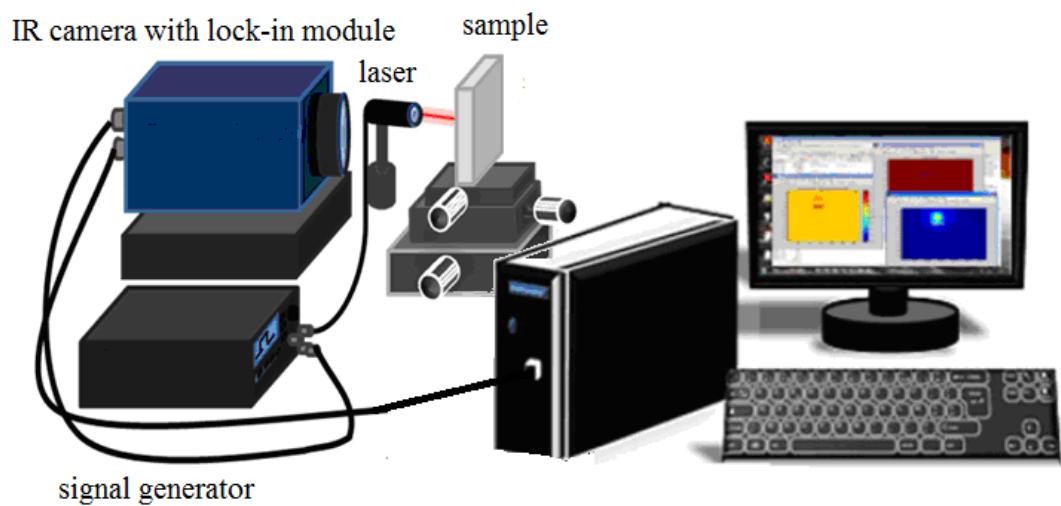


Figure 7: experimental set-up.

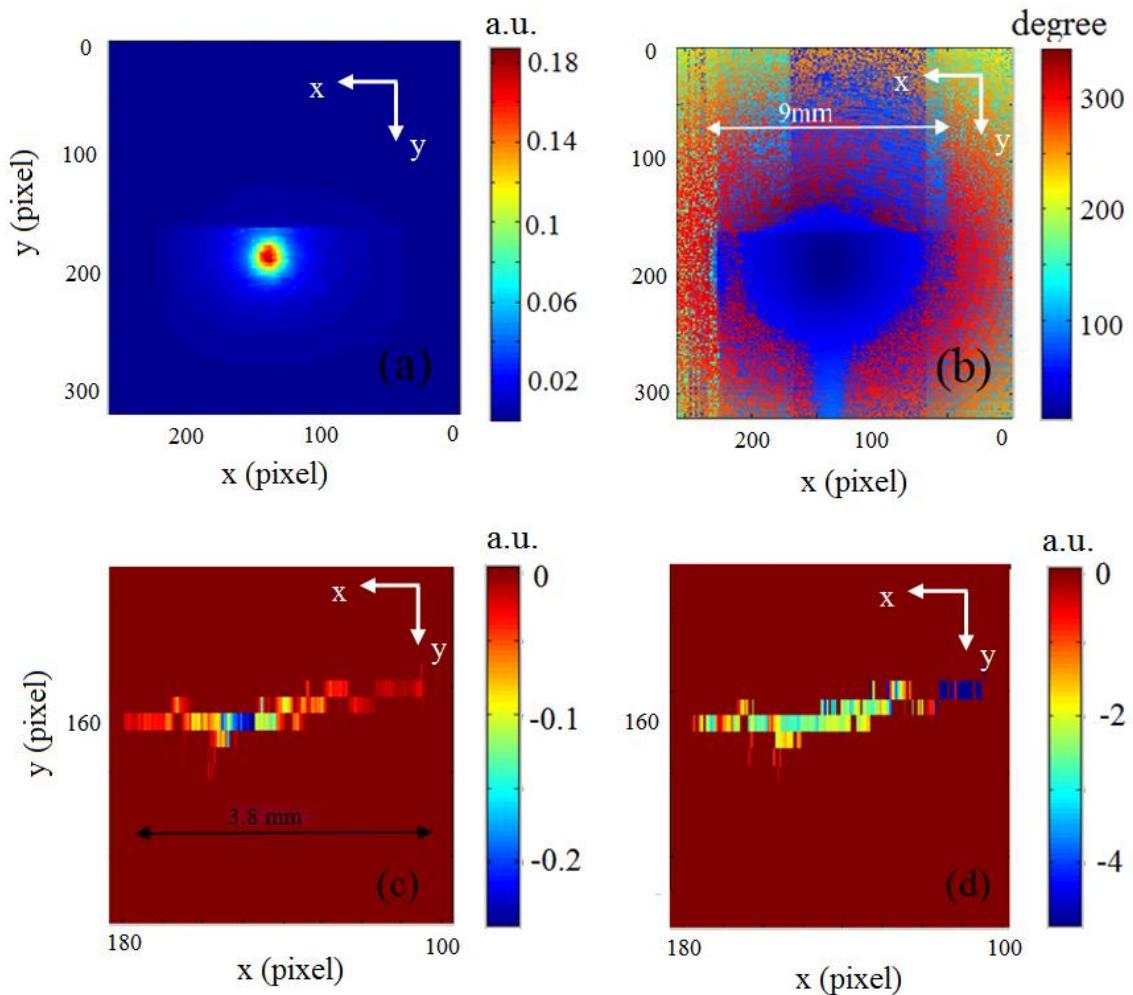


Figure 8: example of a fatigue crack. (a) Amplitude image, (b) Phase image, (c) Laplacian of the amplitude of the selected $P(x,y)$ pixels calculated using Figure 8a, (d) Laplacian of the phase for the selected pixels calculated using Figure 8b. $f = 0.05$ Hz, 1 pixel corresponds to $50 \mu\text{m} \times 50 \mu\text{m}$. a.u.: arbitrary unit.

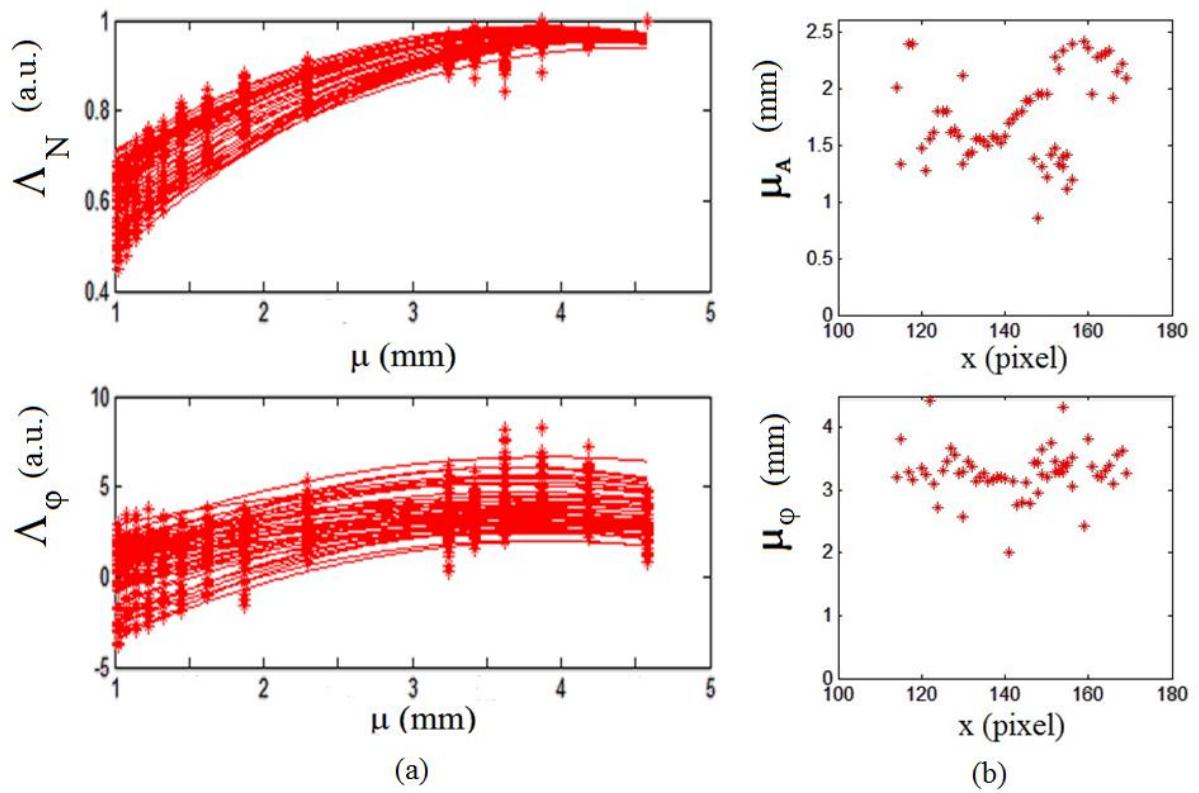


Figure 9: (a) Experimental Λ_N and Λ_ϕ and polynomial fits as a function of μ for the selected pixels in the sample of figure 8. (b) Deduced μ_A and μ_ϕ for the selected pixels along the x direction.
 f is varying from 0.05 Hz to 1.5 Hz.

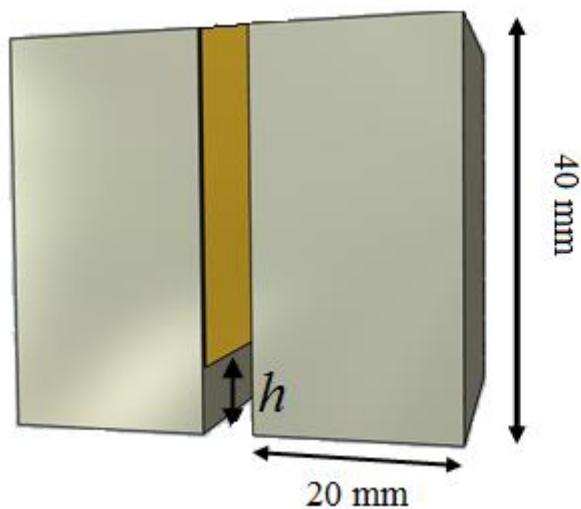


Figure 10: simulation of a controlled depth crack in a steel or aluminum alloy sample. The brass sheet exhibits a constant h cut.

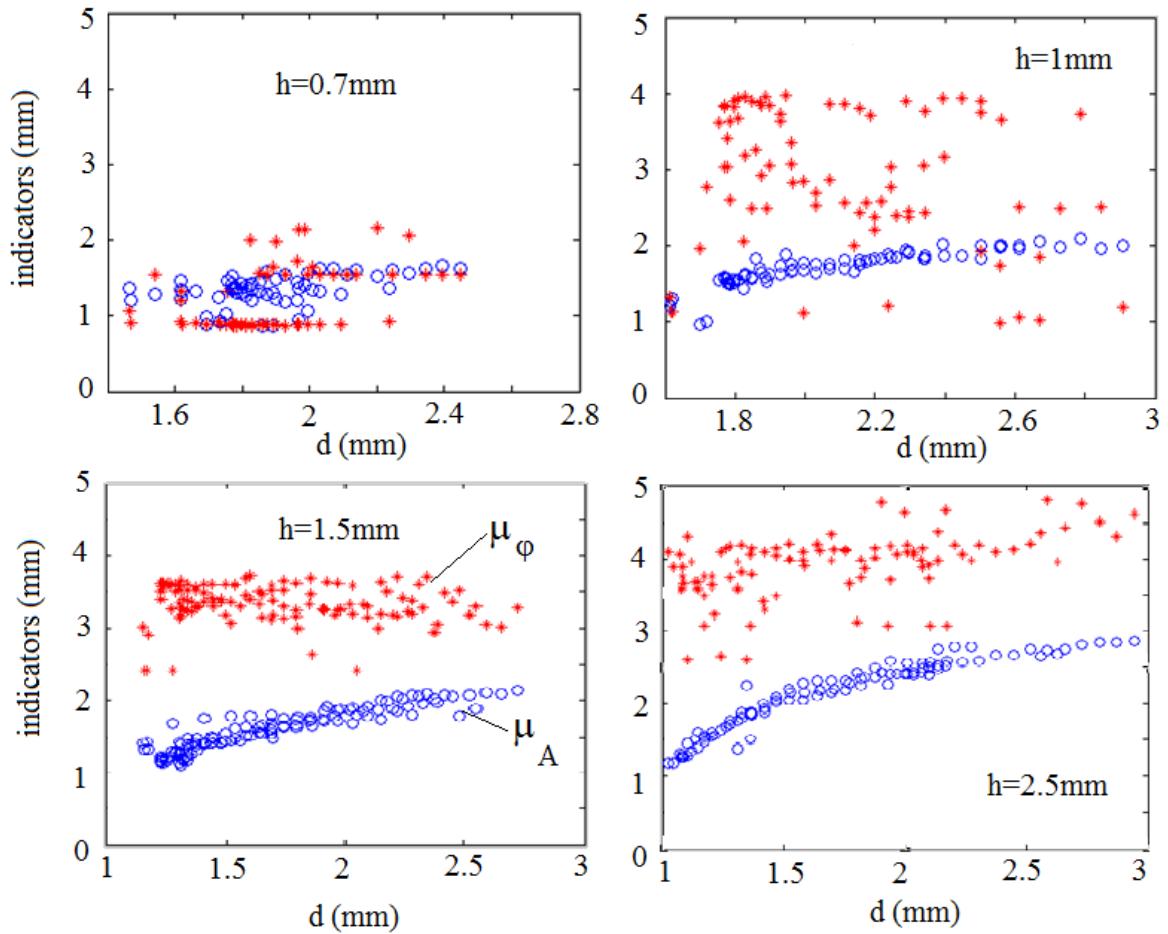


Figure 11: μ_A (blue circle) and μ_ϕ (red star) responses as a function of distance d for four depth h .

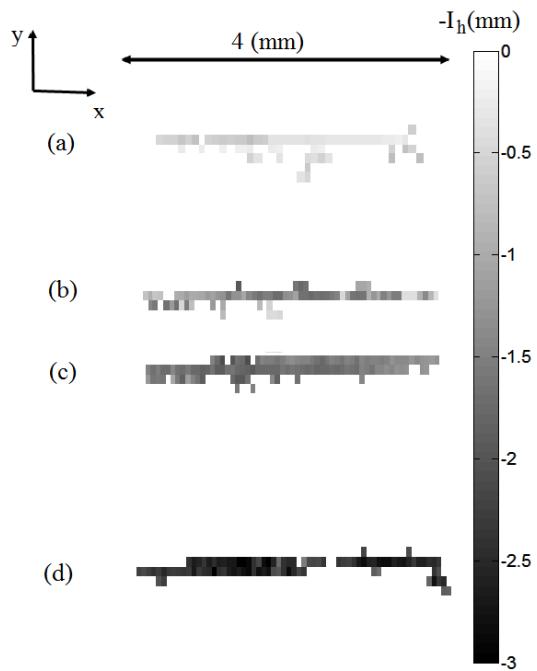


Figure 12: I_h for the selected pixels in x,y plane. 1 pixel represents $80 \mu\text{m} \times 80 \mu\text{m}$, $a = 2$.

The color indicates (negative) I_h evaluation of each selected pixels. (a) Steel sample F1, expected depth around 0.7 mm, (b) Aluminum alloy sample G, expected depth around 1 mm, (c) Steel sample F2, expected depth around 1.5 mm, (d) steel sample F3, expected depth around 2.5 mm.

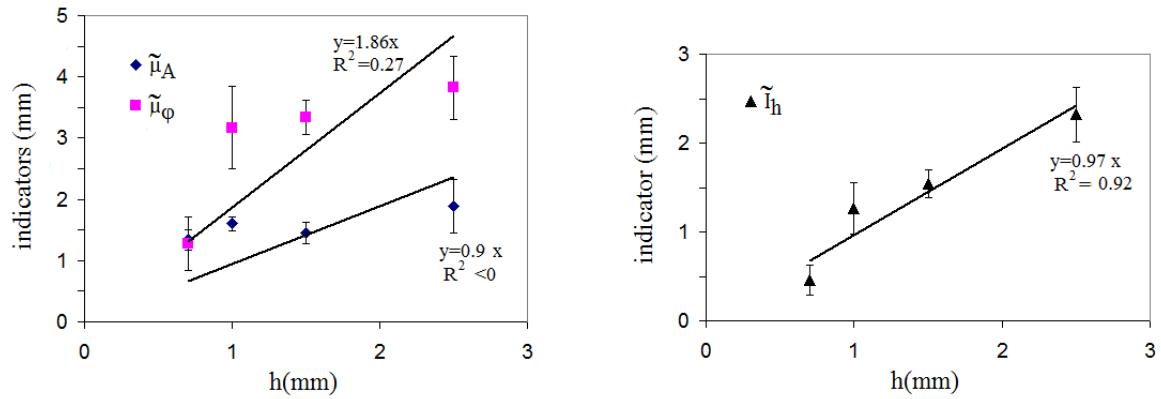


Figure 13: $\tilde{\mu}_A$, $\tilde{\mu}_\phi$ indicators (left) and \tilde{I}_h indicator (right) as a function of expected h values obtained on the steel and aluminum alloy samples. Estimate slope of linear regression through the origin and R-squared coefficient R^2 are added for each indicators.

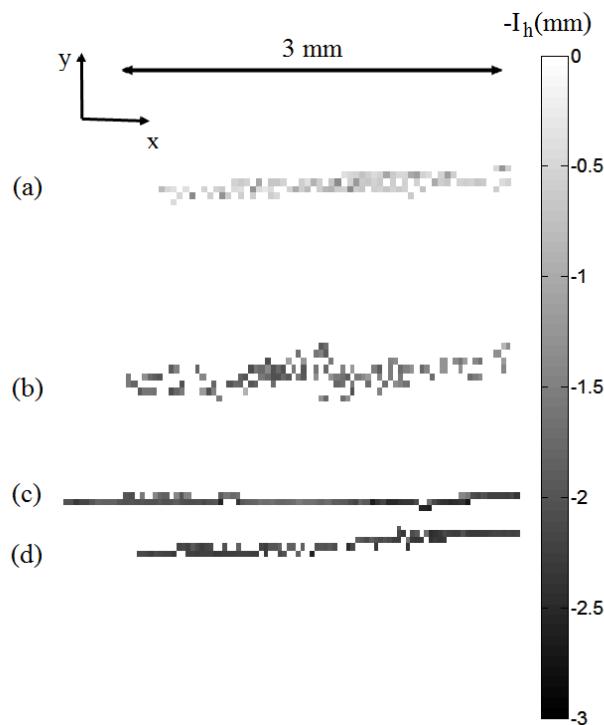


Figure 14: I_h for the selected pixels as a function of pixel location in (x,y) plane,
1 pixel represents $50 \mu\text{m} \times 50 \mu\text{m}$. (a) Inconel sample A, expected depth around 0.4 mm, (b) Inconel sample B,
expected depth around 1.6 mm, (c) Inconel sample D, expected depth around 2.1 mm, (d) Inconel sample E,
expected depth around 3.2 mm.

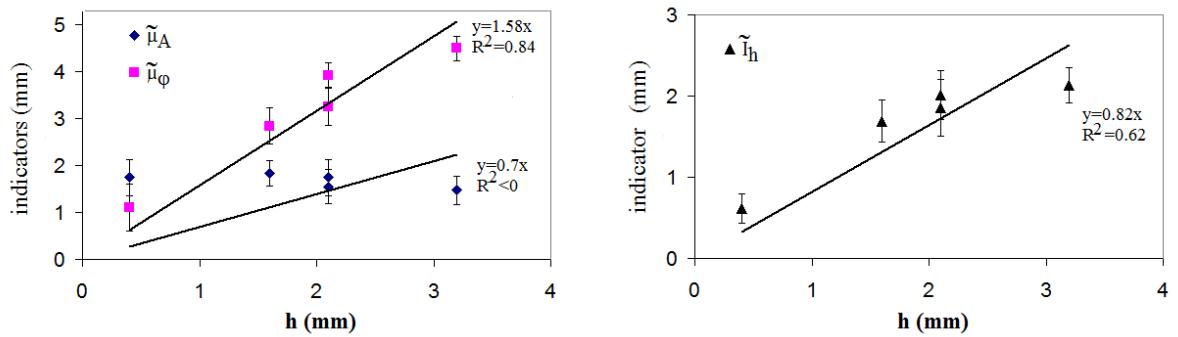


Figure 15: $\tilde{\mu}_A$, $\tilde{\mu}_\phi$ indicators (left) and \tilde{I}_h indicator (right) as a function of expected h values obtained on the Inconel alloy samples. Estimate slope of linear regression through the origin and R-squared coefficient R^2 are added for each indicators.

Table 1: crack sizes of steel and aluminum alloy samples

	Controlled depth h (mm)	crack width (μm) measured by optical microscopy
F1 - steel	0.7 ± 0.1	30 ± 5
G - aluminum alloy	1 ± 0.1	30 ± 5
F2 - steel	1.5 ± 0.1	30 ± 5
F3 - steel	2.5 ± 0.1	30 ± 5

Table 2: crack sizes of Inconel alloy samples

Inconel	Depth (mm) on the two lateral sides of the sample		averaged depth (mm) deduced from the values of the first columns	crack width (μm)
Sample A	0.2 \pm 0.1	0.6 \pm 0.1	0.4	20 \pm 10
Sample B	1 \pm 0.1	2.1 \pm 0.1	1.6	40 \pm 10
Sample C	1.1 \pm 0.1	3.1 \pm 0.1	2.1	35 \pm 5
Sample D	1.7 \pm 0.1	2.4 \pm 0.1	2.1	65 \pm 5
Sample E	2.8 \pm 0.1	3.6 \pm 0.1	3.2	45 \pm 5