

FINAL REPORT FOR CALIBRATED SYNTHETIC APERTURE SONAR

DEREK R. OLSON AND DANIEL C. BROWN

Current high-frequency synthetic aperture sonar (SAS) imaging systems and processing software have been largely successful in producing high-quality imagery in a variety of oceanographic and seafloor environments. In most high-frequency real aperture and SAS systems, the image pixel values have been adjusted for optimal viewing by operators. The dynamic range of the pixel values may be truncated and the mean background level may vary from image to image, even across images of the same geographic area. Automated Target Recognition (ATR) algorithms that process these images require that pixel energy levels are normalized to remove systematic variations in amplitude that can cause these algorithms to fail. However, the adaptive nature of image normalization algorithms can introduce artifacts as well. In many cases the discarded original pixel values may contain valuable information. Calibrated SAS images, defined as images whose pixel values represent only the scattering properties of the seafloor or target, present a promising input to ATR because 1) they preserve amplitude information, which may provide stable feature measurements to ATR, and 2) they can be normalized without artifacts by taking into account propagation and scattering physics, potentially offering a superior method than current normalization techniques.

For this grant, measured system parameter were used to form calibrated images from a high frequency synthetic aperture sonar system. Systems with large bandwidth and wide beams cannot directly estimate the scattering cross section, since that quantity is defined using a plane wave of a single frequency for the incident field. Correspondingly, the quantities produced by calibrated images must be understood as quantities averaged over the systems bandwidth and beamwidth. Details of this method, and scattering cross section results are presented in the enclosed OCEANS2019 conference paper entitled “Analysis of Backscatter Measurements from Calibrated Synthetic Aperture Sonar Images” by Peter D. Romain, Derek R. Olson, and J. Tory Cobb.

Another aspect explored in this grant is to determine whether the scattering cross section estimated using a broadband system is independent of the system resolution, or bandwidth over which the system forms the image. To answer this question, a numerical method was developed using the Helmholtz integral theorem and Fourier synthesis to estimate the scattering cross section in the time domain from a finite ensemble of rough surfaces. This work is detailed in the enclosed manuscript that was submitted to the Journal of the Acoustical Society of America, entitled “Resolution dependence of rough surface scattering using a power law roughness spectrum” by Derek R. Olson and Anthony P. Lyons.

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14. ABSTRACT Current high-frequency synthetic aperture sonar (SAS) imaging systems and processing software have been largely successful in producing high-quality imagery in a variety of oceanographic and seafloor environments. In most high-frequency real aperture and SAS systems, the image pixel values have been adjusted for optimal viewing by operators. The dynamic range of the pixel values may be truncated and the mean background level may vary from image to image, even across images of the same geographic area. Automated Target Recognition (ATR) algorithms that process these images require that pixel energy levels are normalized to remove systematic variations in amplitude that can cause these algorithms to fail. However, the adaptive nature of image normalization algorithms can introduce artifacts as well. In many cases the discarded original pixel values may contain valuable information. Calibrated SAS images, defined as images whose pixel values represent only the scattering properties of the seafloor or target, present a promising input to ATR because 1) they preserve amplitude information, which may provide stable feature measurements to ATR, and 2) they can be normalized without artifacts by taking into account propagation and scattering physics, potentially offering a superior method than current normalization techniques. For this grant, measured system parameter were used to form calibrated images from a high frequency synthetic aperture sonar system. Systems with large bandwidth and wide beams cannot directly estimate the scattering cross section, since that quantity is defined using a plane wave of a single frequency for the incident field. Correspondingly, the quantities produced by calibrated images must be understood as quantities averaged over the systems bandwidth and beamwidth. Details of this method, and scattering cross section results are presented in the enclosed OCEANS2019		

Analysis of Backscatter Measurements from Calibrated Synthetic Aperture Sonar Images

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Abstract— Calibrated Synthetic Aperture Sonar (SAS) data can provide monostatic backscattering strength measurements as a function of grazing angle at high resolution. Motivated by an interest in obtaining invariant metrics from vehicle based sonar acquisition systems, an end-to-end processing methodology for recovering calibrated scattering cross section levels from seafloor swaths is presented. Calibrated SAS beamforming software developed for this work was based on the narrowband sonar equation and a delay-and-sum beamformer. Modern SAS processing techniques such as redundant phase center micronavigation, and an autofocus algorithm to estimate sound speed, were employed in the beamforming stage. Data from overlapping sections of seafloor acquired in the Gulf of Mexico during 2012 provide the basis for analysis of comparative scattering results, which are examined in an effort to demonstrate adherence to established acoustic principles.

Keywords— *Calibrated Synthetic Aperture Sonar (SAS), acoustic backscattering strength, grazing angle, absolute intensity*

I. INTRODUCTION

High-frequency sonar imagery is commonly used for target detection and mapping purposes [1]. The dynamic range of these images is quite high and the image power depends on range due to terms in the sonar equation, such as spherical spreading, sensor directivity, attenuation, and processing gain [2]. Image analysis by human operators or machines requires a certain dynamic range to achieve good performance, so these images are typically normalized using time-varying gain (TVG). A TVG can either be based on the sonar equation [3], or empirical analysis. A simple and common empirical method of normalization is to estimate the received intensity across all channels, then divide each channel by the mean amplitude (square root of the mean power). The normalized channel data are then used to form the image using standard imaging algorithms [4], such as backprojection.

After beamforming, a secondary image pre-processing step is usually applied to the beamformed imagery to reduce the dynamic range of the pixel intensity to the appropriate bit depth of the human visual system (HVS). Typically, these approaches involved linear or non-linear compression of the tail of the low- and high-ends of the intensity distribution [5] or anchoring the scene intensity dynamic range to the statistics of the current scene [6]. These suppressions, especially when coupled with an adaptive TVG scheme as mentioned previously, distort the data and prevent estimation of quantitative acoustic metrics from the resulting image. Fig. 1 depicts an example beamformed image before Fig. 1(a) and after Fig. 1(b) applying the image normalization scheme from [6]. The seafloor and the faint target in the upper right hand corner of Fig. 1(a) are now clearly visible in Fig. 1(b) once the high-end of the dynamic range is compressed.

While these normalization schemes produce imagery pleasing to the human visual system, they distort absolute backscattering measurements from the seafloor and target. Here we seek to invert the sonar equation for synthetic aperture sonar (SAS) images to estimate backscattering strength. The scattering due to seafloor roughness and inhomogeneities is quantified by the scattering cross section per unit area per unit solid angle (hereafter called the “scattering cross section” or “cross section”). When converted to decibels, it is called scattering strength. This quantity is independent of measurement system and geometry, apart from the usual dependence of scattering strength on grazing angle, which in SAS images shows up as a range-dependent intensity. Inverting the sonar equation removes the large-scale intensity changes that are due to the measurement system, and reduces the dynamic range in a predictable way, as opposed to the empirical TVG methods described above.

Scattering strength is proportional to the mean intensity of the scattered field resulting from an interaction with a patch of seafloor [7]. The narrowband sonar equation typically used to describe scattered field behavior can be shown to apply to broadband SAS sonar systems found on unmanned underwater

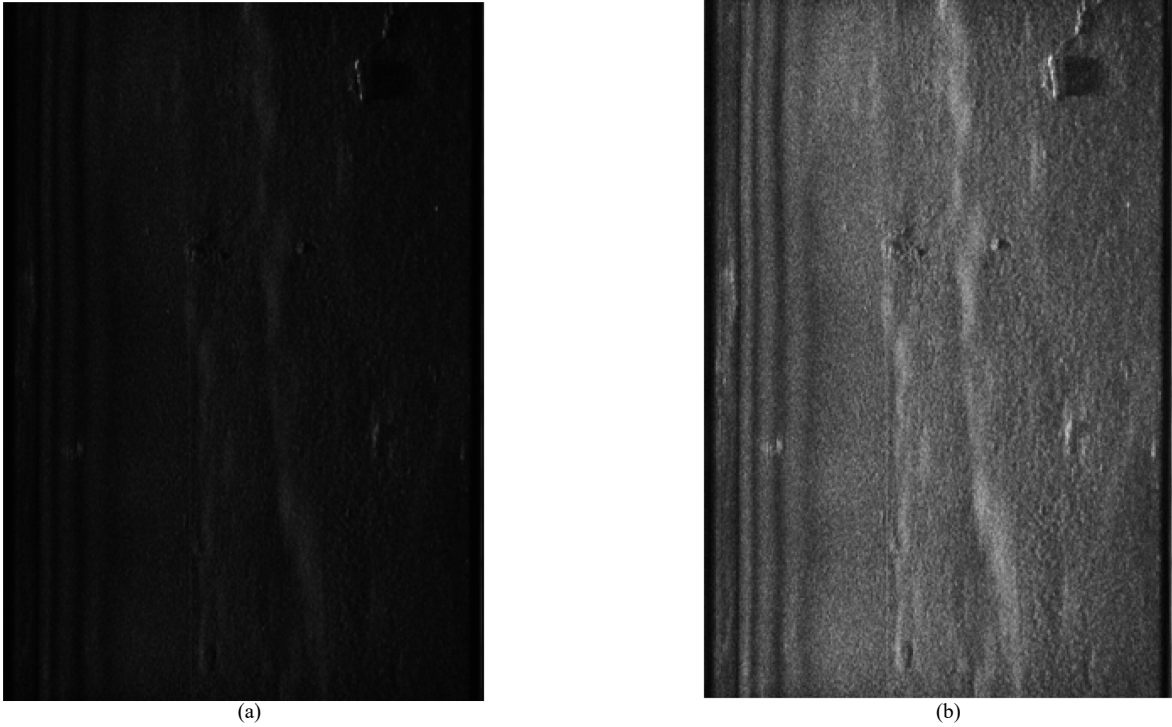


Figure 1. The beamformed image before (a) and after (b) applying the image normalization scheme from (Cook 2007). The seafloor and the faint target in the upper right hand corner of (a) are now clearly visible in (b) once the high-end of the dynamic range is compressed.

vehicles (UUV). This extension is shown for the two dimensional case as outlined by work completed by Olson and Lyons [8], which concluded that the mean intensity of the scattered field can be compared across systems having varying bandwidths, so long as the center frequency is held constant. An overview of the SAS beamforming process implemented to obtain absolute scattering strength levels from the data is presented, along with the methodology for processing subsequent imagery to calculate average scattering strength as a function of grazing angle. A description of the data set provides context for discussion of the scattering strength analysis. Results are shown that compare measurements of overlapping sections of seafloor and conclusions are drawn.

II. SCATTERING MEASUREMENTS

A. The Narrowband Sonar Equation

The scattered field measured by a sonar system depends on the imaging geometry, system parameters, and seafloor properties. When observed from a UUV platform, the transmitter and receivers are close to one another and can be considered to be a monostatic geometry. In Olson et al [9], a method to estimate the scattering cross section for each pixel in a SAS image was detailed. In that reference, the calibration method was based on system parameters that were estimated in several different ways. The element beam patterns were measured in the laboratory, but the overall source strength and receiver sensitivity was estimated by comparing data to a scattering model with known inputs. This method automatically took into account all multiplicative constants in the sonar equation, and therefore the method was able to be simplified. The method used here is based entirely on

laboratory measurements of the sonar parameters, and estimates of sonar location from on-board navigation instrumentation and micronavigation [10] and requires a more rigorous accounting of all the variables. We have therefore altered the calibrated processing, resulting in the following estimate of per-pixel scattering cross section:

$$\tilde{\sigma}_i = \frac{|q_i|^2}{\Gamma_i(\sum_j w_{ij})^2}, \quad (1)$$

$$|q_i|^2 = \sum_j^{N_r} a_{mf} \sqrt{\frac{2 \cos \theta_{ij}}{s_r s_o c \tau \cos(\theta_{ij} - \theta_o)} \frac{1}{r_{ij} \Psi_{SE}} \frac{r_{ij}^2 e^{2\alpha r_{ij}}}{|b_{tx}(\theta_{ij}) b_{rx}(\theta_{ij})|^2}} w_{ij} v_{ij}, \quad (2)$$

where $\tilde{\sigma}_i$ is the unaveraged scattering cross section of the i^{th} pixel, r_{ij} is the range from the j^{th} sensor to the i^{th} pixel, v_{ij} is the received voltage from the j^{th} sensor delayed for the i^{th} pixel, s_r is the receiver voltage sensitivity in V/Pa, s_o is the source level in Pa \times m, α is the frequency dependent absorption coefficient in radians per meter, and c is the sound velocity. The transmitted pulse length is τ , and the factor a_{mf} takes into account the gain if a matched filter is used to compress the pulse. $b_{tx}(\theta_{ij})$ is the transmitter vertical directivity pattern, and $b_{rx}(\theta_{ij})$ is that of the receiving transducer. The output of the initial stages of the beamforming process is represented by the variable q_i , for pixel i . In the beamforming stage, Eq. (2), the sonar equation has been inverted for each channel before beamforming. Eq. (1) accounts for the change in energy due to beamforming. In Eq. (2), j represents an individual sensor

channel, and the subscript ij represents variables that depend on both pixel location and sensor location. The reference angle, θ_0 , represents the depression angle of the main response axis and θ_{ij} the grazing angle from the j^{th} sensor channel to the i^{th} pixel. The factor Ψ_{SE} represents the angular beamwidth of a single array element, so that $r_{ij} \Psi_{SE}$ is the azimuthal dimension of the ensonified area. The parameter Γ_i takes into account both the partially coherent gain of the beamforming stage, as well as a conversion of the azimuthal resolution between a single sensor channel and a weighted linear array. The weights applied to the synthetic aperture for each pixel are specified by w_{ij} , and the coherent gain is removed from the beamformed image.

B. Broadband Effects

In the above analysis, the narrowband sonar equation was employed, although broadband frequency modulated pulses are commonly used in SAS systems. It is key to understand any effect of using a broadband signal on scattering strength estimates. We report result from Olson and Lyons [10], in which the Helmholtz-Kirchhoff integral equation was solved for a rough surface over a large frequency band. Time-domain signals were formed through Fourier synthesis, and Monte-Carlo estimates of scattering strengths were formed for different spatial resolutions. In Fig. 2, we plot scattering strength for a pressure-release rough surface following a one dimensional power law with a spectral strength of $\llbracket 10 \rrbracket^{-5}$ m, and a spectral exponent of 2. The simulations used a center frequency of 100 kHz (the center frequency of the system used in Olson et al [8], which motivated [10]) and various bandwidths. From this figure, no discernable dependence on bandwidth is observed. This independence of scattering strength on pulse length for power law roughness spectra indicates that the narrowband sonar equation is appropriate. Although this simulation was performed for a one-dimensional roughness spectrum, we expect that the same relationship holds for isotropic two-dimensional roughness. Measurements in anisotropic environments may not have the same relationship, which we will note later in our analysis.

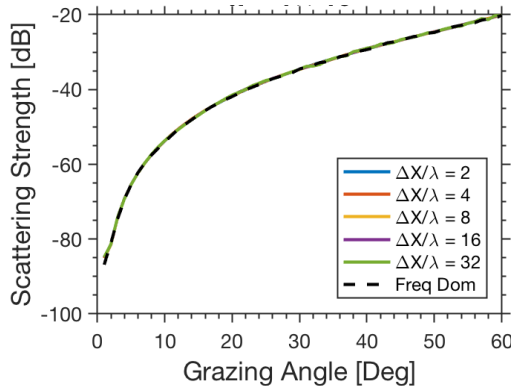


Figure 2. Result of numerical simulation of the time-domain scattering cross section estimated for different pulse spatial resolution, ΔX compared to the acoustic wavelength, λ . Results for a pure frequency domain simulation at the center frequency were also performed.

III. DATA SETS AND RESULTS

A. Overview

The narrowband sonar equation was used to estimate $\tilde{\sigma}$ for data collected off the coast of Panama City, FL in 2012. This data set is of particular importance for our purposes because it was collected with many overlapping tracks, allowing for independent measurements over the same seafloor swaths. This collection geometry allows the comparison of scattering strength for the same patch of seafloor measured from disparate azimuthal angles. The data collection comprised lawnmower tracks in 18.8 – 19.8 m of water depth in an operation area understood to be predominantly sand. This minimal change in depth over the experimental area enables the assumption of small large-scale seafloor slope in the image. The runs processed for this analysis were of undisturbed seafloor sections and did not contain any intentionally positioned targets or objects of interest.

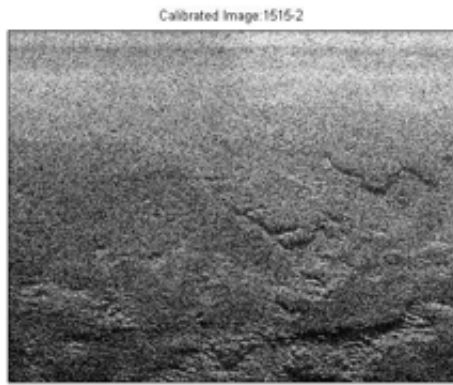
Several examples of calibrated images are presented in Figs. 3-5 and display backscatter strength in decibels. In Fig. 3 (a) and (b), the two images depict a similar section of seafloor having approximately 80% of overlap for the full swath. Distinguishing features of this bottom section are discernable in each image. These images were acquired from the same SAS array at vehicle headings that differed by 5°. Individual pixels from these data files were averaged in the across track (range) dimension and analyzed as a function of the corresponding grazing angle, then plotted as shown in Fig. 3 (c). These results correspond to the images (a) and (b) and show close agreement in averaged scattering levels having a mean difference of approximately 0.08 dB across applied grazing angles. Also shown in dotted line is the empirical Lambert curve, defined for the monostatic condition in equation 3.

$$\sigma = \mu \sin^2 \theta_i \quad (3)$$

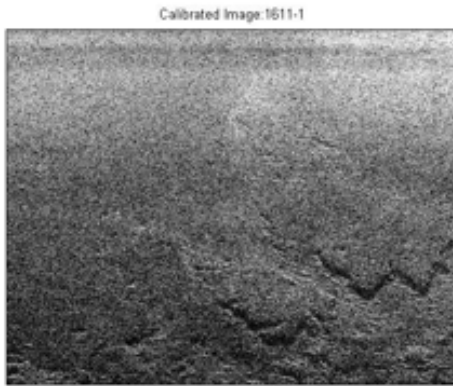
where μ is an empirically determined constant equal to -20 dB for this scene. This value is consistent with other values of the Lambert parameter for sandy seafloors summarized in Fig 12.3 of Jackson and Richardson [7]. Scattering strength values presented were calculated based on pixel sections depicted in the corresponding images..

B. Analysis

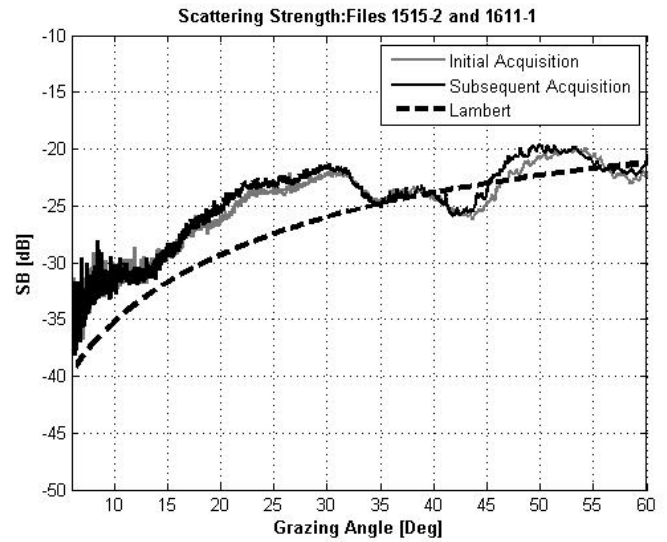
The approach outlined above was applied to the larger set of data to determine whether calibrated scattering strength is consistent for different collection geometries in the 2012 test data. This initial analysis of the calibrated processing focused on establishing repeatability of measurements taken separately but over similar sections of seafloor. The study also serves to establish a baseline from which future work can be analyzed and seeks to determine when scattering strength measurements from calibrated SAS images can be appropriately applied.



(a)



(b)



(c)

Figure 3. The calibrated image obtained during initial acquisition (a) and (b) the calibrated image processed during subsequent vehicle run. Scattering strength results from both compared to Lambert Curve (c).

Data were examined from 110 port-side images. Of these data files there were 42 instances where the images comprised seafloor sections exhibiting a nominal 60% or more of overlap. Mean difference of scattering strength in dB was compared for a relevant range of grazing angles. This was done for images acquired from similar vehicle heading, defined as within $\pm 10^\circ$. Analysis was also performed for images with opposing heading, defined as vehicle heading difference $> 170^\circ$.

Complete image files were trimmed in both the across track and along track axis to discard potential edge effects. Scattering strength values for each file were then averaged in the along track dimension. Grazing angles were calculated from vehicle position and image coordinates. Overall, the scattering strengths calculated via calibrated processing described above showed a 0.82 dB difference when repeat pass data was acquired under similar reflection pose and a 1.32 dB difference when acquired at opposite reflection pose, or azimuth angle. These differences represent the mean scattering strength at system relevant grazing angles and are shown in Table 1:

TABLE 1.

	Sample Size	Mean Absolute SB Difference (dB)
Same Vehicle Heading (Difference $< 10^\circ$)	24	0.82
Opposite Vehicle Heading (Difference $> 170^\circ$)	18	1.32

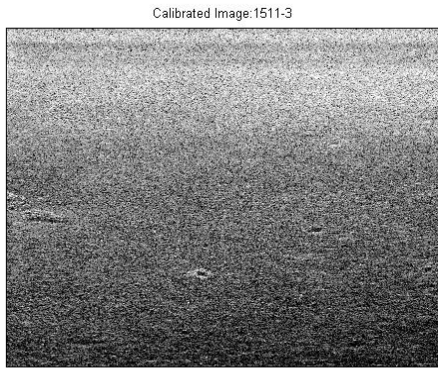
Additionally for all pose scenarios the less overlap that existed between swaths, the greater the difference in scattering measurements i.e. images having less similar sections had less similar scattering strengths, a result depicted in Table 2:

TABLE 2.

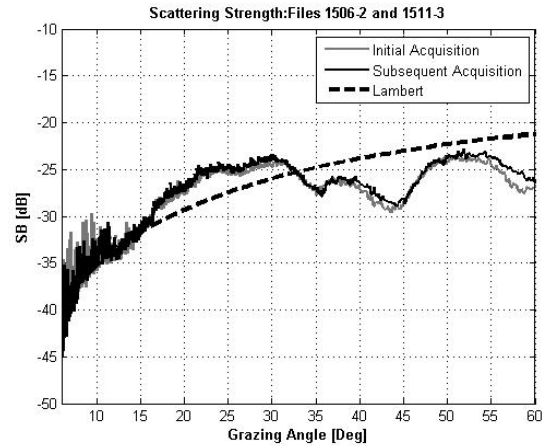
Nominal % Overlap	Mean Scattering Difference (dB)
60-69	1.21
70-79	0.91
80-89	0.63

In instances where the truncated image appeared homogeneous, calibrated scattering strengths were similar even if acquired at different azimuth angles. Fig. 4 depicts the robustness in scattering strength measurements encountered when acquisition geometries varied according to a heading difference of 180° . This result supports the acoustic understanding that a diffusely scattered field lacks directionality.

Swaths where the seafloor was discernably anisotropic (primarily directional ripples) showed large scattering strength differences. An example is shown in Fig. 5. The pockmarked bottom and rippling structure presented an interface that gave rise to an incident field highly dependent on vehicle acquisition geometry. Since the entire image was used to estimate the average scattering strength, systematic changes, such as the relative pixel fraction of the image composed of ripples,

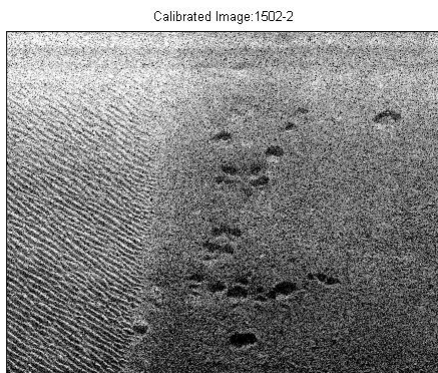


(b)

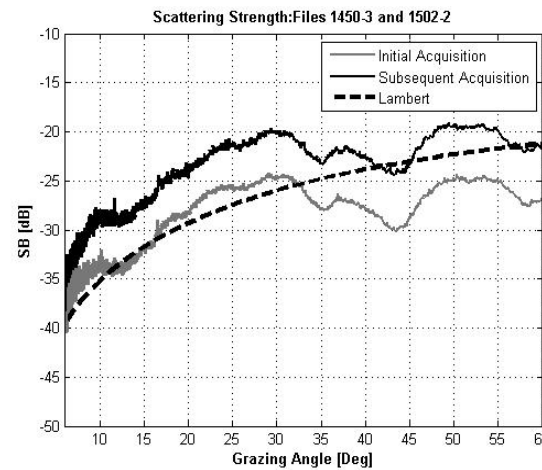


(b)

Figure 4. The calibrated image (a) and (b) scattering strength results from both compared to Lambert Curve.



(a)



(b)

Figure 5. The calibrated image (a) and (b) scattering strength results from both compared to Lambert Curve.

pockmarks, and homogeneous sand may be responsible for the discrepancy in Fig. 5(b). Analysis using smaller image patches may reveal more consistency across imaging geometry. This is an opportunity for future research.

IV. CONCLUSION

A calibrated beamformer was developed and utilized to generate scattering strength measurements for individual pixels contained within SAS data files. The narrow band sonar equation formed the basis for this process and was shown to be extendable to broadband sonar systems through numerical simulation and the processing of sonar data from a 2012 sea test. The acquisition of multiple overlapping sections of seafloor provided an opportunity to assess the calibrated process for conformity to established acoustic principles. Analysis demonstrated that absolute scattering strengths approximated documented results for like bottom types, and

review of multiple images indicated that similar seafloor sections (as defined by approximate percentage of overlap) have similar scattering strengths. Further, measured scattering strength differences are inversely proportional to percentage overlap. Anisotropic surface types presented highly variable scattering strengths when changes to vehicle geometry were introduced, whereas homogenous surface types were robust to changes in measurement when analyzed from various azimuth angles.

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Resolution dependence of rough surface scattering using a power law roughness spectrum

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1 Contemporary high-resolution sonar systems use broadband pulses and long real or
2 synthetic arrays to achieve high resolution for mapping and target detection tasks.
3 The spatial resolution of these systems can approach the acoustic wavelength, and
4 it is important to understand any affects this situation might have on quantitative
5 measures of the scattered field, such as the scattering cross section or scintillation
6 index. In this work, we numerically investigate the dependence of these two acous-
7 tic measures on pulse length (or equivalently bandwidth) using rough surfaces with
8 power-law spectra. Using the boundary element method and Fourier synthesis, we
9 found that there is no resolution dependence of the scattering cross section. We found
10 that the scintillation index increases as resolution increases, grazing angle decreases,
11 and spectral strength increases. This trend is confirmed for center frequencies of 100
12 kHz and 10 kHz, as well as for power law spectral exponents of 1.5, 2, and 2.5. The
13 hypothesis that local tilting at the scale of the acoustic resolution is responsible for
14 intensity fluctuations was examined. It was found that local tilting is responsible in
15 part for the fluctuations, but other effects, such as non-local multiple scattering and
16 shadowing likely also play a role.

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17 **I. INTRODUCTION**

18 Theoretical treatment of wave scattering from rough interfaces is generally performed
 19 using an incident plane wave, which by definition has a single direction and frequency, and
 20 exists over infinite spatial extent. However, experimental measurements of the scattered
 21 field typically employ broadband pulses to achieve high spatial resolution - desirable for
 22 seafloor mapping or target detection. Performance of such systems typically depends on
 23 the mean scattered intensity of the scattered field from the seafloor, and more generally its
 24 probability density function. The mean intensity is usually characterized in terms of the
 25 scattering cross σ section per unit area per unit solid angle, σ (hereafter referred to as the
 26 “cross section”, “scattering cross section,” or “scattering strength” for the decibel version.).
 27 Variability in the scattered intensity is often characterized using the scintillation index, SI
 28 (Ishimaru, 1978; Lyons *et al.*, 2009; Tatarski, 1961).

29 The scattering cross section is defined as the ratio of the scattered intensity to the incident
 30 intensity, normalized by propagation effects, as well as the energy flux of the incident acoustic
 31 wave. For some finite resolution systems at low grazing angles, this definition may be
 32 simplified to (Jackson and Richardson, 2007, p. 32),

$$\sigma = \frac{A \langle |p_s|^2 \rangle}{r_s^2 |p_i|^2} \quad (1)$$

33 where A is the ensonified area, r_s is the distance between a resolved patch of area A on the
 34 seafloor (or any rough surface, generally), p_i is the incident pressure, and p_s is the scattered
 35 pressure measured by the system. Note that this definition is valid only for geometries with
 36 well-defined incident and scattered field directions. Strictly, this definition of the scattering

37 cross section is only true in the limit as the ensonified area becomes large compared to all
38 length scales of interest (i.e. outer scale of the rough surface, or the acoustic wavelength),
39 since plane waves interact with the entire rough surface. It has been demonstrated ([Gauss](#)
40 [et al., 1996](#); [Henye](#)[y et al., 1995](#)), that for finite resolution systems, the scattering strength is
41 independent of pulse length - a consequence of Parseval's theorem ([Oppenheim et al., 1999](#),
42 p. 60) and linear time-invariance, and thus generally valid. However, if the properties of
43 the ensemble of rough surfaces vary with resolution, the scattering environment has very
44 strong frequency dependence, or is time-varying, then the measured scattering strength may
45 depend on resolution. For high-resolution systems, it is plausible that the ensemble used
46 to estimate the scattering cross section may change as a function of resolution, and has
47 potential to cause a resolution dependence of the scattering cross section.

48 The interface scattering cross-section characterizes the mean scattered power from an
49 interface, but a more general property of the scattered field is the probability density function
50 (pdf) of the pressure or its complex magnitude, termed the envelope pdf. The envelope pdf is
51 connected to performance of target detection systems, and has potential for remote sensing of
52 the environment using high resolution systems ([Lyons et al., 2009](#); [2016](#); [Olson et al., 2019](#)).
53 The pressure due to scattering from a rough, homogeneous interface with Gaussian height
54 statistics has commonly been assumed to follow a Gaussian distribution for the real and
55 imaginary components, and a Rayleigh distribution for its complex magnitude ([Jakeman](#),
56 [1980](#))). In this situation, the scintillation index, or normalized intensity variance is unity.
57 For heavy-tailed statistics (with more frequent large amplitude events), the scintillation
58 index is greater than unity.

59 Arguments for Rayleigh distributed scattered pressure magnitude follow from the assump-
60 tion of a large number of independent surface elements contributing to the scattered field
61 (Abraham and Lyons, 2002; Jakeman, 1980). So long as the ensonified area of a Gaussian,
62 homogeneous rough interface is large (so there are many independent scatterers contributing
63 to the field), this assumption holds true. Another argument for Rayleigh magnitudes follows
64 from perturbation theory and the interpretation in terms of Bragg scattering. In this frame-
65 work, the scattered pressure is proportional to the amplitude spectrum of the roughness
66 evaluated at the Bragg wavenumber, $2k_w \cos(\theta_i)$, where k_w is the acoustic wavenumber in
67 the water column, and θ_i is the incident grazing angle. If the surface has Gaussian statistics
68 in the spatial domain, then the wavenumber components will have a Rayleigh distributed
69 envelope via the central limit theorem. Since the the acoustic pressure magnitude is directly
70 proportional to the acoustic spectrum at the Bragg wavenumber, it follows that the envelope
71 pdf will be Rayleigh distributed as well.

72 Contemporary high-resolution seafloor imaging systems, such as synthetic aperture sonar
73 (SAS) have spatial resolutions on the order of the center wavelength. Small resolution cell
74 sizes may result in ensembles that vary with the resolved area of the seafloor, thereby causing
75 a departure from Rayleigh statistics. The resolution dependence of SI has implications for
76 target detection performance, synthetic aperture autofocus algorithms (e.g. (Marston and
77 Plotnick, 2015)), and preprocessing algorithms for SAS images (Williams, 2015).

78 It was observed in (Lyons *et al.*, 2016) that measurements of the scintillation index from
79 SAS images of homogeneous random rough interfaces had a strong dependence on range,
80 which was interpreted as a result of modulation of the local slope by roughness components

81 at the scale of the acoustic resolution or larger. This interpretation corresponds to the
 82 composite roughness model (Jackson *et al.*, 1986b; McDaniel and Gorman, 1983). Combined
 83 with interpretations in (Lyons *et al.*, 2016), this results in a dependence of the scintillation
 84 index on the acoustic resolution, the underlying pixel statistics, range (through grazing
 85 angle), as well as roughness spectrum parameters. These interpretations, while plausible,
 86 suffer from a lack of experimental confirmation.

87 In this paper, we examine the question of whether there is a dependence on resolution of
 88 the scattering cross section and scintillation index. Since the cross section is strictly defined
 89 for incident and scattered plane waves, a finite resolution version must be used, which is
 90 detailed below. When discussing the cross section, it is always noted whether we mean the
 91 plane wave version or finite resolution approximation. The acoustic resolution is defined as
 92 the full width half maximum spatial extent of the square of the incident pulse envelope, and
 93 is equal to $\Delta X = c/(2aB_w)$, where ΔX is the spatial resolution of the pulse¹, $1/(2aB_w)$ is
 94 the temporal resolution of the pulse in the backscattering direction, B_w is the 3dB full width
 95 bandwidth of the transmitted pulse, and c is the wave speed. The constant a depends on
 96 the shape, or point spread function of the pulse used.

97 These questions were investigated through numerical solution of the Helmholtz-Kirchhoff
 98 integral for the scattered pressure using the boundary element method (BEM) (Sauter and
 99 Schwab, 2011; Wu, 2000) using pressure-release boundary conditions. This method is similar
 100 to that used by (Thorsos, 1988). Fourier synthesis was used to construct the broadband
 101 scattered pressure at various spatial resolutions, and metrics were computed based on the
 102 scattered time-domain pressure. Comparisons were made to the ensemble averaged cross

103 section performed in the frequency domain (i.e. computed at a single frequency, which is a
 104 good approximation of the plane wave case). These simulations were performed for center
 105 frequencies of both 100 kHz and 10 kHz, and for one dimensional rough surfaces with power
 106 law spectra, whose parameters are the spectral strength and spectral exponent.

107 Through these numerical experiments, it was found that the scattering strength does not
 108 vary as a function of bandwidth for the parameters investigated in this study. The error
 109 of this comparison is within the Monte Carlo error of this study. For scintillation index we
 110 found that it becomes greater than one as resolution increases, grazing angle decreases, and
 111 spectral strength increases. Specifically, for larger spectral exponents, the scintillation index
 112 is more sensitive to changes in spectral strength, resolution and grazing angle.

113 We first present an overview of the geometry and roughness statistics in Sec. II. The
 114 integral equations and discretization methods are given in Sec. III, and the incident field in
 115 Sec. IV. Methods to estimate the scattering cross section and scintillation index are given
 116 in Sec. V. We give a discussion on how the parameters of the numerical simulations were
 117 selected in Sec. VI. Results are presented in Sec. VII, with a discussion and some preliminary
 118 hypotheses given in Sec. VIII. Conclusions are given in Sec. IX

119 II. GEOMETRY AND ENVIRONMENT

120 The geometry of the scattering problem is presented in Fig. 1. The problem takes place
 121 in two dimensions with position vector $\mathbf{r} = (x, z)$. The rough interface is defined as $z = f(x)$
 122 and is shown as the thick black line in this figure. In this figure, the nominal incident and
 123 scattered wave directions are shown with their grazing angles and nominal wave vectors.

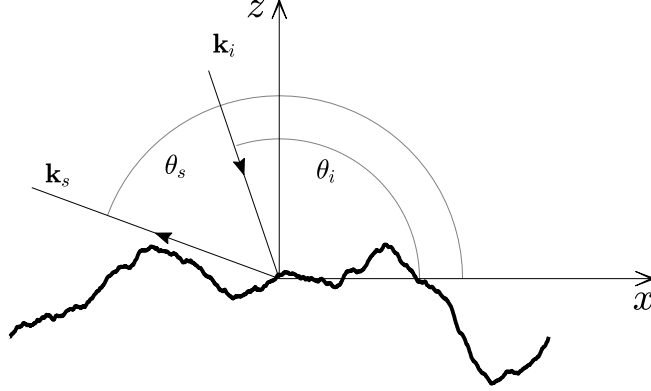


FIG. 1. caption

124 The sound speed in the upper medium is c , which is taken to be 1500 m/s, but our results
 125 can be applied to other sound velocities by performing the appropriate dimensional scaling.
 126 The acoustic frequency is f , and is related to the wavenumber $k = 2\pi f/c$. Simulations
 127 are performed at a center frequency f_0 , and bandwidth BW . The center wavelength and
 128 wavenumber are λ_0 and k_0 respectively

129 The rough interface is assumed to have wide-sense homogeneity (spatial stationarity) and
 130 a Gaussian pdf. Its second order properties can be completely described by its autocovari-
 131 ance function,

$$B(x) = \langle f(y)f(y+x) \rangle \quad (2)$$

132 and power density spectrum

$$W(K) = \frac{1}{2\pi} \int B(x)e^{iKx} dx. \quad (3)$$

133 Several second-order properties of this spectrum are useful for the analysis performed in this
 134 paper. In particular, the root mean square (rms) height, h^2 is given by

$$h^2 = \int_{-\infty}^{\infty} W(K) dK = B(0). \quad (4)$$

135 The rms slope s^2 is

$$s^2 = \int_{-\infty}^{\infty} K^2 W(K) dK = \left. \frac{\partial^2 B(x)}{\partial x^2} \right|_{x=0}. \quad (5)$$

136 The power density spectrum used in this work is the truncated power law,

$$W = \frac{w}{|K|^\gamma}, \quad (6)$$

137 for $k_l \leq |K| \leq k_u$, and zero otherwise. The spectral strength is w with units of $m^{3-\gamma}$
 138 and γ is the dimensionless spectral exponent. The lower wavenumber cutoff is $k_l = 2\pi/L$,
 139 where L is the outer scale. The upper wavenumber cutoff is $k_u = 2\pi/\ell$, where ℓ is the
 140 inner scale. Random realizations are produced from this power spectrum using the Fourier
 141 synthesis technique detailed in (Thorsos, 1988). Since $f(x)$ is a real function, both positive
 142 and negative wavenumbers are included in the Fourier integral. We identify the outer scale
 143 with the length of the rough surface, and the inner scale with the sampling interval of the
 144 rough surface realization.

145 For the power-law form used here, the non-dimensional mean square slope and mean
 146 square height are

$$s^2 = \frac{2k_0^{3-\gamma} w}{3-\gamma} \left[\left(\frac{k_u}{k_0} \right)^{3-\gamma} - \left(\frac{k_l}{k_0} \right)^{3-\gamma} \right] \quad (7)$$

$$k_0^2 h^2 = \frac{2k_0^{3-\gamma} w}{\gamma-1} \left(\frac{k_0^{\gamma-1}}{k_l^{\gamma-1}} - \frac{k_0^{\gamma-1}}{k_u^{\gamma-1}} \right). \quad (8)$$

147 These parameters have been expressed in a form where the terms outside and inside the
 148 parentheses are dimensionless. Although the upper wavenumber is set by the discretization
 149 interval, δx , the way in which the rough surfaces enter into the acoustical simulations may
 150 be subject to an effective upper limit, $k_{ueff} = 2\pi/\ell_{eff}$, where ℓ_{eff} is an effective inner scale.
 151 Roughness components with wavenumbers much greater than k likely have an insignificant
 152 effect on the scattered field, making the effective upper limit much less than that defined
 153 by the surface sampling. RMS height is insensitive to the upper cutoff, and more sensitive
 154 to the low-wavenumber cutoff. RMS slope is sensitive to the upper cutoff, and insensitive
 155 to the lower cutoff, so long as it is sufficiently small. To make the upper limit explicit, we
 156 will use the notation s_ℓ to denote the rms slope computed using $k_u = 2\pi/\ell$ for some length
 157 scale ℓ .

158 III. INTEGRAL EQUATIONS AND DISCRETIZATION

159 We perform this study numerically using a discretized form of the 2D Helmholtz-Kirchhoff
 160 integral equation for Dirichlet boundary conditions (Thorsos, 1988). Although our motiva-
 161 tion for this work is seafloor scattering, the assumption of a Dirichlet boundary allows us to
 162 focus solely on the role of the rough interface. For a single frequency, this integral equation
 163 is

$$p(\mathbf{r}_p) C_p = p_i(\mathbf{r}_p) + \int_S \frac{\partial p(\mathbf{r}_s)}{\partial n_s} G_k(|\mathbf{r}_s - \mathbf{r}_p|) dS, \quad (9)$$

164 where C_p is 1/2 on the rough interface, unity in the fluid medium above the rough interface,
 165 and zero below the interface. p_i is the incident pressure, p is the total pressure, and $\partial p/\partial n$

166 is the total pressure normal derivative. $\mathbf{r}_p = (x_p, z_p)$ is a point in space, $\mathbf{r}_s = (x_s, z_s)$ is a
 167 point on the rough surface, and $G_k(R) = (i/4)H_0^{(2)}(kR)$ is the 2D free-space Green function
 168 (Devaney, 2012, p. 6), where $H_0^{(2)}(z)$ is the zeroth-order Hankel function of the second
 169 kind. Note that this integral equation can also describe electromagnetic scattering from 1D
 170 corrugated surface with perfectly conducting boundary conditions subject to an incident
 171 wave with TM (p) polarization (Toporkov *et al.*, 1998).

172 The scattering problem is solved in two steps. First, the point \mathbf{r}_p is taken to the boundary.
 173 Application of the boundary conditions results in the equation

$$-\int_S \frac{\partial p(\mathbf{r}_s)}{\partial n_s} G_k(|\mathbf{r}_s - \mathbf{r}_p|) dS = p_i(\mathbf{r}_p) \quad (10)$$

174 This equation is numerically solved for $\partial p/\partial n$ on the surface. Once this quantity is known,
 175 Eq. (9) is then evaluated with \mathbf{r}_p in the far field.

176 Numerical solution of Eq. (10) is performed by discretization of the integral equation
 177 using boundary element method (Sauter and Schwab, 2011; Wu, 2000). In particular, we
 178 use piecewise linear basis functions to approximate $\partial p/\partial n$, and collocation to compare the
 179 true and approximate solution at discrete points. These two methods convert the integral
 180 equation into a linear system,

$$Vy = b, \quad (11)$$

181 where y is the solution vector consisting of the basis function coefficients used in the approx-
 182 imation for $\partial p/\partial n$, and $b = p_i$ evaluated at the discrete collocation points $\mathbf{r}_m = (x_m, z_m)$.

183 The matrix V has elements

$$V_{mn} = -\int G_k(|\mathbf{r}_m - \mathbf{r}_s|) \phi(\xi_n(\mathbf{r}_s)) dS \quad (12)$$

184 Here, $\phi(\eta)$ is a linear basis function defined on the interval $\eta \in [-1, 1]$ - a symmetric
 185 triangular function centered at zero with a maximum of 1. Outside of the interval, ϕ is zero.
 186 The function ξ_n maps the basis function centered at the n -th point from physical space, \mathbf{r}_s
 187 to the η domain. In this case, the basis functions are centered at the same collocation points
 188 \mathbf{r}_m , resulting in a square matrix. Integration is carried out using Gauss-Legendre quadrature
 189 (Abramowitz and Stegun, 1972). Due to the weak singularity in the Green's function, the
 190 diagonal elements of the matrix are computed using a sixteen point quadrature rule combined
 191 with a variable transformation whose Jacobian exactly cancels the singularity (Wu, 2000).
 192 Nonsingular matrix elements were computed using an eight point quadrature rule. LAPACK
 193 routines were used to solve the linear system using LU decomposition and back substitution
 194 (Anderson *et al.*, 1999).

195 Collocation points are defined on the rough surface, (x_m, z_m) with equal spacing, δx on
 196 the horizontal axis. From these points, a cubic spline approximation is used to construct
 197 a continuous and smooth surface. This interpolation forces the surface normal, and thus
 198 $\partial p / \partial n$ to be continuous, which improves the convergence rate of the discretization of the
 199 integral operator (Atkinson, 1997).

200 Once the surface pressure normal derivative is found, it is propagated to the field using
 201 Eq. (9) with \mathbf{r}_p in the far field. In this work, the field points are equally spaced intervals of one
 202 degree at radius R . The far field radius is defined to be about 25 times the Rayleigh distance
 203 from the surface, d^2 / λ_0 . This criterion for the far-field is quite conservative (Jackson and
 204 Richardson, 2007, Appendix J), (Lysanov, 1973; Winebrenner and Ishimaru, 1986), although
 205 it enables the use of asymptotic expansions for the Hankel function.

206 **IV. INCIDENT FIELD**

207 The incident fields used in this work are broadband pulses whose spatial dependence is
 208 an approximation of a plane wave. The nominal direction of the incident wave vector are
 209 specified as vectors in Fig. 1. Their lengths vary due to the broadband nature of the field,
 210 although the center wave vectors can be defined at the center frequency by the expressions
 211 $\mathbf{k}_{0i} = (k_{0ix}, k_{0iz})$ and $\mathbf{k}_{0s} = (k_{0sx}, k_{0sz})$. The components are defined in terms of the grazing
 212 angles θ_i and θ_s (with respect to the horizontal axis) by

$$k_{0ix} = -k_0 \cos \theta_i \qquad k_{0sx} = k_0 \cos \theta_s \qquad (13)$$

$$k_{0iz} = -k_0 \sin \theta_i \qquad k_{0sz} = k_0 \sin \theta_s \qquad (14)$$

213 The center wavenumber k_0 is defined by an average of the wavenumber weighted by power
 214 spectrum of the transmitted source

$$k_0 = \int_{-\infty}^{\infty} k S^2(f) df, \qquad (15)$$

215 where $S(f) = \int s(t) \exp(-i2\pi ft) dt$ is the linear (amplitude) spectrum of the transmitted
 216 pulse, $s(t)$. The transmitted pulse used here was a complex exponential multiplied by a
 217 Gaussian envelope, with the form $s(t) = p_0 \exp(-t^2/\tau^2 + i\omega_0 t)$, where τ is a parameter of
 218 the pulse length, and ω_0 is the center angular frequency, related to the center frequency, f_0
 219 by $f_0 = \omega_0/(2\pi)$. The factor p_0 is the pressure at the center of the pulse envelope and is
 220 included to make the dimensions consistent, taken to be 1 Pa in this work.

221 The temporal resolution of the pulse, $\Delta\tau$ is defined by the duration of the pulse envelope
 222 between its half power points. For the Gaussian pulse used, this quantity can be obtained

223 by solving the equation $\exp(-(\Delta\tau/2)^2/\tau^2) = 1/\sqrt{2}$, resulting in $\Delta\tau = \tau\sqrt{2\ln 2}$. This
 224 definition of the temporal resolution results in a full width half power bandwidth of $BW\Delta\tau =$
 225 $2\ln(2)/\pi \approx 0.44$. For reference, if a rectangular function with full width of BW is used
 226 for $S(f)$, then $BW\Delta\tau \approx 0.88$. The same relationship is obtained if constant envelope
 227 pulse of length $\Delta\tau$ is used. Although the rectangular pulse has a larger time-bandwidth
 228 product, the Gaussian pulse has no appreciable sidelobes in the time-domain, but requires a
 229 computational bandwidth much larger than BW to approximate a true Gaussian function.

230 Broadband fields are synthesized from single frequency approximations of a plane wave.
 231 This narrowband field is the extended Gaussian beam developed in (Thorsos, 1988) that
 232 provides tapering to guard against edge effects entering into the scattering calculation. The
 233 form of this field (adapted to our time convention) is given by

$$p(\mathbf{r}) = \exp(-i\mathbf{k}_i \cdot \mathbf{r}(1 + w(\mathbf{r})) - (x - z \cot \theta_i)^2), \quad (16)$$

234 where

$$w(\mathbf{r}) = (kg \sin \theta_i)^{-2} [2(x - z \cot \theta_i)^2 / g^2 - 1], \quad (17)$$

235 and g is a width parameter of the incident field. For broadband simulations, Eq. (16) is
 236 used for each frequency. Since the Gaussian function has an infinite domain of support, it
 237 must be truncated to use in numerical simulations.

238 The function $w(\mathbf{r})$ improves the agreement between the numerical solution of the
 239 Helmholtz-Kirchhoff integral equation and its normal derivative. Discrepancies between
 240 these two solutions can result because the incident field satisfies the Helmholtz equation
 241 approximately to order $(kg \sin \theta_i)^2$ (Thorsos, 1988). Good agreement between the two solu-

242 tions was observed when $kg \sin \theta_i$ is large. Therefore, to analyze low-grazing angles, (which
 243 are important to contemporary synthetic aperture sonar systems, e.g. (Bellettini and Pinto,
 244 2009; Dillon, 2018; Fossum *et al.*, 2008; Pinto; Sternlicht *et al.*, 2016)), the parameter g must
 245 grow as θ_i approaches zero. This requirement can be thought of as enforcing the constraint
 246 that the angular width of the incident beam (full width half max),

$$\Delta\theta = \frac{2\sqrt{2\log(2)}}{kg \sin \theta_i} \quad (18)$$

247 should be small compared to θ_i . When the relative angular width, defined as $\Delta\theta/\theta_i$ is not
 248 small, the direction of the incident field is spread over a large range of angles, and is thus
 249 not well defined.

250 V. ESTIMATING TIME-DOMAIN QUANTITIES OF THE SCATTERED FIELD

251 The time-domain pressure is computed by

$$p(t, \theta_i, \theta_s) = \int_{-\infty}^{\infty} S(f)p(f, \theta_s, \theta_i)e^{i(2\pi f/c)R+i2\pi ft} df \quad (19)$$

252 where $p(f, \theta_s, \theta_i)$ is the scattered pressure measured at θ_s with an incident field having
 253 frequency f , and incident grazing angle θ_i . The first term in the exponential removes the
 254 time delay associated with propagation to the far field, so that the $p(t)$ can be mapped to
 255 the rough interface. The scattered grazing angle, θ_s is computed using the location at which
 256 the pressure is calculated in the far field, $\theta_s = \tan^{-1}(z_s/x_s)$. In practice this integral is
 257 computed using the fast Fourier transform (FFT).

258 An example realization of the scattered pressure the frequency domain is plotted in
 259 Fig. 2(a). The frequency domain pressure is plotted as the raw scattered pressure, and also

260 weighted by the amplitude spectrum. The time-domain pressure squared after weighting
 261 by $S(f)$ and an inverse Fourier transform, is plotted in Fig. 2(b). It contains fluctuations
 262 on the time-scale of the pulse length, as well as deterministic changes due to the incident
 263 beam used in the Helmholtz integral calculations. The deterministic component must be
 264 removed before estimating the scattering cross section. This can be performed by dividing
 265 the pressure magnitude squared by the energy flux density, and multiplying it by $2\rho c/\sin\theta_i$.
 266 This quantity is equal to

$$\begin{aligned}
 e_f(x) = \exp(-2x^2/g^2) \times \\
 \left(1 + \frac{2x \cot \theta}{ikg^2} + \frac{4x^2 \cot^2 \theta_i \csc^2 \theta_i}{g^2 k^2} + \right. \\
 \left. + (2x^2/g^2 - 1) \frac{\csc^2 \theta_i}{g^2 k^2} \right)
 \end{aligned} \tag{20}$$

267 If $kg \sin \theta_i \gg 1$, then this expression can be approximated with $\exp(-2x^2/g^2)$. However,
 268 the full version of Eq. (20) is used in all cases here. The position argument, x is the position
 269 corresponding to time t in the scattered pressure. It is a function of the incident grazing
 270 angle and is defined as $x(t, \theta_i) = -ct/(2 \cos \theta_i)$. The negative sign is a consequence of the
 271 definition of the angles in Fig. 1. This coordinate change is used later to map the scattered
 272 time series to locations on the rough surface.

273 To remove effects of the pulse length and cylindrical spreading, the squared magnitude
 274 is multiplied by r and divided by $c\Delta\tau/(2 \cos \theta_i)$. To summarize, the dimensionless scaled
 275 intensity is defined as

$$q(t)^2 = \frac{r|p(t)|^2}{e_f(x(t, \theta_i))} \frac{2 \cos(\theta_i)}{c\Delta\tau} \tag{21}$$

276 This quantity is plotted in Fig. 2 for size different pulse resolutions.

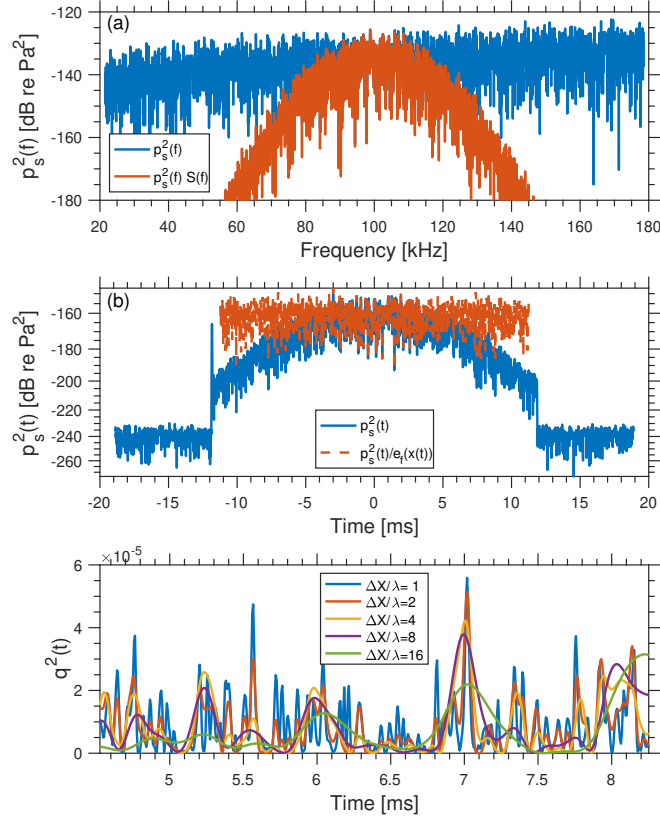


FIG. 2. (color online) Steps to estimate scattering cross section.

277 The scattering cross-section can be estimated by a simple average of q^2 ,

$$\sigma = \langle q^2 \rangle_{t, N_e}. \quad (22)$$

278 This ensemble average is over time, t as well as N_e , the number of rough surface realiza-
 279 tions. Averaging over time was performed after $q(t)$ was decimated to obtain statistically
 280 independent samples, following (Abraham and Lyons, 2004). Averaging over the roughness
 281 ensembles is performed to reduce uncertainty in the σ or SI estimate and is achieved by
 282 concatenating the time series obtained from each realization and averaging the resulting vec-
 283 tor. In practice, only 95% of the time series from the rough surface is used in the ensemble

284 average, to ensure that edge effects do not contaminate the time series (see the spike at -12
 285 ms in Fig. 2(b)). An example of $q^2(t)$ is plotted in Fig. 2(c) for different resolutions.

286 We also examine the scintillation index, SI , which is the variance of the scaled intensity,
 287 $I = q^2$ divided by the square of the mean intensity (Ishimaru, 1978, p. 437),

$$SI = \frac{\langle I^2 \rangle - \langle I \rangle^2}{\langle I \rangle^2}. \quad (23)$$

288 The scintillation index characterizes the fluctuations in the scattered field. If $SI = 1$, then
 289 the magnitude of the complex pressure (known as the envelope) has a Rayleigh distribution
 290 and its real and imaginary components are Gaussian. If $SI > 1$, then the pdf of the scattered
 291 field is heavy-tailed, which means that there is a higher probability of occurrence of high
 292 amplitude events compared to the Rayleigh distribution.

293 VI. PARAMETERS OF NUMERICAL EXPERIMENTS

294 A. Signal Parameters

295 The objective of this work is to study the resolution (or bandwidth) dependence of the
 296 scattered field. Our experiments covered the resolutions typically used in narrowband scat-
 297 tering experiments (Jackson *et al.*, 1986a; Williams *et al.*, 2002; Williams and Jackson, 1998),
 298 with the resolution cell on the order of 10 or more wavelengths, down to a value of one wave-
 299 length, which is on the order of what is achievable by modern SAS systems. We used specific
 300 values of $\Delta X/\lambda = (1, 2, 4, 8, 16)$. The proportional spatial resolutions correspond to tempo-
 301 ral resolutions $\Delta\tau f_0 = (2, 4, 8, 16, 32)$ at small grazing angles, since $\Delta X = \Delta\tau/(2 \cos \theta_i)$ for
 302 backscattering.

303 High frequency acoustic imaging systems provided the motivation for this work, and thus
 304 the simulations used a center frequency of 100 kHz. However, the parameters of the simu-
 305 lation were specified in a non-dimensional fashion. As long as every dimensional quantity
 306 is scaled properly, results of these simulations should be valid for lower frequencies with
 307 larger roughness parameters. To check whether the non-dimensional scaling was valid, we
 308 performed one of the simulations at 10 kHz as well, and scaled the roughness parameters
 309 accordingly. A sound speed of 1500 m/s was used for all simulations.

310 B. Roughness parameters

311 Roughness parameters were specified using only two dimensional constants, wavenumber,
 312 k (through a combination of f_0 and c) and spectral strength, w . Three spectral exponents
 313 were used, $\gamma = 1.5, 2, 2.5$, since this parameter has been observed to vary for measured
 314 seafloor roughness ([Jackson and Richardson, 2007](#), Ch 6). Only cases of $\gamma = 2$ were per-
 315 formed for both 100 kHz and 10 kHz. The $\gamma = 1.5$ and $\gamma = 2.5$ simulations used a center
 316 frequency of 100 kHz only.

317 It now remains to specify the spectral strength. For $\gamma = 2$, we used spectral strengths of
 318 $w_{\gamma=2} = (1 \times 10^{-6}, 1 \times 10^{-5}, 2 \times 10^{-5}, 3 \times 10^{-5}, 4 \times 10^{-5})\text{m}$, where we have used a subscript on
 319 the spectral strength to denote that it is used for a specific value of the spectral exponent.
 320 These values resulted in $SI \approx 1$ for the smallest w , and $SI > 1$ for larger values. These
 321 values span the smaller end of the roughness measurements with spectral exponent ≈ 2
 322 summarized in Table 6.1 of ([Jackson and Richardson, 2007](#)).

323 It was desired that the simulation parameters were set such that they resulted in similar
 324 values of SI for different spectral exponents. Based on interpretation that roughness at
 325 horizontal scales larger than the resolution cause changes in the local slope of the seafloor,
 326 we set the spectral strengths for $\gamma \neq 2$, such that the rms slope (Eq. (7)) was equal across
 327 different γ . Since pulses with a finite spatial resolution were used, it seems reasonable that
 328 an effective upper limit k_{ueff} is imposed on the rough surface based on the spatial resolution.
 329 With this requirement, w for other γ (denoted by w_γ), were computed using

$$w_\gamma = w_{\gamma=2}(3 - \gamma) \frac{k_{ueff} - k_l}{k_{ueff}^{3-\gamma} - k_l^{3-\gamma}} \quad (24)$$

330 where k_l is the lower limit of the wavenumber spectrum, and k_{ueff} is the effective upper
 331 wavenumber limit. We set $k_{ueff} = 0.5k_0$ based on the values of $\Delta X/\lambda$ studied here.

332 Roughness parameters for the 100 kHz simulations are summarized in Table I. For the
 333 10 kHz simulations, we chose to keep the same s^2 and kh . This condition can be satisfied
 334 if the spectral strength, surface length, and sampling interval for 100 kHz and $\gamma = 2$ are all
 335 multiplied by 10.

336 C. Sampling parameters

337 The sampling interval, δx was specified to minimize the discretization error of the integral
 338 equation. Since very large numbers of independent samples were used, our estimates of the
 339 scattering cross section had uncertainty of about 0.2-0.3 dB. It was found that an observable
 340 bias with this uncertainty in the scattering strength occurred if $\delta x/\lambda_0 > 12$, which is due to

γ	w	$s_{\delta x}^2$	$s_{\Delta X}^2$	$s_{2\Delta X}^2$	$s_{4\Delta X}^2$	$s_{8\Delta X}^2$	$s_{16\Delta X}^2$	$k_0 h$
-	$m^{3-\gamma}$	o	o	o	o	o	o	-
2	1.00e-06	6.40	1.66	1.17	0.83	0.58	0.41	0.95
	1.00e-05	19.52	5.23	3.70	2.62	1.85	1.30	2.99
	2.00e-05	26.63	7.37	5.22	3.70	2.61	1.84	4.23
	3.00e-05	31.55	9.00	6.39	4.52	3.20	2.25	5.18
	4.00e-05	35.34	10.37	7.37	5.22	3.69	2.60	5.98
1.5	1.02e-07	14.58	1.95	1.16	0.69	0.41	0.24	0.34
	1.02e-06	39.44	6.16	3.67	2.18	1.30	0.77	1.06
	2.04e-06	49.31	8.68	5.18	3.09	1.84	1.09	1.51
	3.06e-06	54.93	10.59	6.23	3.78	2.25	1.34	1.84

341 discretization error. A conservative value of $\delta x/\lambda_0 = 15$ was therefore used to minimize this
 342 bias.

343 Our focus in this work is moderate to low grazing angles. The lower limit to the grazing
 344 angles that can be reliably estimated is set by the surface length, which in turn is set by the
 345 memory limitations and acceptable number of CPU hours used. These latter constraints
 346 limited us to $g = 250\lambda$. At the center frequency, the relative angular width for that value of
 347 g was about 3.5% at 10° grazing angle. Lower frequencies are more of a problem, since for a
 348 constant value of g , decreasing the frequency will increase the angular width of the field. At
 349 the lower 6 dB down point of the largest bandwidth signal used (equivalent to $0.844f_0$), the
 350 angular width for $g = 250\lambda$ was approximately 4.2° at a nominal angle of 10° grazing angle.
 351 With these angular widths, 10° was taken to be an acceptable, if conservative, lower limit
 352 to the grazing angles that can be reliably estimated in this work. Lower angles result in a
 353 larger relative angular width which increases quadratically with θ_i . In choosing the surface
 354 length, we increased g to 400λ and did not observe any change in the behavior of σ or SI
 355 above 10° . Grazing angles less than 10° likely require fast approximate methods to solve the
 356 integral equation, such as the fast multipole method (Liu, 2009), or hierarchical matrices
 357 (Hackbusch, 2015).

358 The total rough surface length, L was set to $4.5g$ to allow the incident field taper to decay
 359 sufficiently at the edges of the computational domain. For the proportional bandwidths
 360 studied in this work, these surface parameters resulted in surfaces with $N = 16,875$ points.
 361 The matrix V resulting from this discretization required 4.3 GB of memory storage for
 362 double precision complex numbers.

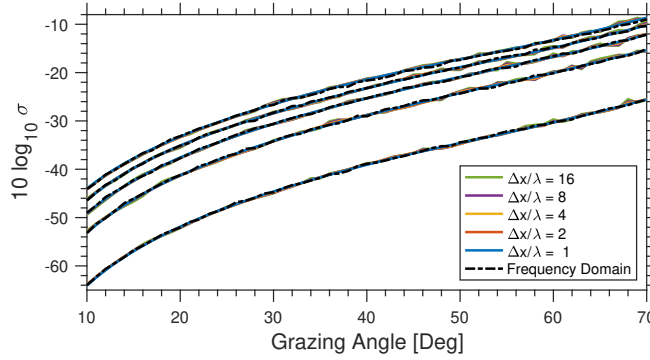


FIG. 3. Scattering Strength comparison. 100 kHz, $\gamma = 2$

363 The frequencies required for the largest bandwidth simulation $\Delta\tau f_0 = 2$, spanned approx-
 364 imately $0.2 f_0$ to $1.8 f_0$. This computational bandwidth was about seven times the largest
 365 3dB bandwidth used for the Gaussian spectrum. Using a frequency spacing of $\delta f = c/(3L)$,
 366 the number of frequencies per simulation was approximately 6000. Simulations were per-
 367 formed on the Hamming high performance computing cluster at the Naval Postgraduate
 368 School.

369 VII. RESULTS

370 We present results for backscattering strength as a function of grazing angle, θ_i , resolution
 371 ΔX , and spectral strength, w , in Fig. 3 for $f_0 = 100$ kHz and $\gamma = 2$. Finite resolution
 372 scattering strength as well as the frequency domain version is plotted on the vertical axis,
 373 with grazing angle on the horizontal axis. Each resolution is plotted as a different color,
 374 with the frequency domain version as a black dashed-dot line. Results for different spectral
 375 strengths are on the same figure since they are well separated from one another, with the
 376 smallest w corresponding to the lowest scattering strength.

Resolution Dependence of Scattered Field

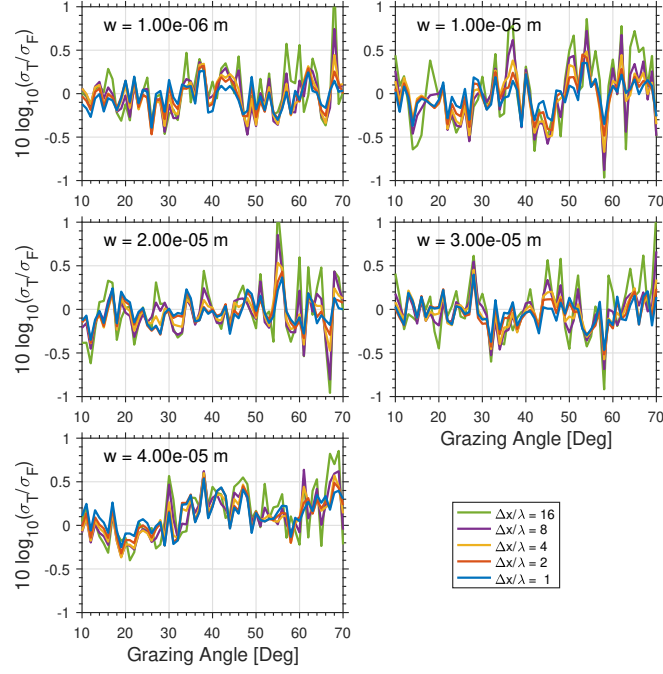


FIG. 4. Scattering Strength ratio. 100 kHz, $\gamma = 2$

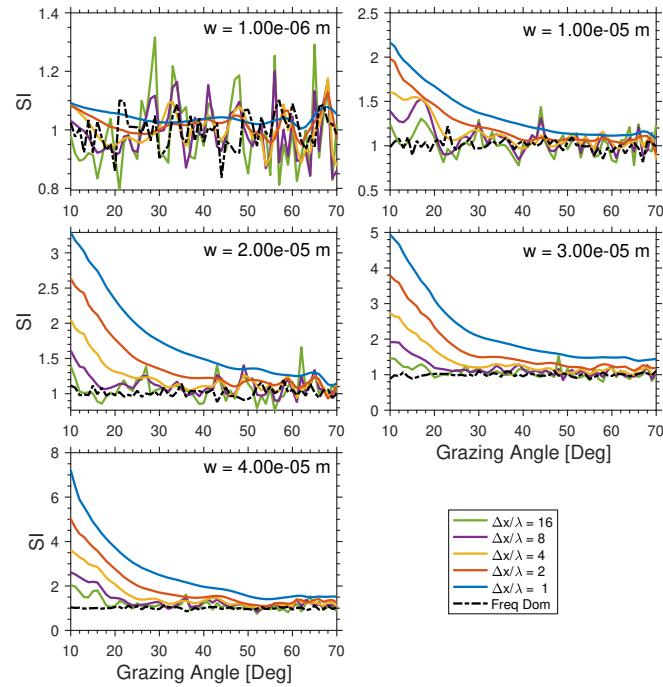


FIG. 5. Scintillation Index. 100 kHz $\gamma = 2$

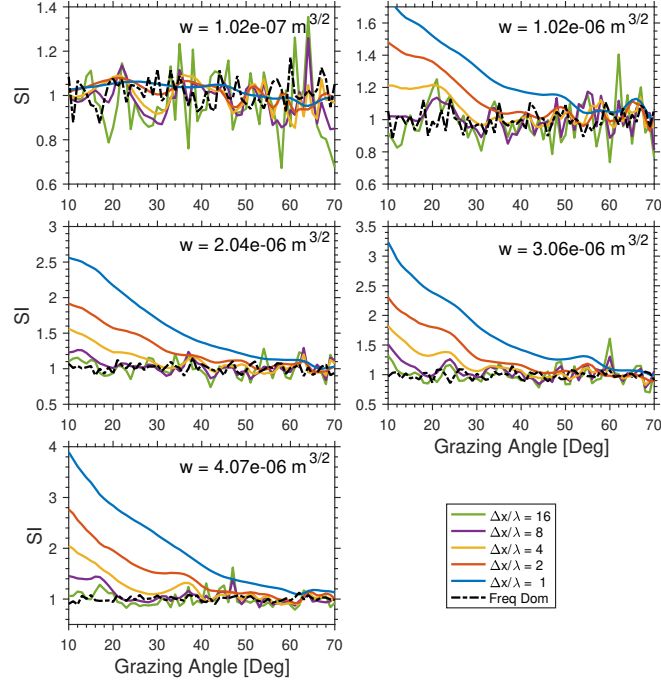


FIG. 6. Scintillation Index 100 kHz, $\gamma = 1.5$

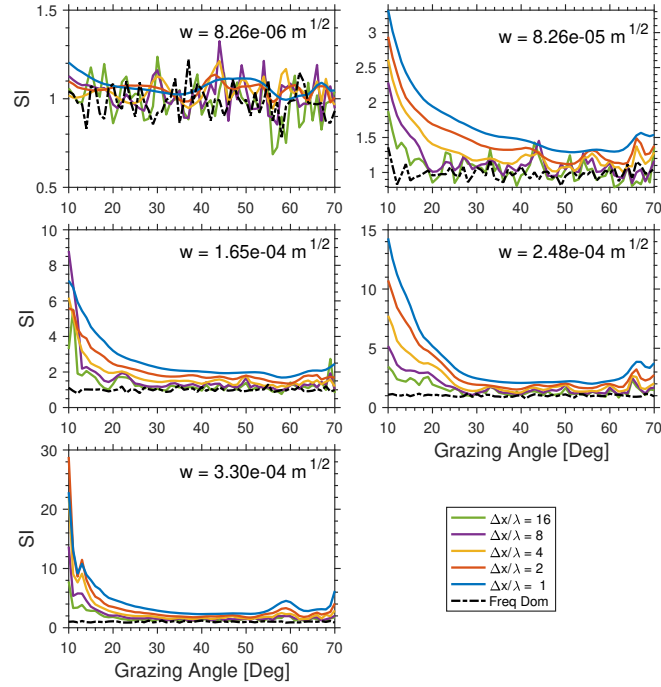


FIG. 7. Scintillation index 100 kHz, $\gamma = 2.5$

377 The angles are limited to a lower limit of 10° grazing angle due to finite surface length,
 378 and an upper limit of 70° , due to the difficulty in estimating the cross section near vertical
 379 using broadband pulses (Hefner, 2015; Hellequin *et al.*, 2003). To compare these results more
 380 closely, the broadband cross section is divided by the frequency-domain cross section and the
 381 dB value taken. This quantity, which we call the dB error, is plotted in Fig. 4. For all cases
 382 in these figures, there is insignificant difference between the apparent cross section in the
 383 time domain, and the frequency-domain cross sections for most w . At the largest spectral
 384 strength, some systematic oscillations as a function of θ_i are present, but cannot be easily
 385 disentangled from the rapid Monte-Carlo fluctuations. Other than that case, all differences
 386 appear to be random and within the uncertainty of the Monte-Carlo simulations. This is
 387 the expected result using Parseval's theorem, discussed in (Gauss *et al.*, 1996; Henyey *et al.*,
 388 1995), and confirms that broadband signals can successfully be used to estimate scattering
 389 strength from a power-law seafloor with spectral exponent $\gamma = 2$.

390 The scintillation index for this case is plotted in Fig. 5 for broadband signals as well
 391 as the single frequency case. Each spectral strength is plotted in its own subfigure, and
 392 each resolution has its own line within the subfigure. The vertical axis is SI , and the
 393 horizontal axis is grazing angle in degrees. The narrowband result is approximately unity
 394 for the entire angular domain shown, for all spectral strengths. This is the expected result
 395 if the central-limit theorem is employed, or if the scattered field is well-described in terms of
 396 Bragg scattering. For the broadband signals, there is a profound dependence on resolution,
 397 with the scintillation index increasing as the resolution cell becomes small. This behavior
 398 can be seen in the example realization shown in Fig. 2(c). As ΔX becomes small, the

399 intensity peaks become higher, even though q^2 is normalized by the ensonified length of the
 400 interface. Additionally, holding resolution constant, the scintillation index increases as the
 401 grazing angle becomes small, monotonically for this case. For most broadband cases, SI
 402 asymptotically approaches unity as grazing increases to its upper limit.

403 However, for the highest resolution cases, $\Delta X/\lambda = \{1, 2\}$, and the largest spectral
 404 strengths, this high angle asymptote is greater than one, indicating that for all angles ex-
 405 amined here, scattered complex pressure magnitude is non-Rayleigh. In (Lyons *et al.*, 2016)
 406 a K distribution was required to describe the pdf of the scattered field at moderate grazing
 407 angles, which agrees with this result. The Monte-Carlo fluctuations are significantly less
 408 than the difference between the high-angle SI asymptote and unity, indicating that this is a
 409 statistically significant finding. We also note that (Lupien, 1999) has observed non-Rayleigh
 410 scattering for broadband scattering from rough surfaces with a power-law exponent of $\gamma = 3$,
 411 but statistical tests barely rejected the Rayleigh distribution. That analysis did not remove
 412 the effect of the Gaussian taper, so the conclusions are not comparable to the present work.

413 Narrowband and broadband scattering scattering strength and scintillation index were
 414 also computed for 10 kHz, $\gamma = 2$, and spectral strengths that were ten times the value in
 415 the previous section. The relative resolution, $\Delta X/\lambda$ was held constant, but consequently
 416 the resolution ΔX was a factor of 10 larger. These parameters were chosen such that the
 417 dimensionless second-order quantities were the same as in the previous 100 kHz simulations.
 418 It was found that the scattering strengths, scattering strength dB error, and the scintillation
 419 index were the same for the 10 kHz and 100 kHz cases, to within Monte-Carlo fluctuation.
 420 This set of simulations was performed to verify that characterizing the simulations non-

421 dimensionally was valid. Since plots for the 10 kHz case do not add significantly new
 422 information, they are not shown here. These results indicate that departure from Rayleigh
 423 statistics is not isolated to high-frequency imaging systems, and may occur in in lower-
 424 frequency sonar systems as well, so long as the seafloor has the appropriate roughness
 425 parameters.

426 The spectral exponent was changed to $\gamma = 1.5$ to examine the effect of changing the shape
 427 of the power spectrum. New values of w were used, specified in Table I. Again, the scattering
 428 strength is the same whether computed at the center frequency, or using broadband pulses
 429 in the time domain. Scattering strength comparisons and the dB error are not shown. The
 430 scintillation index is plotted in Fig. 6, and it depends on angle, resolution, and spectral
 431 strength, as in the previous two cases. The SI curves were expected to behave similarly for
 432 $\gamma = 2$ and $\gamma = 1.5$ at $\Delta x/\lambda = 2$. While SI depends on resolution in a similar fashion, it is
 433 less than the SI for $\gamma = 2$, and is less sensitive to resolution, spectral strength, and grazing
 434 angle.

435 Finally, the spectral exponent was changed to $\gamma = 2.5$. The values of spectral strength can
 436 be found in Table I. Again, the scattering cross section computed at the center frequency and
 437 using broadband pulses were the same to within the Monte-Carlo error of the simulations,
 438 and are not shown here. The scintillation index is plotted in Fig. 7, and again depends on
 439 resolution, spectral strength and grazing angle. As the spectral strength is increased in the
 440 same proportions as the earlier plots (the second through fifth spectral strengths are 10, 20,
 441 30, and 40 times the smallest spectral strength respectively), the scintillation index increases

442 much more rapidly than either the $\gamma = 2$ and $\gamma = 1.5$ cases. For large values of spectral
 443 strength, the SI has some peaks near 60° .

444 VIII. DISCUSSION

445 In the results presented above, we have shown that scattering strength when estimated
 446 using broadband pulses is indistinguishable from its narrow-band quantity, and is indepen-
 447 dent of pulse length. This conclusion is not surprising, given the requirements of Parseval's
 448 theorem. However, as the pulse length changes, the properties of the ensemble used to
 449 estimate scattering strength changes as well. It is encouraging to see that although the
 450 ensemble is changing with respect to resolution (i.e the rough patch within a resolution cell
 451 is different for each resolution), σ is invariant to pulse length. We expect this result to hold
 452 for 3D environments as well if the roughness is isotropic. High-resolution systems are able
 453 to reliably estimate scattering strength, and it may be a stable feature for use in seabed
 454 classification. However, for highly anisotropic, non-stationary scenarios, such as those stud-
 455 ied by (Lyons *et al.*, 2010; Olson *et al.*, 2019, 2016), the measured scattering strength may
 456 depend on pulse length.

457 We have also shown that the scintillation index (also called structure (Wang and Bovik,
 458 2002; Wang *et al.*, 2004), lacunarity (Williams, 2015), or contrast (Marston and Plotnick,
 459 2015)) is highly dependent on all the parameters studied: resolution, grazing angle, spectral
 460 strength, and spectral exponent. For moderate to low grazing angles, SI monotonically in-
 461 creases as grazing angle decreases, resolution cell decreases, and spectral strength increases.
 462 SI is larger for larger γ . Contrary to scattering strength, SI , and therefore the scatter-

463 ing process in general, is fundamentally different in the frequency and time domains for
 464 broadband pulses.

465 In (Lyons *et al.*, 2016), it was hypothesized that the physical cause of heavy-tailed statis-
 466 tics in high resolution sonar imagery was local tilting of the seafloor due to roughness wave-
 467 lengths larger than the acoustic resolution. The scattered amplitude (envelope) was modeled
 468 as a product between a Rayleigh-distributed random variable (due to sub-resolution rough-
 469 ness), and a random variable that took into account the scattering cross section of the small
 470 scale roughness, evaluated at the nominal grazing angle modified by the local slope. Through
 471 the composite roughness approximation, local tilting (due to large-wavelength roughness
 472 components) modulates the Rayleigh-distributed field and causes the intensity variance to
 473 be greater than that of the Rayleigh distribution (i.e greater than unity). An rms slope with
 474 an upper cutoff of $2\pi/\Delta X$ was used as the input parameter to the composite roughness
 475 model, which was then used to compute the scintillation index. Given this interpretation,
 476 SI increases monotonically with rms slope, holding grazing angle constant.

477 A broad test of this hypothesis can be performed if we examine the resolution dependence
 478 of SI at a single angle and single spectral strength, but vary the spectral exponent. If γ is
 479 large, then the roughness will have less energy at high wavenumbers, and more if γ is small.
 480 If ΔX is decreased by, eg. a factor of two, then the effective rms slope will change as a
 481 function of γ . For γ of 1.5, 2 and 2.5 respectively, the effective rms slope will increase by
 482 a factor of 1.68, 1.41, and 1.19 respectively. Based on this interpretation, we would expect
 483 SI to be most sensitive to resolution for small γ

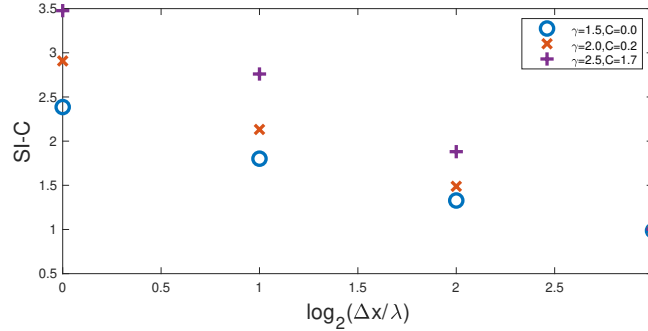


FIG. 8. SI for three different value of γ . A constant has been subtracted from SI for $\gamma = 2$, and $\gamma = 2.5$, such that the values at $\Delta x/\lambda = 8$ are all equal. Using this form, the sensitivity of SI on acoustic resolution can be seen to increase as γ increases. The grazing angle has been set to 20° , and the spectral strength is the second-largest value in Table XYZ.

484 This comparison is made in Fig. 8, which plots the scintillation index minus a constant
 485 C as a function of resolution and spectral exponent. Grazing angle is held constant at 20° ,
 486 and the second-highest values of spectral strength is used. The constant C was specified
 487 so that at $\Delta X/\lambda = 8$, each spectral exponent had the same abscissa on the plot, to better
 488 show the difference in slope. SI is least sensitive to ΔX for $\gamma = 1.5$, and most sensitive
 489 when $\gamma = 2.5$. This plot contradicts the hypothesis that rms slope at scales greater than
 490 the acoustic resolution is the sole driver of intensity fluctuations. Therefore local tilting is
 491 not, or not the only cause of heavy-tailed statistics. In the following subsections, we first
 492 examine the local tilting hypothesis in greater detail, and then explore other mechanisms
 493 that might be driving the SI behavior.

494 **A. Examination of the local tilting hypothesis**

495 We examine the local tilting hypothesis in finer detail first. If this hypothesis is true,
 496 then the intensity as a function of space ($I(t)$ mapped to x through $x = -ct/(2 \cos(\theta))$
 497 for backscattering), should increase if the large-scale slope at x is negative, and decrease if
 498 positive. Consequently there should be a statistical correlation between $I_{\Delta x}(x)$, the intensity
 499 at a given resolution at position x , and $s_{\Delta x}(x)$, the slope field low-pass filtered with an upper
 500 limit of Δx . This correlation can be quantified in a crude but straightforward manner using
 501 the Pearson product moment correlation coefficient, ρ , defined by

$$\rho(s_{\Delta x}, I_{\Delta x}) = \frac{E[s_{\Delta x}(I_{\Delta x} - E[I_{\Delta x}])]}{\sqrt{E[s_{\Delta x}^2] E[(I_{\Delta x} - E[I_{\Delta x}])^2]}}, \quad (25)$$

502 where the slope field is assumed to be a zero-mean process, but intensity is not. This
 503 coefficient quantifies the linear variation between a dependent and an independent variable.

504 We test the hypothesis of local tilting by filtering the slope field at different scales and
 505 forming the correlation coefficient with the intensity at a single resolution. The surface filter
 506 scale that maximizes the correlation is estimated, and this process is repeated for all acoustic
 507 resolutions. The acoustic resolution is ΔX , and the surface filter size is ΔX_{surf} . If the ΔX_{surf}
 508 that maximizes ρ , which is denoted ΔX_{max} , varies in proportion to Δx , then we can conclude
 509 1) that slope modulation is responsible in part for the intensity fluctuations, and 2) slopes
 510 at the scale of the acoustic resolution are responsible in part for the fluctuations.

511 The correlation coefficient for parameters of $\gamma = 2$, $\theta_i = 20^\circ$, and $w = 2 \times 10^{-5}\text{m}$ is
 512 plotted in Fig. 9. ΔX is on the horizontal axis, ΔX_{surf} is on the vertical axis, and ρ is
 513 denoted by grayscale. Holding ΔX constant, there is a distinct peak in ρ as a function of

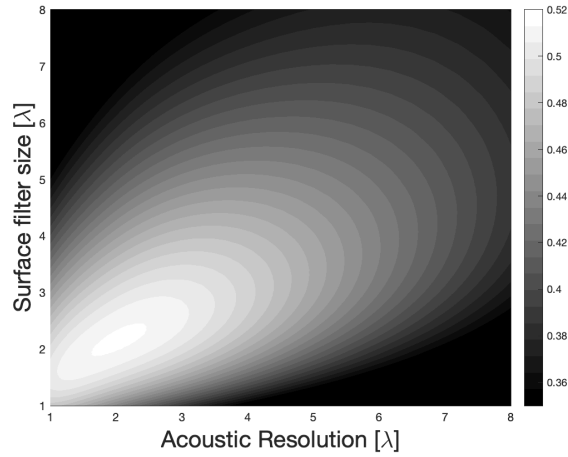


FIG. 9. Correlation coefficient as a function of the acoustic resolution, ΔX , and the surface filter size, ΔX_{surf} .

514 ΔX_{surf} . The peak value of ρ increases as the acoustic resolution becomes small, indicating
 515 that the role of slope modulation is greater for smaller resolution cells. Additionally, the
 516 peak in ΔX_{surf} varies with ΔX , indicating the local tilting hypothesis may be correct. This
 517 plot has a similar structure for other θ_i , γ , and w .

518 ΔX_{max} , the value of ΔX_{surf} that maximizes ρ , as a function of ΔX is plotted in Fig. 10.
 519 θ_i is constant at 20° grazing angle, each spectral strength is plotted as its own line. The
 520 different values of γ appear in subfigures. For each spectral strength and exponent, the
 521 maximum slope correlation scale varies monotonically with the acoustic resolution, except
 522 for the smallest spectral strength for $\gamma = 1.5$. For $\gamma = 2$, this trend is approximately linear,
 523 except for the smallest spectral strength that exhibits a slight negative curvature. The
 524 smallest spectral strengths for all values of γ resulted in $SI \approx 1$. For these situations, local
 525 tilting is not required to explain the behavior of SI .

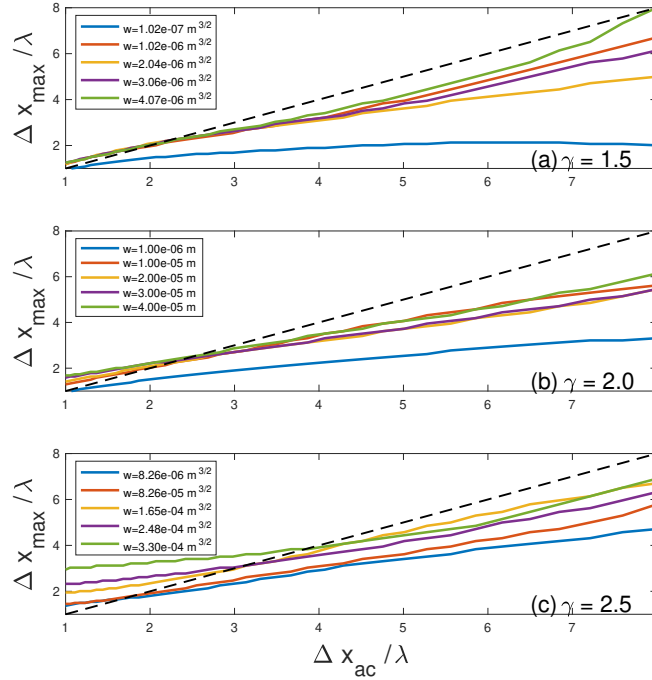


FIG. 10. Surface slope filter scale that maximizes the product moment cross correlation coefficient for each acoustic resolution. Grazing angle has been held constant at 20° , and each line represents a different spectral strength. Each subplot contains a different spectral exponent with (a) $\gamma = 1.5$, (b) $\gamma = 2$, and (c), $\gamma = 2.5$. Dashed black lines have a slope of unity and intercept of zero for reference.

526 For the other spectral exponents, the dependence of ΔX_{\max} on ΔX is also mostly linear,
 527 but has some negative or positive curvature. A line with unit slope and zero intercept is
 528 also plotted for reference. Although there is some departure from linearity, the monotonic
 529 dependence of ΔX_{\max} on ΔX indicates that, to within a proportionality constant, slopes at
 530 or larger than the resolution scale account for a significant part of the intensity fluctuations.
 531 Since the slopes are generally less than unity, this proportionality constant is less than unity,

532 a result that prompts further study into this effect. This plot provides evidence that the
533 local tilting hypothesis is one component of the intensity fluctuations.

534 **B. Other effects**

535 We have shown through the correlation coefficient that local tilting is in part responsible
536 for the intensity fluctuations, but cannot be the sole origin. The effect of local tilting is
537 modeled physically using the composite roughness approximation ([Jackson *et al.*, 1986b](#);
538 [McDaniel and Gorman, 1983](#)), in which the roughness is split into large-scale and small-
539 scale components. Perturbation theory is applied to the small-scale component, and the
540 Kirchhoff approximation to the large-scale component. When the grazing angle is small, the
541 Kirchhoff approximation becomes inaccurate, and multiple scattering and shadowing can
542 become important ([Liszka and McCoy, 1982](#)). Based on the analysis in ([Thorsos, 1988](#)),
543 shadowing of the incident field is important when the incident grazing angle is comparable
544 to the rms slope angle. Multiple scattering is expected to be important, when the incident
545 or scattered angles are comparable to twice the rms slope angle. Shadowing of the scattered
546 field is important when the scattered grazing angle is comparable to the rms slope angle.

547 In multiple scattering, the incident field interacts with distant parts of the rough surface
548 (defined as portions outside the resolution cell) multiple times before being radiated back to
549 the acoustic medium ([Liszka and McCoy, 1982](#)). Multiple scattering would cause an additive
550 component of the field with a lower amplitude and larger effective ensonified area than the
551 field due to the surface within the resolution cell. If the effective region of multiple scattering
552 were large enough, then scattered field would have contributions from many independent

553 locations, driving the statistics toward a Rayleigh distribution. Coherent combination of the
554 field scattered by the surface roughness located within the resolution cell (including tilting),
555 and the multiply scattered field would then result in a smaller SI than just scattering from
556 within the resolution cell (Johnson *et al.*, 2009; Lyons *et al.*, 2009; Watts, 1987).

557 Based on the analysis in (Thorsos, 1988, 1990), the Kirchhoff approximation fails due to
558 the presence of multiple scattering when there is significant energy in the roughness spectrum
559 at scales smaller than the wavelength. For power-law surfaces, a surface with a small spectral
560 exponent has more high-frequency energy than one with large spectral exponent. Therefore,
561 simulations with $\gamma = 1.5$ should have more significant multiple scattering than simulations
562 with $\gamma = 2$ or 2.5, and consequently the SI should be lower due to a large effective multiple
563 scattering region. Since we observe this trend in these numerical simulation, we believe that
564 this interpretation based on multiple scattering is a likely cause of the smaller scintillation
565 indices seen for $\gamma = 1.5$, and a lower sensitivity to the acoustic resolution. At this time,
566 this hypothesis cannot be confirmed more specifically, but requires further research.

567 Shadowing is also important when the grazing angles become small. A shadowing correc-
568 tion, which is essentially the expected value of the ratio of illuminated to the total surface
569 area, has been derived by (Wagner, 1967). A simplified form of this approximation is widely
570 used in the literature (Jackson *et al.*, 1986b; Thorsos, 1988, 1990) that depends on only
571 grazing angle and rms slope. This approximate form is based on very rough surface, and
572 vanishing correlation length, which seems to be useful for surfaces with a Gaussian roughness
573 spectrum. However the assumption of a vanishing lengths scale is inappropriate for power
574 law spectra, as they exhibit long-range correlation (Mandelbrot and Ness, 1968) when the

575 hurst exponent, $H = (\gamma - 2)/2 > 1/2$. A consequence of long-range correlation is that the
576 local mean tends to a non-zero value, even though its expected value is everywhere zero.
577 Thus if the surface has a local maximum, it may stay below that local maximum for some
578 distance before returning to it. Consequently at low grazing angles, the shadows cast by
579 local maxima can be quite long on power-law surfaces, especially when energy is concen-
580 trated in the low-wavenumber portion of the spectra, as it does for large γ . In terms of the
581 scintillation index, the geometric interpreting of shadowing causes parts of the surface to
582 be either illuminated, or not. This binary effect will increase the variance of the intensity,
583 leading to a higher scintillation index. Since we observe much higher scintillation indices
584 for $\gamma = 2.5$ ($H = 3/4$), and these surfaces also have larger rms roughness due to their
585 concentration at low wavenumbers, we believe that shadowing is a likely contributor to the
586 scintillation index. As before, this hypothesis cannot be confirmed here, but requires more
587 detailed analysis.

588 A physically accurate theoretical model that includes all of these effects is evidently
589 required to predict the scintillation index at low grazing angles. Such a model is, at this
590 time, not available, and is a fruitful opportunity for future research. We have noted that the
591 usual form of the shadowing correction for the intensity may be inappropriate for power-law
592 surfaces, especially when the spectral exponent is large, and the more general version derived
593 in (Wagner, 1967) may be required. The independence of the scattering cross section on
594 resolution established here imposes a useful constraint. Any theoretical model that breaks
595 apart the solution of the exact integral equation into physically interpretable phenomena
596 (such as tilting, shadowing, or multiple scattering) must also satisfy this constraint.

597 **IX. CONCLUSION**

598 In this work Fourier synthesis combined with numerical solution of the Helmholtz integral
599 equation has been performed to analyze the scattered field in terms of the scattering cross
600 section and scintillation index. We have examined the dependence of these two quantities on
601 acoustic resolution to understand the effects that modern high-resolution acoustic imaging
602 systems have on quantitative measurements of seafloor scattering. We have found that for
603 power-law surfaces, the scattering strength is independent of pulse length, which indicates
604 it is a stable quantity to use across measurement systems with different geometries. The
605 scintillation index depends strongly on pulse length. The behavior of SI on pulse length
606 indicates that scattering is fundamentally different in the time and frequency domains, and
607 that further research is needed to understand (or predict) intensity fluctuations in high
608 resolution broadband sonar systems.

609 Although simulations were performed in two dimensions, these results may be present in
610 three dimensions as well, if the roughness spectrum is assumed to be isotropic. The exact
611 values of the scintillation index will be different for 3D environments, as the rms slope is
612 calculated differently, and out of plane effects may be important.

613 Heavy-tailed, or non-Rayleigh scattering is commonly observed in scattering measure-
614 ments and is usually attributed to nonstationary, or patchy environments ([Abraham and](#)
615 [Lyons, 2004](#); [Lyons *et al.*, 2009](#)). Heavy-tailed statistics have been observed in seemingly
616 homogeneous seafloors by ([Lyons *et al.*, 2016](#)), and these numerical simulations have ver-
617 ified that statistically homogeneous surfaces can produce heavy-tailed statistics when in-

618 interrogated by a broadband high-resolution system. The composite roughness model used
619 in (Lyons *et al.*, 2016) was investigated, and found that it is in part responsible for inten-
620 sity fluctuations. Other sources of fluctuations, such as shadowing and multiple scattering
621 may be required to fully predict the scattered field. Further research is required to fully
622 understand this aspect of seafloor scattering.

623 We note a few consequences of heavy-tailed statistics arising in high-resolution systems
624 in homogeneous roughness environments. Heavy-tailed statistics are a significant source of
625 false alarms in acoustic target detection systems. Since scintillation index increases at low
626 angles, long range systems may suffer from decreased performance. The benefit of high
627 resolution systems, more pixels per target, may be offset due to the increased false alarms.

628 There may be some benefits of resolution dependence of the pdf of the scattered field.
629 Some autofocus algorithms for synthetic aperture systems (e.g. (Blacknell *et al.*, 1992; Cal-
630 low, 2003; Marston and Plotnick, 2015)), use the scintillation index, or contrast as their cost
631 function. If the acoustic field is entirely due to point scatterers, as is commonly assumed
632 (Brown *et al.*, 2017), then the pdf of the scattered field will be Rayleigh for all resolutions.
633 If an autofocus algorithm is applied, and the point spread function of the imaging algorithm
634 becomes smaller, then the field will still be Rayleigh and contrast will not increase. There-
635 fore, an autofocus algorithm based on SI or lacunarity will not be sensitive to the focus,
636 unless there are discrete scatterers in the scene. However, as show in this paper, SI is a
637 strong function of resolution for homogeneous power law surfaces, especially at low grazing
638 angles. Improving the focus at low grazing angles will lead to an increase in SI , and thus
639 the autofocus algorithm will be more sensitive to the actual degree of focus.

640 Median filters are often used to “remove” the speckle or intensity fluctuations from acous-
641 tic or electromagnetic images before use in remote sensing or target detection algorithms
642 (e.g. ([Williams, 2015](#), [2018](#)) and references therein). Although the scattering cross section,
643 which uses the arithmetic mean of the intensity, is insensitive to resolution, the median is
644 not, since the median is highly dependent on the probability density function. Thus either
645 the cross section can be used as a resolution independent quantity that has a large vari-
646 ance, or its variance can be reduced with the consequence that the pixel intensity will no
647 longer be directly related to scattering strength. Other methods of speckle reduction, such
648 as multilook, must be used if a reliable estimate of scattering strength.

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653 School.

654 ¹If phase coding and pulse compression is employed, this analysis deals with the resolution of the compressed,
655 or matched-filtered pulse.

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