
A framework to quantify uncertainties of seafloor backscatter from swath mapping echosounders

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Abstract :

Multibeam echosounders (MBES) have become a widely used acoustic remote sensing tool to map and study the seafloor, providing co-located bathymetry and seafloor backscatter. Although the uncertainty associated with MBES-derived bathymetric data has been studied extensively, the question of backscatter uncertainty has been addressed only minimally and hinders the quantitative use of MBES seafloor backscatter. This paper explores approaches to identifying uncertainty sources associated with MBES-derived backscatter measurements. The major sources of uncertainty are catalogued and the magnitudes of their relative contributions to the backscatter uncertainty budget are evaluated. These major uncertainty sources include seafloor insonified area (1–3 dB), absorption coefficient (up to > 6 dB), random fluctuations in echo level (5.5 dB for a Rayleigh distribution), and sonar calibration (device dependent). The magnitudes of these uncertainty sources vary based on how these effects are compensated for during data acquisition and processing. Various cases (no compensation, partial compensation and full compensation) for seafloor insonified area, transmission losses and random fluctuations were modeled to estimate their uncertainties in different scenarios. Uncertainty related to the seafloor insonified area can be reduced significantly by accounting for seafloor slope during backscatter processing while transmission losses can be constrained by collecting full water column absorption coefficient profiles (temperature and salinity profiles). To reduce random fluctuations to below 1 dB, at least 20 samples are recommended to be used while computing mean values. The estimation of uncertainty in backscatter measurements is constrained by the fact that not all instrumental components are characterized and documented sufficiently for commercially available MBES. Further involvement from manufacturers in providing this essential information is critically required.

Keywords : Multibeam echosounder, Calibration, Incidence angle

34 **1. Introduction**

35 Amongst acoustic sensors, multibeam echosounders (MBES) are commonly the tool of choice
36 for most seafloor studies because they concurrently offer high-resolution, co-located bathymetry
37 and backscatter (Hughes Clarke et al. 1996; Mayer 2006; Anderson et al. 2007; Lucieer et al.
38 2017). Historically the analysis of multibeam sonar data has focused on the bathymetric
39 component and the critical role it plays in nautical charting and in offering insights into geologic
40 and tectonic processes of the seafloor. The rich history of the use of MBES for critical mapping
41 applications has resulted in significant progress over the last two decades in quantifying the
42 sources of uncertainty associated with the bathymetric component of MBES (Hare et al. 1995;
43 Hare 2001; Calder and Mayer 2003; Lurton and Augustin 2010; Lucieer et al. 2015) adding
44 tremendously to the credibility and value of bathymetric data.

45 More recently, the interpretation of the second component of MBES systems, namely seafloor
46 backscatter, is playing an increasingly important role in many ocean-mapping applications
47 including habitat characterization, environmental monitoring, geological and geotechnical
48 studies, and natural resource prospecting (Lucieer et al. 2017). In support of these applications,
49 efforts have been made to use MBES backscatter to characterize the nature of the seafloor,
50 typically through broad descriptions of seafloor or sediment type (e.g. rock, sand, mud) or in
51 other instances, to further estimate basic parameters like grain size or acoustic properties (Hasan
52 et al. 2014 and references therein). Unlike for bathymetry, however, there has been little efforts
53 made to understand the uncertainty associated with MBES backscatter measurements and thus
54 methods of seafloor characterization using backscatter are not constrained with respect to
55 associated uncertainty.

56 The interpretation of backscatter data for seafloor characterization is typically done through the
57 analysis of backscatter mosaic texture or seafloor backscatter angular response. The backscatter
58 mosaic is a georeferenced image of the signal intensity scattered back to the sonar. With different
59 seafloor materials showing different intensity levels, mosaics can be used to segment the seafloor
60 into different types either subjectively by an interpreter, or more objectively through image
61 processing approaches (e.g. Reed and Hussong 1989; Brown et al. 2011; Diesing et al. 2016). As
62 the echo intensity varies with the angle of incidence of the acoustic signal at the seafloor, the
63 angular variations of backscatter have to be normalized (typically at 45°) for the mosaic to be
64 interpretable. As a result of this normalization process, a key quantitative aspect of the seafloor
65 properties (its angular response) is lost, hence limiting the use of mosaics to qualitative
66 interpretation (Schimel et al. 2015). Even when viewed qualitatively, the lack of knowledge of
67 the uncertainty associated with the backscatter levels depicted on a mosaic calls into question the
68 meaning of the interpretation. Issues of uncertainty in seafloor backscatter measurements have
69 become apparent when combining and comparing data sets from different MBES surveys (e.g.,
70 Hughes Clarke 2012; Lacharité et al. 2017) where surveys from different systems resulted in
71 wildly different backscatter results.

72 Unlike the backscatter mosaic, the analysis of the backscatter angular response allows for the
 73 extraction of quantitative features and algorithm-based seafloor characterization approaches
 74 (e.g., Fonseca and Mayer 2007). Such approaches can provide useful predictions of seafloor type
 75 provided that uncertainties are appropriately constrained (Fonseca et al. 2009; Rzhhanov et al.
 76 2012; Hasan et al. 2014) but suffer from the current lack of understanding of uncertainties in the
 77 underlying backscatter measurements. With more emphasis on automated and physical model
 78 driven characterization techniques, quantification of backscatter data is becoming more
 79 important (Alevizos et al. 2017) involving efforts in MBES calibration, and in better
 80 understanding, modelling, and estimating the associated uncertainty.

81 The aim of this paper is to identify the major sources of uncertainty for MBES-derived seafloor
 82 backscatter values, evaluate (when possible) their causes and estimate their magnitudes. In doing
 83 so, we hope to establish a framework for further analyses that may be broadly applied to various
 84 systems and situations so that end-users and operators may aspire to a more quantitative
 85 understanding of seafloor backscatter. We begin with a review of the basics of seafloor
 86 backscatter measurements. We then seek to identify the significant sources of uncertainty and
 87 quantify their respective magnitudes. Finally suggestions are made that might help mitigate the
 88 major sources of uncertainty.

89 2. Preliminary notions

90 2.1. Elements of backscatter measurement

91 MBES backscatter data result from the measurements of seafloor target strength (see e.g. Urick
 92 1983), a quantity that relates the incident and scattered pressure fields from a given target - in our
 93 case a small patch of the seafloor instantaneously insonified by the sonar signal. The ensemble
 94 average of squared scattered pressure $\langle |p_s|^2 \rangle$ is proportional to the insonified area A and the
 95 squared incident pressure $|p_i|^2$, and inversely proportional to the sonar-target squared distance
 96 r_s^2 , neglecting absorption and refraction effects:

$$\langle |p_s|^2 \rangle = |p_i|^2 A \sigma_b \frac{1}{r_s^2} \quad [\text{Eq. 1}]$$

97 where the proportionality coefficient σ_b is referred to as the “backscattering cross-section per
 98 unit area per unit solid angle” (Jackson and Richardson 2007 p 23); its logarithmic equivalent is
 99 the “bottom scattering strength” (Urick 1983):

$$S_b = 10 \log_{10} \sigma_b . \quad [\text{Eq. 2}]$$

101 The target strength (TS in dB re 1 m²) of the seafloor area A is then related to the scattering
 102 strength by:

$$TS = S_b + 10 \log_{10} A. \quad [\text{Eq. 3}]$$

104 $10\log_{10}A$ is used here instead of the correct form $10\log_{10}(A/A_0)$ for notation simplicity where A_0
 105 $= 1 \text{ m}^2$ is the reference unit surface. In the practical situation where TS is measured by a
 106 directional transmitter and receiver, the mean square voltage at the receiver output is expressed
 107 in dB as:

$$108 \quad 10 \log_{10} \left[\langle |V_{r(t)}|^2 \rangle \right] = EL + RS_o = SL_o + D_{TX} - 2TL + 10 \log_{10} A + S_b + RS_o + D_{RX} \quad [\text{Eq. 4}]$$

109 where $\langle |V_{r(t)}|^2 \rangle$ is the average squared voltage at receiver, EL the echo level at the receiver, RS_o
 110 the sensitivity of the receiver transforming the incident acoustic pressure into an electrical signal
 111 along its maximum response axis, SL_o the source level along its maximum response axis, $2TL$
 112 the two-way transmission loss, A the insonified area, D_{TX} and D_{RX} the transmit (Tx) and receive
 113 (Rx) directivity function values in the sonar-target propagation direction (Lurton 2010). The
 114 received voltage is then converted to a digital number DN through an Analog-to-Digital
 115 Converter (ADC) and recorded; this operation introduces a specific offset G_{AD} so that:

$$116 \quad DN = 10 \log_{10} \left[\langle |V_{r(t)}|^2 \rangle \right] + G_{AD}. \quad [\text{Eq. 5}]$$

117 The value of G_{AD} is related to how the digitization process is carried out, including the ADC's
 118 technological characteristics (Schimel et al. 2015). The measured backscatter strength can then
 119 be expressed from [Eq. 4] and [Eq. 5] as:

$$120 \quad S_b = DN - RS_o - G_{AD} - SL_o + 2TL - D_{TX} - D_{RX} - 10 \log_{10} A. \quad [\text{Eq. 6}]$$

121 For a given seafloor type and frequency, this value of S_b is also related to the seafloor incidence
 122 angle θ . The various uncertainty sources contributing to the measured S_b and θ are analyzed in
 123 the rest of this paper. In the following the S_b uncertainty expressed in dB relates to the
 124 percentage uncertainty in σ_b ; for example, a 1 dB uncertainty in S_b relates to a 10% uncertainty
 125 in σ_b .

126 **2.2. Sources of seafloor backscatter measurement uncertainty**

127 The expression [Eq. 6] for seafloor backscatter strength can be grouped as:

$$128 \quad S_b = \{DN - RS_o - G_{AD} - SL_o - D_{TX} - D_{RX}\} + \{2TL\} - \{10 \log_{10} A\} \quad [\text{Eq. 7}]$$

129 suggesting three main components of uncertainty:

- 130 1. The first component $\{DN - RS_o - G_{AD} - SL_o - D_{TX} - D_{RX}\}$ is the practical output of the
 131 target strength measurement, combining the measured echo level (DN), the source level
 132 (SL_o), the sonar Rx sensitivity (RS_o and G_{AD}) and directivity (D_{TX} and D_{RX}), but excluding
 133 the transmission losses ($2TL$). In the following it is conventionally designated as the
 134 "compensated echo level". Sources of its uncertainty include:

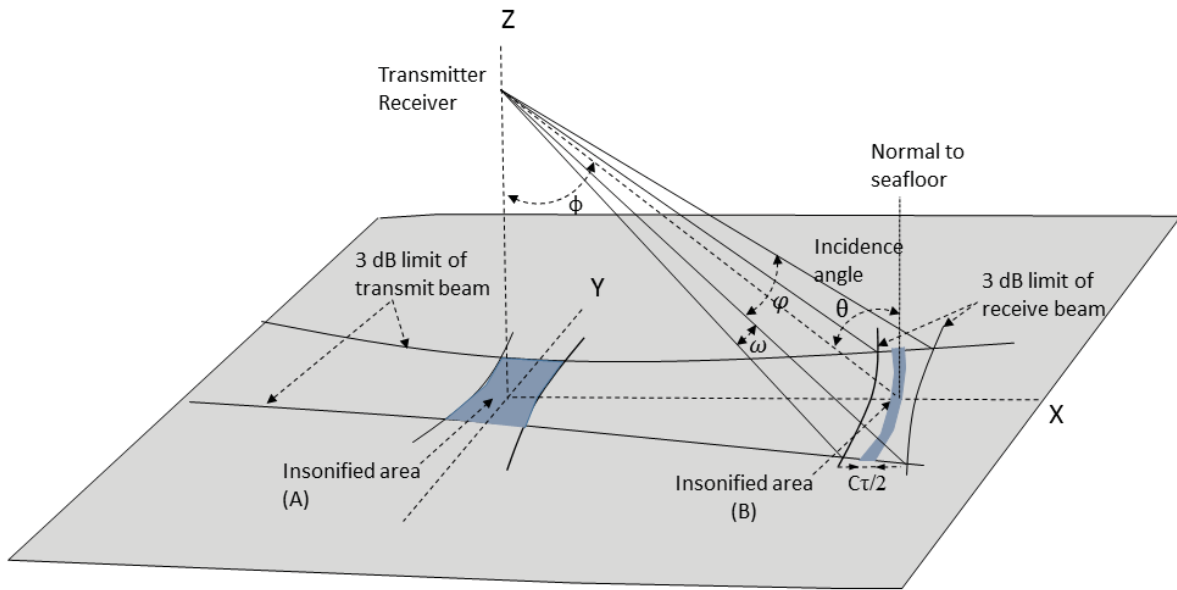
- 135 a) the stochastic nature of the physical echo intensity variations. An ensemble average
136 process helps in reducing the variance around the estimated mean, but as the number of
137 available samples is limited, some uncertainty remains in the backscatter estimate;
- 138 b) the sonar characteristics including electroacoustic (transducer sensitivity and directivity),
139 and electronic characteristics (Tx power amplification, Rx pre-amplification, various
140 gains, filtering, A/D conversion);
- 141 c) the environmental conditions (noise level added to the echo level).

142 The details of MBES-related uncertainty sources in (b) are not always available to end-users
143 and in the absence of this information the sounder must be considered a "black box", without
144 a real estimate of the uncertainty related to its actual transfer function. This uncertainty may
145 be globally determined from experimental data on a controlled target (Lurton and Lamarche
146 2015), but this can be an expensive, logistically difficult, and time-consuming process.

147 Additionally, not all MBES systems provide an estimate of S_b in the recorded data, but rather
148 only the DN values. Even when S_b values are explicitly provided in datagrams, they must
149 still be considered cautiously. Specific gains (either static or time-varying gain TVG) are
150 applied before digitization to keep the signal within the ADC input range; these must be
151 removed in order to retrieve the original physical S_b values. Such system-specific processing
152 steps, if not correctly implemented by the manufacturer, may result in large offsets in the
153 reported S_b . Several studies have highlighted these MBES-design shortcomings (Fonseca et
154 al. 2006; Gavrilov and Parnum 2010; Greenaway and Weber 2010; Hughes Clarke 2012;
155 Brown et al. 2015).

- 156 2. The second component $\{2TL\}$ is the two-way transmission loss between the sonar and the
157 target. It features both the geometrical divergence loss (function of the oblique range) and the
158 absorption loss (depending on both the range and the local absorption coefficient, a function
159 of frequency and water properties). The uncertainty in TL is mainly controlled by both the
160 range estimation accuracy and the knowledge of the seawater characteristics involved in
161 absorption.
- 162 3. The third component $\{10 \log_{10} A\}$ is the insonified footprint area instantaneously active in
163 the backscatter process delimited by the sounder beam pattern and/or the pulse duration. This
164 component also depends on the propagation range and the incident angle of the signal on the
165 seafloor (to be considered in a 3-D geometry) (Fig. 1).

166



167

168 Figure 1: Measurement geometry of MBES and area insonified for near nadir (A) and at
 169 oblique angle (B).

170 Additionally to these three sources of radiometric measurement uncertainty, the incidence angle
 171 estimation, upon which S_b is dependent, can be another major cause of uncertainty. The S_b
 172 dependence on seafloor incidence angle and frequency is a fundamental characteristic of seafloor
 173 backscatter data. With the MBES frequency fixed (or slightly varying with different T_x sectors),
 174 the mean seafloor angular response (AR) is characterized by its S_b values associated with
 175 incidence angles. Given such relationships, many research efforts have used comparisons of
 176 measured AR to theoretical models as a basis for seafloor segmentation and characterization (e.g.
 177 De Moustier and Alexandrou 1991; Hughes Clarke et al. 1997; Fonseca and Mayer 2007;
 178 Rzhanov et al. 2012). The uncertainty of the incidence angle is a function of T_x - R_x angle
 179 estimation accuracy, refraction by the sound-speed profile, and seafloor local slope. The position
 180 of the backscatter samples, similar to bathymetric samples, is determined through use of MBES
 181 geometry and positioning of the vessel. The Total Horizontal Uncertainty (THU) in the position
 182 of soundings, at the 95 percent confidence level, is not expected to exceed 5 meters + 5 percent
 183 of the depth (IHO 2008). For backscatter samples, the effect of position uncertainty is therefore
 184 assumed to be negligible in this paper.

185 3. Elementary analysis of major uncertainty components

186 As outlined above, the elementary analysis proposed here focuses on the magnitude of the S_b
 187 uncertainty broken down into the four parameters controlling the $S_b(\theta)$ estimate: compensated
 188 echo level, seafloor incidence angle, transmission loss, and insonified area. In evaluating the

189 sources of uncertainty, two significance thresholds of 1 dB and 1° are adopted here for
190 radiometric and geometric uncertainties respectively. These values are selected based on the
191 observation that in order to differentiate confidently between seafloor types, differences in
192 backscatter levels of approximately 1 dB are needed (Lucieer et al. 2017).

193 **3.1.Compensated echo level**

194 **3.1.1. Random fluctuations of the echo level and SNR**

195 The stochastic nature of the backscatter process results in a randomly-fluctuating sonar echo
196 level (Urick 1983). A simplified but widely used theoretical model assumes backscatter
197 amplitudes to follow a Rayleigh distribution, implying a standard deviation of 5.57 dB for
198 elementary backscatter samples (Dyer 1970; Jackson and Richardson 2007). Physically
199 interpreted, this model assumes an instantaneous insonified area (signal footprint) wide enough
200 to enclose a large number of simultaneously activated scatterers with statistically independent
201 random phases (Stanic and Kennedy 1992). In order to reduce the resulting uncertainty
202 associated with randomly fluctuating sonar echo levels, the backscatter level can be averaged
203 over an increasing number of signal samples (Peritsky 1973), however at the cost of degraded
204 resolution. For MBES measurements, the number of samples available for averaging depends on
205 depth, system parameters, and angular region of the measurement, and ultimately controls the
206 random uncertainty of the mean backscatter (Jackson and Richardson 2007). A more detailed
207 discussion of the statistical uncertainty of the echo-level can be found in Appendix A.

208 The echo level measurement uncertainty also depends on the signal-to-noise ratio (*SNR*). Noise
209 sources in ocean are numerous and highly variable (Urick 1983), including noise caused by sea-
210 surface agitation, biology, and bubbles created by the ship motion and/or surface wave action.
211 Also, the sonar performance may be limited by reverberation in the water column due to
212 biological, gaseous or inorganic scatterers. Self-noise caused by the sonar and its carrier platform
213 adds to these environment-related causes. A *SNR* better than 10 dB (Lurton and Augustin 2010)
214 can be taken as a reasonable lower limit for acceptable measurements of bathymetry according to
215 today's standards (IHO 2008). A generalized prediction of uncertainty caused by *SNR* is not
216 suggested here as there are too many causes and individual cases may degrade *SNR* up to a level
217 such that backscatter measurements are no longer possible. The S_b uncertainty due to *SNR* can be
218 simply modelled as:

$$\delta S_b = 10 \log_{10} \left(\frac{S+N}{S} \right) = 10 \log_{10} \left(1 + 10^{-\frac{SNR}{10}} \right) \quad [\text{Eq. 8}]$$

219 where S and N are the intensities of the expected signal and the additive noise respectively,
220 defining $SNR=S/N$. Assuming the worst case of a 10 dB *SNR*, the corresponding uncertainty in
221 backscatter measurements is around 0.4 dB (increase in the resulting average intensity for
222 {signal + noise} compared to signal alone). Therefore, while *SNR* can be a major uncertainty
223 source in some individual measurement scenarios, *SNR* can be practically considered as a minor
224 source of uncertainty for MBES data if currently acceptable quality for bathymetry is achieved.

225 Recommendations for improving the MBES data reliability in relation to *SNR* can be found in
226 Rice et al. (2015).

227 **3.1.2. Uncertainty of source level and receiver sensitivity**

228 A detailed characterization of uncertainty in the MBES parameters is still lacking (Lamarche and
229 Lurton 2017). MBES manufacturers have only offered nominal magnitudes of uncertainty related
230 to backscatter measurements. For example, for Kongsberg systems Hammerstad (2000) provided
231 a typical uncertainty of ± 1 dB related to MBES transducer sensitivities but cautioned that this
232 uncertainty might be larger for a specific system. Although several studies have attempted to
233 measure sonar sensitivity in calibration tanks and by field comparisons (Fonseca et al. 2006;
234 Lanzoni and Weber 2011; Rice et al. 2012; Welton 2014), MBES electronics are complex and
235 there are many causes of instrumental uncertainty that users cannot be expected to measure and
236 estimate, let alone keep track of the various engineering parameters needed to confidently
237 estimate these uncertainties. Involvement of MBES manufacturers is therefore critically needed
238 to model the MBES characteristics essential for calibration.

239 **3.1.2.1. Relative sonar calibration**

240 In absence of a readily available calibration documentation, users have to rely on empirical data
241 to derive the calibration offsets. Often, while repeating backscatter measurements over the same
242 seafloor using different settings or with different MBES systems, discrepancies in the observed
243 backscatter values are observed. These differences can then be estimated to adjust backscatter
244 values to match in a relative sense. This empirical method to make backscatter data consistent
245 among different settings or MBES systems is called relative calibration. The adjustment
246 protocols for relative calibration operations and the removal of systematic artifacts have been
247 studied extensively (Hughes Clarke et al. 1996; Hellequin et al. 2003; Augustin and Lurton 2005;
248 Llewellyn 2006; Fonseca et al. 2006, Chu and Hufnagle 2006; Fonseca et al. 2009; Parnum and
249 Gavrilov 2011; Teng 2011; Hughes Clarke 2012; Hiroji 2016). These relative calibration
250 protocols can provide valuable information about the overall health of the MBES including
251 system degradation due to transducer aging or bio-fouling (Lehaitre et al. 2008) and therefore are
252 also being incorporated into sonar acceptance protocols (Beaudoin et al. 2012; Rice and Malik
253 2015; Hauser et al. 2015). While such relative calibrations provide a means to have the same
254 seafloor appear to have consistent backscatter irrespective of different settings or MBES systems
255 used, it provides no indication of the actual backscatter uncertainty.

256 **3.1.2.2. Absolute sonar calibration**

257 As individual MBES systems may show differences in calibration from system to system, the
258 only alternative to manufacturer-provided information is to subject MBES to empirical checks in
259 a tank or at sea. The aim of this MBES calibration is to estimate the device-related parameters
260 required for S_b estimation including: transmit and receive beam patterns, pulse length and the
261 quantitative impact of gain changes applied during the data acquisition. Absolute calibration
262 using reference spheres is a well-accepted method developed for fisheries sonars and proposed

263 for application to MBES (Foote et al. 2005, Lanzoni and Weber 2011, Demer et al. 2015): using
 264 this method, the combined transmit and receive characteristics of the sonar are measured. The
 265 two-way beam pattern thus obtained can be used as a single correction to the measured
 266 backscatter. Since accurate placement and controlling motion of a reference sphere inside MBES
 267 narrow beam patterns are challenging, a calibration approach using extended targets has also
 268 been demonstrated (Heaton et al. 2013). An alternate method to target calibration (either sphere
 269 or extended target) is the use of a reference hydrophone (Demer et al. 2015); this method is
 270 required if transmitter and receiver characteristics need to be determined separately (Johannesson
 271 and Mitson 1983). For practical reasons, the use of hydrophones and transducers in a tank is
 272 suitable only for high-frequency portable systems with small arrays. Alternately this method has
 273 also been used to measure the beam pattern of a large array by fitting a hydrophone on an ROV
 274 (Fusillo et al. 1996), however, this approach is complex and expensive. Finally, using a reference
 275 seafloor patch as a benchmark (Eleftherakis et al. 2018; Ladroit et al. 2018) is an attractive
 276 option although the seafloor backscatter itself may change depending on a number of factors
 277 including temporal changes due to sediment movement and the formation of bedforms and other
 278 features that can cause seafloor backscatter to have strong dependence on azimuth (Lurton et al.
 279 2017).

280 Given that a general model for this class of drifting uncertainty cannot be defined and hence
 281 applied to quality control of backscatter data, the reality is that if a reduction in this source of
 282 uncertainty is desired, it is currently the user's responsibility to conduct regular calibration
 283 operations, either by test tank measurements, surveys on reference seafloor areas, or by
 284 comparison with calibrated sonar systems (Lurton and Lamarche 2015).

285 **3.2. Incidence angle**

286 The incidence angle considered in seafloor backscatter computations is the angle between the
 287 signal arrival direction at the seafloor and the local perpendicular to the interface (considered as
 288 locally flat although possibly tilted). The incidence angle uncertainty depends on three
 289 components:

- 290 A. The angle measured by the sounder at the receiving array (R_x), relative to the vertical.
 291 This measurement depends both on the intrinsic performance of the sensor array
 292 processing and on the platform motion (normally compensated for, with some
 293 instrumental uncertainty). The angles associated with the backscatter signal samples are
 294 referenced to the arrival angle at the sounding point (bottom detect) of the beam. Hence
 295 this instrumental uncertainty is equivalent to the one considered for the bathymetry
 296 uncertainty budget (Hare et al. 1995). Considering that most of the bathymetry relative
 297 error is given by its angle component (Lurton and Augustin 2010; Hare 2001):

$$\frac{\delta z}{z} = \tan\theta \cdot \delta\theta \quad [\text{Eq. 9}]$$

298 and using typical magnitudes met for acceptable-quality bathymetry data measured by
299 MBES, one finds an angle error around 0.15° for limit values of $\delta z/z = 1\%$ and $\theta = 75^\circ$.

300 This angular uncertainty is increased by the beam-pointing uncertainty caused by the ship
301 motion, but considering the high accuracy of today's motion sensors (typical uncertainty
302 for roll, pitch and heading accuracy is below 0.1°) the quadratically-cumulated angular
303 uncertainty due to both sensor and ship motion can be considered to stay below 0.2° and
304 hence can be neglected.

305 B. The effect of refraction due to propagation inside the water column. Uncertainties in the
306 estimated sound speed profile impact the accuracy of compensation for the refraction
307 effect. The sound speed profile has a twofold effect on incidence angle estimation: (1) the
308 beam steering angle at the sonar's head; and (2) refraction in the water column. Angular
309 uncertainty introduced in the computation of beam steering by a sound speed uncertainty
310 δc_s at the sonar head is given by (Hare et al. 1995):

$$\delta \phi_s = \frac{\tan \phi_s}{c_s} \delta c_s \quad [\text{Eq. 10}]$$

311 where ϕ_s is the beam steering angle from nadir and c_s is the sound speed at the sonar
312 head used for beam steering. In most MBES, the sound speed at the sonar head is
313 continuously measured by a dedicated probe, and therefore the sound speed uncertainty is
314 not expected to be more than ≈ 1 m/s. Considering a pessimistic $\delta c_s/c_s = 0.1\%$ (i.e.
315 $\delta c_s = 1.5$ m/s), the uncertainty in beam steering will be $\delta \phi_s \approx 0.2^\circ$ at $\phi_s = 75^\circ$.

316 Using the complete sound speed profile to compute an average value c_p the effect of an
317 uncertainty δc_p upon the incidence angle θ (referenced to nadir) can also be estimated as:

$$\delta \theta_p = \frac{\tan \theta}{c_p} \delta c_p \quad [\text{Eq. 11}]$$

318 giving the same magnitude of 0.2° in the pessimistic case of $\delta c_p/c_p = 0.1\%$ and $\theta = 75^\circ$.
319 So considering independent errors on c_s and c_p , the incident angle error magnitude
320 should stay within 0.3° .

321 In summary, the effect of beam steering and refraction on seafloor incidence angle is
322 negligible considering sound speed uncertainties remain smaller than 0.1% .

323 C. The seafloor local slope. This is best estimated from the Digital Terrain Model (DTM)
324 built from the MBES bathymetry. Three cases can be considered for evaluating the
325 seafloor slope influence on incidence angle uncertainty:

326 (i) The slope is completely ignored i.e., the seafloor is assumed to be flat and
327 horizontal. The error in the incidence angle will be equal to the slope of the
328 seafloor. This simplification is still commonly applied at basic levels of

329 backscatter processing but should be avoided in case of requirements of a good
330 quality backscatter level;

331 (ii) The seafloor topography is accounted for using a previously-determined DTM.
332 This is normally achievable by most modern seafloor-mapping sonars providing
333 both bathymetry and backscatter data. However, DTM slopes are subject to
334 uncertainties linked to the bathymetry measurement accuracy and to the details of
335 the processing steps applied for their construction;

336 (iii) Even for seafloor slopes inferred from a DTM, small-scale slopes in the
337 bathymetry may be unresolved and hence affect the estimate of local incidence
338 angle. Little can be derived from MBES bathymetric data about unresolvable
339 small-scale slopes and thus remains an unquantifiable uncertainty source.

340 In DTM slope calculations, the random vertical uncertainty in the soundings is
341 considered the most critical uncertainty source. Determining the uncertainty in slope
342 estimation, based on resolution, DTM uncertainty, analysis scale and computation
343 algorithm, is an active area of research in terrain analysis and modeling. Dolan and
344 Lucieer (2014) and Zhu et al. (2014) have shown uncertainties in slopes to reach up to 5°-
345 6° when using a MBES-derived DTM. Furthermore, assumptions about the macro-relief
346 of the surveyed seafloor at the spatial resolution of the backscatter samples are needed for
347 an *a-priori* estimate of slope uncertainty; for most MBES this cannot be assessed by
348 using only the bathymetry available from the MBES.

349 Although uncertainty due to the above individual sources (beam pointing angle, refraction and
350 seafloor slope) cannot be differentiated from the beam pointing angle measurement itself, the
351 incidence angle uncertainty affects the S_b measurement in two ways:

- 352 • the angle at which measured S_b is reported;
- 353 • the footprint area computation that impacts the echo level computation term ($\{10\log_{10}A\}$
354 in [Eq. 6]) as it is related to the incidence angle.

355 The magnitude of the impact of a wrong angle estimate on the resulting angular backscatter
356 curve can be demonstrated using the derivative (vs. angle) of a canonical angular backscatter
357 model. Using for instance the GSAB model (Lamarche et al. 2011) in its simplest form (a
358 Gaussian law for specular regime and Lambert's law at oblique incidences) leads to the results
359 presented in Fig. 2. The expressions for σ_b , its differential $\frac{\partial\sigma_b}{\partial\theta}$ and the corresponding uncertainty
360 δS_b in dB are given by:

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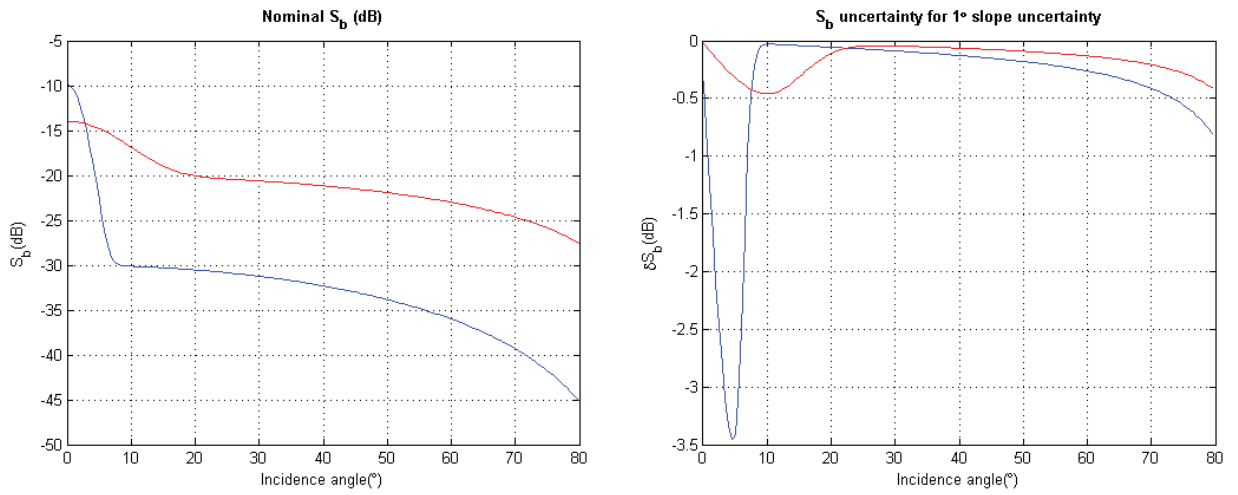
$$\sigma_b = A \exp\left(-\frac{\theta^2}{2B^2}\right) + C \cos^D \theta$$

$$\frac{\partial \sigma_b}{\partial \theta} = -\frac{A\theta}{B^2} \exp\left(-\frac{\theta^2}{2B^2}\right) - CD \cos^{(D-1)} \theta \sin \theta \quad [\text{Eq. 12}]$$

$$\delta S_b = \frac{10}{\ln 10} \frac{\partial \sigma_b}{\partial \theta} \frac{\delta \theta}{\sigma_b}$$

364

365 where A is the specular maximum amplitude, B is the facet slope standard deviation, C quantifies
 366 the average backscatter level at oblique incidence and D is the backscatter angular decrement.



367

368 Figure 2: Effect of incident angle uncertainty on backscatter. Two nominal angular backscatter curves
 369 representing different seafloor types (*blue* and *red*, *left*), and the effect of a 1° slope angle uncertainty
 370 on the backscatter values (*corresponding colors*, *right*). The impact is maximal in the specular region,
 371 where the cut-off effect corresponds to the strongest angular variations (0° to 10° or 0° to 20° according
 372 to the case); it is negligible in the “plateau” angle sector (10°-20° to 50°-60°) and increases at high
 373 incidence angles.

374 The two cases illustrated in Fig. 2 are typical angular backscatter curves for a soft-sediment (*in*
 375 *blue*, high narrow specular backscatter, decreasing in oblique region with $\cos^2 \theta$) and a coarse
 376 sediment (*in red*, low and wide specular backscatter, decreasing in oblique region with $\cos \theta$).
 377 For most seafloors the oblique-regime average angle dependence lies between the $\cos \theta$ and
 378 $\cos^2 \theta$ curves shown here. The model input parameters (A , B , C , D) are respectively (0.1; 2°;
 379 0.001; 2) and (0.03; 7°; 0.01; 1). As expected, the impact of incidence angle uncertainty is
 380 maximal for the specular regime; in this region its magnitude depends on the specular lobe slope
 381 and may reach several dB for a 1° angular change. On the other hand, the sensitivity to incidence
 382 angle uncertainty becomes negligible on the “plateau” regime (10°-20° to 50°-60°) where the S_b
 383 variation with angle is small. At higher angles (>70° in this example) the angle dependence
 384 increases again. In summary the angular dependence at steep angles varies strongly with the
 385 specular lobe, while the oblique regime shows a much more stable behavior regardless of the

386 seafloor type. This stability with angle is one of the major advantages of using the plateau region
 387 of incidence for backscatter measurements by MBES and should be preferentially used while
 388 comparing one backscatter survey to the other. This approach is also taken by space-borne radars
 389 which measure reflectivity only using a subset of the oblique regime (Long and Skouson 1996;
 390 Prigent et al. 2015).

391 **3.3. Transmission loss**

392 The transmission loss includes two effects (Urlick 1983): geometrical divergence (energy
 393 spreading along propagation path) and absorption (due to physicochemical properties of
 394 seawater). The one-way transmission loss (TL) referenced to a 1 m conventional range is
 395 classically written:

$$TL = 20 \log_{10} R + \alpha R \quad [\text{Eq. 13}]$$

396 where R is the range (in m), $20 \log_{10} R$ is the spherical spreading loss ($20 \log_{10} R$ is used instead
 397 of the correct form $20 \log_{10}(R/R_0)$ for notation simplicity, where $R_0 = 1$ m is the reference unit
 398 distance), and α is the absorption coefficient. Hence the uncertainty in TL will include the
 399 combined effects of uncertainties in the measured range (present in both terms) and the
 400 absorption coefficient.

401 **3.3.1. Range impact upon spreading loss**

402 The two-way spreading loss considered here is given by $2TL_s = 40 \log_{10} R$. The geometrical
 403 range R is determined by measurement of the time-of-flight t and the average sound speed \bar{c}
 404 between source and target, through the elementary relation

$$R = \frac{\bar{c}t}{2}. \quad [\text{Eq. 14}]$$

405 Therefore, the range uncertainty is due to both uncertainties in time measurement and average
 406 sound speed ($\delta\bar{c}$); its relative value is the quadratic summation of the values for time and sound
 407 speed, assumed to be independent:

$$R = \frac{\bar{c}t}{2} \Rightarrow \frac{\delta R}{R} = \sqrt{\left(\frac{\delta t}{t}\right)^2 + \left(\frac{\delta \bar{c}}{\bar{c}}\right)^2}. \quad [\text{Eq. 15}]$$

408 The minimum travel-time uncertainty δt is bounded by the sampling step of the digitized time
 409 signal, normally smaller than half the pulse duration. For instance, for a high-frequency MBES
 410 transmitting 0.2 ms pulses in a 50 m water depth (z), the range uncertainty is bounded by
 411 $\delta t = T/2 = 0.1$ ms, compared to a minimum two-way travel time of $2z/c = 66$ ms; so the relative
 412 error in this case is $\delta t/t \approx 0.15\%$. Note that an approximate linear scaling exists for the various
 413 categories of MBES for pulse duration vs. depth range; e.g. a low-frequency MBES typically

414 transmits 20 ms pulses in a 5000 m water depth, hence the same magnitude for $\delta t/t$ is expected
415 for different operational depths.

416 The $\delta \bar{c}$ magnitude arises from the sound speed measurement uncertainty, which is expected to be
417 better than 0.5 m/s (e.g. Sea-Bird Electronics Inc. 2010), as well as due to spatial and temporal
418 water column variability (Beaudoin et al. 2009). The relative uncertainty $\delta \bar{c}/\bar{c}$ integrated over
419 the water depth is not expected to be more than 0.1% ($\delta \bar{c} \sim 1.5$ m/s).

420 With these magnitudes of $\delta \bar{c}/\bar{c} = 0.1\%$ and $\delta t/t = 0.1\%$ the range-relative uncertainty expressed
421 in [Eq. 15] is about $\delta R/R = 0.18\% \approx 0.2\%$.

422 Finally, the associated spreading loss uncertainty is given by:

$$\delta_R(2TL_s) = 40 \log_{10}\left(1 + \frac{\delta R}{R}\right) \approx 40 \log_{10}(1.002) \approx 0.035 \text{ dB.} \quad [\text{Eq. 16}]$$

423 This result is independent of the range and is valid for all MBES categories and propagation
424 ranges. Moreover, the range term featured in the transmission loss is partly compensated by its
425 role in the footprint area A expression, proportional either to R or to R^2 . Thus, the actual final
426 dependence of the S_b value upon range will be $20 \log_{10} R$ or $30 \log_{10} R$, instead of $40 \log_{10} R$
427 and the maximum uncertainty in spreading loss, corresponding to [Eq. 16] should be either 0.018
428 dB (for $20 \log_{10} R$) or 0.027 dB (for $30 \log_{10} R$). To conclude, the S_b uncertainty caused by the
429 range uncertainty on the geometrical divergence component of the propagation loss is less than
430 0.03 dB and can be considered negligible.

431 **3.3.2. Range impact upon absorption loss**

432 The absorption loss is given by $2TL_{abs} = 2\alpha R$. Hence its range-dependent uncertainty for a δR
433 range variation is:

$$\delta_R(2TL_{abs}) = 2\alpha \delta R = 2\alpha R \frac{\delta R}{R} \quad [\text{Eq. 17}]$$

434 with the right-hand term containing the product of the absorption loss and the relative uncertainty
435 in range. The relative uncertainty in range is typically 0.2% or less; hence for a numerical
436 estimation of [Eq. 17] the magnitude of $2TL_{abs} = 2\alpha R$ has to be specified. Four cases are
437 considered here for different frequencies and maximum oblique ranges typical of various MBES
438 categories (deep, medium, shallow, very shallow) (Tab. 1). The results in Table 1 show that the
439 S_b uncertainty due to range in the absorption effect can reach a magnitude of 0.08 dB in the
440 worst cases (extreme oblique range, intermediate frequencies 30-100 kHz with a 0.1%
441 uncertainty in range) – and hence is a negligible effect.

442

443

444

445 Table 1: Uncertainty [Eq. 17] in transmission loss due to range uncertainty for four typical categories of multibeam
 446 echosounders.

MBES category	Deep	Medium	Shallow	Very Shallow
Frequency (kHz)	12	30	100	300
Approximate absorption coeff. α (dB/km)	1.2	6.7	33.2	72.5
Max depth z (m)	5000	2000	300	50
Max oblique range (m) $R_{max} = z/\cos 75^\circ \approx 4z$	20000	8000	1200	200
Max absorption loss (dB) $2\alpha R_{max}$	48.0	107.2	79.7	29.0
Uncertainty [Eq. 17] (dB) for $\delta R/R = 0.1\%$	0.04	0.1	0.08	0.03

447

3.3.3. Absorption coefficient

448 The $2TL_{abs}$ uncertainty due to an absorption coefficient uncertainty $\delta\alpha$ is given by:

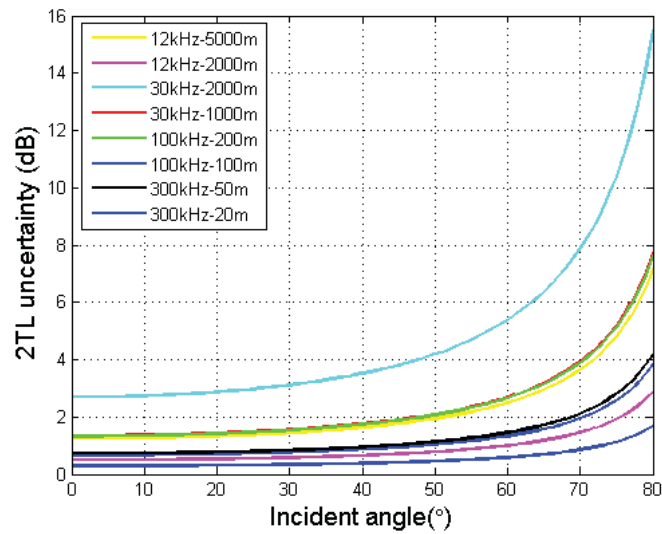
$$\delta_\alpha(2TL_{abs}) = 2R\delta\alpha = 2\alpha R \frac{\delta\alpha}{\alpha} \quad [\text{Eq. 18}]$$

449 where the relative uncertainty in absorption coefficient $\delta\alpha/\alpha$ has been made explicit. The
 450 absorption effect is a combination of the intrinsic absorption coefficient of the seawater
 451 (depending both on the absorption model reliability and on the accuracy of the measurements of
 452 estimates of local water properties) and the possible additional absorption caused by events in the
 453 water column such as bubble clouds (close to the surface or the ship's hull) or suspended
 454 sediments (close to the seafloor). The latter effect is more prone to impact high-frequency
 455 systems in shallow waters, while surface bubbles can impact systems in any water depth.
 456 Unfortunately, it is very difficult to assume *a priori* realistic magnitudes for such causes of
 457 uncertainty. The underlying physical phenomena controlling the intrinsic absorption coefficient
 458 of seawater are well understood and several models exist, based on fitting datasets of empirical
 459 measurements. Although more recent models have been proposed (Ainslie and McColm 1998),
 460 the model by Francois and Garrison (1982) is the most commonly used today, with a reported
 461 accuracy of 5%. To reduce this uncertainty, more direct observations of absorption coefficients
 462 are needed (Doonan et al. 2003). A rough estimate of uncertainty in transmission loss is
 463 proposed in Table 2 for an assumed $\delta\alpha/\alpha$ ranging from 1% to 10%.

464 Table 2: Uncertainty in transmission loss due to absorption coefficient uncertainties (1% and 10%) for four
 465 typically-used frequencies of MBES.

MBES category	Deep	Medium	Shallow	Very Shallow
Frequency (kHz)	12	30	100	300
Absorption coeff. α (dB/km)	1.2	6.7	33.2	72.5
Max depth z (m)	5000	2000	300	50
Max oblique range (m) $R_{max} = z/\cos 75^\circ \approx 4z$	20000	8000	1200	200
Max absorption loss (dB) $2\alpha R_{max}$	48.0	107.2	79.7	29.0
Max. $2TL_{abs}$ uncertainty (dB) for $\delta\alpha/\alpha = 1\%$	0.48	1.0	0.8	0.3
Max. $2TL_{abs}$ uncertainty (dB) for $\delta\alpha/\alpha = 10\%$	4.8	10	8	3

466 Therefore in the most probable practical cases of a few percent of relative uncertainty $\delta\alpha/\alpha$
 467 considered at the maximum oblique range of the sounder, the absorption uncertainty may reach
 468 several dB (up to 10 dB in the worst case of Tab.2). These estimates can be refined through a
 469 computation as a function of incident angle, for various frequencies and water depths; Fig. 3
 470 presents such results for a pessimistic $\delta\alpha/\alpha=10\%$. This figure illustrates that uncertainty in
 471 seawater absorption coefficient, even at lower levels, can be expected to be a major factor in the
 472 final S_b estimation accuracy, especially in the case of medium frequencies (30 and 100 kHz).



473
 474 Figure 3: Expected uncertainty (on $2TL$, or on S_b) resulting from a 10% uncertainty in absorption coefficient,
 475 based on the same parameters (frequency – water depth) as in Table 2.

476 In summary, the main factor to consider for the backscatter uncertainties due to transmission loss
 477 is the absorption coefficient which can result in uncertainties in backscatter estimates of several
 478 dB (Tab. 2, Fig. 3). The effects caused by the propagation range uncertainty are negligible in
 479 comparison.

480 3.4. Insonified area

481 Knowledge of the insonified area A is required to determine the backscatter strength defined per
 482 unit area ($10\log_{10}A$ in [Eq. 6]). In the classical Mill's cross configuration for MBES arrays, the
 483 insonified area extent in the along-track direction is defined by the T_x sector beamwidth (Lurton
 484 2010). For the oblique incidence region, the across-track extent of the insonified area is bounded
 485 by the pulse length projection over the seafloor, while in the normal incidence region, it is
 486 bounded by the receiver beamwidth (Fig. 1). The detailed accurate computation of the insonified
 487 area is complicated if both the full T_x and R_x beam patterns are considered, however,
 488 approximate formulas are commonly used. At oblique incidence (short-pulse regime, see Lurton
 489 2010) the insonified area can be approximated as:

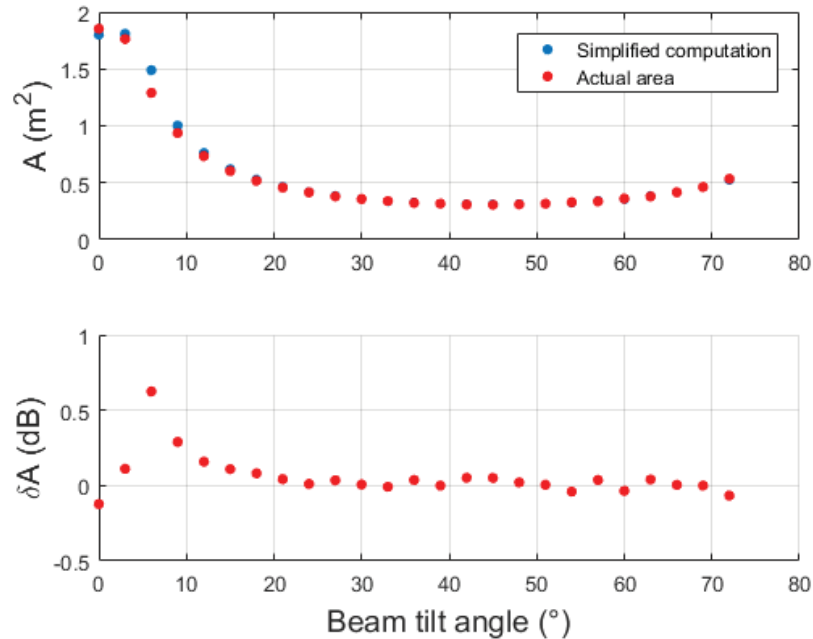
$$A \approx \varphi R \frac{cT}{2 \sin \theta \cos \gamma} \quad [\text{Eq. 19}]$$

490 and around normal incidence (long-pulse regime) as:

$$A \approx \varphi \omega R^2 \frac{1}{\cos \theta \cos \gamma} \quad [\text{Eq.20}]$$

491 with R the range; φ and ω the along-track and across-track two way equivalent apertures
 492 respectively (Fig. 1); T the pulse length; c the local sound speed; θ the across-track incidence
 493 angle; and γ the along-track slope. The pulse length T considered here is either the length of the
 494 physically transmitted pulse in case of continuous waves (CW) or the compressed pulse length
 495 after matched filtering in the case of frequency modulated (FM) transmitted signals (Lurton,
 496 2010).

497 These approximations [Eq. 19, Eq. 20] may lead to biases in the backscattering strength
 498 estimates. For narrow beams, this bias can practically be ignored (Hellequin et al. 2003).
 499 However, for wide beams, the bias can be significant, as shown for radar (Kim et al. 1982; Ulaby
 500 et al. 1983; Wang and Gogineni 1991) and sonar backscatter measurements (Matsumoto et al.
 501 1993). MBES beamwidths are today usually less than 2° so only a minimal effect on the
 502 insonified area is expected. Using a point-scatterer model (Ladroit et al. 2012) for a shallow-
 503 water MBES (0.15 ms pulse length, 1.5° along- and across-track beamwidths), a numerical
 504 simulation is presented here (Fig. 4) to illustrate the possible bias caused by the approximated
 505 formulae used for the insonified area. The area estimated using the simplified equations [Eq. 19,
 506 Eq. 20] matches fairly well with the simulated area defined by the idealized beam-pattern, for a
 507 range of depths (Fig. 4 shows an example in 50 m depth) except for a narrow intermediate
 508 angular range at the transition between the near-nadir and the oblique-angle regimes where the
 509 computed and simulated areas differ more significantly (up to ~ 0.5 dB in this example). Thus,
 510 the approximations used in footprint area computations can be applied to MBES data without
 511 causing significant uncertainty beyond the near-nadir region. The contribution of other terms in
 512 [Eq. 19, Eq. 20] in the insonified area estimation are discussed below.



513

514 Figure 4: Example of comparison of insonified area estimates based on simplified computation [Eq. 19, Eq.
 515 20] and actual area obtained by numerical simulation. At $\sim 6^\circ$, the simplified formula shifts from
 516 insonification limited by beam aperture to insonification limited by pulse length, resulting in a slight
 517 mismatch with the simulation results. Depth 50 m; pulse duration 0.15 ms; beamwidth 1.5° .

518 3.4.1. Range dependence

519 The impact of range uncertainty on footprint area is not considered here. It was considered above
 520 in the divergence transmission loss analysis and shown to be a parameter of secondary
 521 importance.

522 3.4.2. Sounder parameters

523 The sonar system parameters considered here are the beam apertures (φ and ω) and the pulse
 524 length T . Uncertainties in these terms can be caused either by shortcomings in the documentation
 525 provided by the manufacturer or by unwanted modifications in the MBES characteristics, for
 526 example failure of sonar array elements or inappropriate motion compensation (Gallaudet 2001;
 527 Hiroji 2016). In all cases, these uncertainties:

- 528 • act as stable biases on the measured/computed backscatter values and can be
 529 corrected *a posteriori* provided that their magnitude is identified;
- 530 • should not exceed a few percent, whatever their cause.

531 Table 3 gives the S_b uncertainties (in dB) associated with uncertainty of 1% to 20% for the input
 532 parameters of footprint A (with the $10\log_{10}$ dependence involved in [Eq. 6]). It is expected that
 533 the impact of these uncertainties in the footprint extent would remain small (although not

534 negligible, especially if accumulated), considering that the relative uncertainty on the sounder's
 535 parameters (beam patterns, pulse length) are likely to stay within few percent.

536 Table 3: S_b uncertainty caused by a relative uncertainty in individual components of insonified area A (beamwidths
 537 or pulse length), from 1% to 20%, expressed in dB (according to the $10\log_{10}A$ dependence in [Eq. 6]).

Relative uncertainty (%)	1	2	3	10	20
Uncertainty in dB	0.04	0.09	0.21	0.41	0.79

538 3.4.2.1. Across-track angle

539 Consider here first the angles in the across-track vertical plane (containing the formed beams).
 540 The sources of angle errors are presented in §3.2. If the across-track slope of the seafloor is taken
 541 into account when computing the incidence angle θ , an uncertainty $\delta\theta$ causes an uncertainty of
 542 A given by (for the short-pulse regime [Eq. 19]):

$$A \propto \frac{1}{\sin \theta} \Rightarrow \frac{\delta A}{A} = \frac{\delta \theta}{\tan \theta}. \quad [\text{Eq. 21}]$$

543 So the S_b uncertainty caused by angle variations in footprint area is given by:

$$\delta_{A,\theta} S_b = 10 \log_{10} \left(1 + \frac{\delta A}{A} \right) = 10 \log_{10} \left(1 + \frac{\delta \theta}{\tan \theta} \right). \quad [\text{Eq. 22}]$$

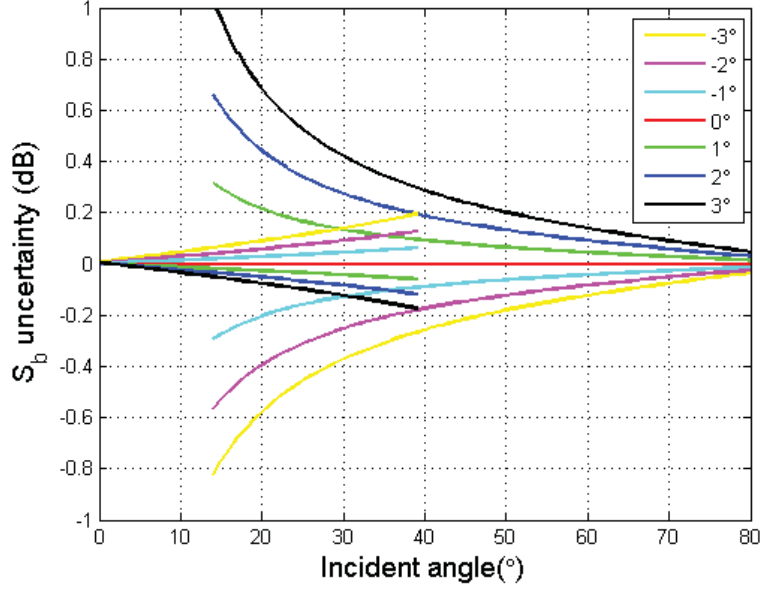
544 Note that normal incidence ($\theta \rightarrow 0$) is not considered here; the angle dependence [Eq. 18] on
 545 $1/\sin \theta$ is not valid in this regime and must be replaced by the long-pulse regime expression
 546 [Eq. 20]:

$$A \propto \frac{1}{\cos \theta} \Rightarrow \frac{\delta A}{A} = -\tan \theta \delta \theta, \quad [\text{Eq. 23}]$$

$$\delta_{A,\theta} S_b = 10 \log_{10} \left(1 + \frac{\delta A}{A} \right) = 10 \log_{10} (1 - \tan \theta \delta \theta). \quad [\text{Eq. 24}]$$

547 Fig. 5 shows the S_b uncertainty considering an uncertainty in the across-track incident angle (θ)
 548 from -3° to 3° , for the long- (0° to 40° incidence) and short-pulse (15° to 80° incidence) cases.
 549 The same slope shows reverse effects on the insonified area uncertainty using short- or long-
 550 pulse regimes, thus giving rise to a step change at the incidence angle where the insonified area
 551 shifts from the beam limited (long-pulse) to pulse limited (short-pulse) regime. Overall, the S_b
 552 uncertainty remains below 0.8 dB for slope-caused angle uncertainties reaching about $\pm 3^\circ$.

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554

555 Figure 5: Uncertainty in backscatter strength (S_b in dB) caused by variations in footprint area due to across-
 556 track incident angle uncertainty ranging from -3° to 3° for the long-pulse (0° to 40°) and the short-pulse
 557 cases (15° to 80°).

558 If no compensation is applied for the seafloor topography (assumed to be flat and horizontal),
 559 then the uncertainty of the footprint area estimate is the difference between the angular
 560 dependences $A \propto 1/\sin\vartheta$ (where ϑ is the incidence angle for an assumed flat topography) and
 561 $A \propto 1/\sin(\vartheta - \beta)$ (accounting for actual terrain slope β). Hence the uncertainty for the short-
 562 pulse regime is expressed in dB as:

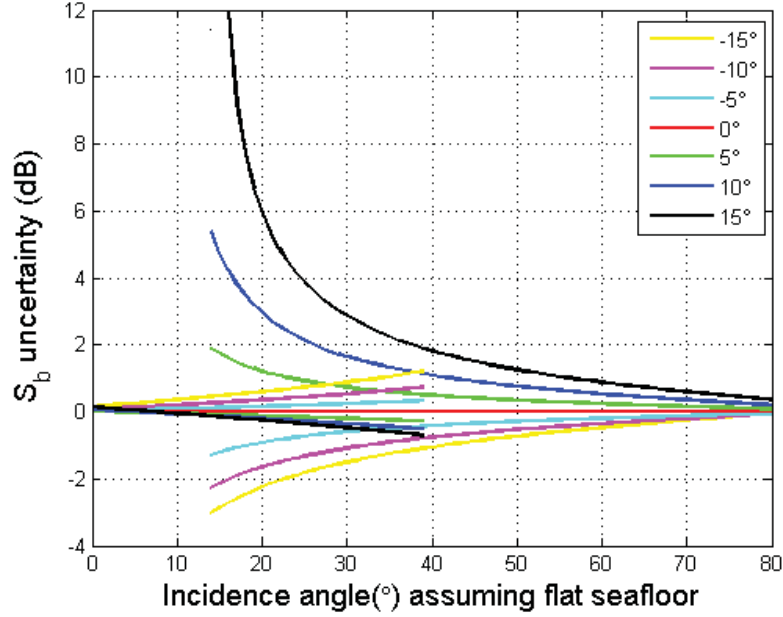
$$\delta_{A,\beta} S_b = 10 \log_{10} |\sin\vartheta / \sin(\vartheta - \beta)| \quad [\text{Eq. 25}]$$

563 Similarly for the long-pulse regime:

$$\delta_{A,\beta} S_b = 10 \log_{10} |\cos\vartheta / \cos(\vartheta - \beta)| \cdot \quad [\text{Eq. 26}]$$

564 The resulting S_b uncertainty is plotted in Fig. 6 as a function of incidence angle (0° to 80°) when
 565 the seafloor slope β (between -15° and $+15^\circ$) is not accounted for, for the long- (0° to 40°
 566 incidence) and short-pulse (15° to 80° incidence) regimes. For the long-pulse case, the
 567 uncertainty is on the order of 1 dB for steeper slopes (15°), however, for the short-pulse region
 568 the uncertainty in the seafloor for slopes facing towards the MBES causes large uncertainty in S_b
 569 (e.g., > 3 dB for $\vartheta = 15^\circ$ at $\theta < 30^\circ$). Figs. 5 and 6 indicate that the impact of across-track seafloor
 570 slope uncertainty is significant and most severe at mid-range incidence angles (20° - 50°).

571



572

573 Figure 6: Uncertainty in backscatter strength (S_b in dB) if the seafloor across-track slope is not considered for
 574 area insonified computation. Unaccounted seafloor slopes from -15° to 15° are considered for the long-
 575 pulse (0° to 40°) and short-pulse cases (15° to 80°).

576 **3.4.2.2. Along-track angle**

577 An uncertainty $\delta\gamma$ in the along-track incidence angle γ causes an uncertainty in the insonified
 578 area A given by [Eq. 19, 20]:

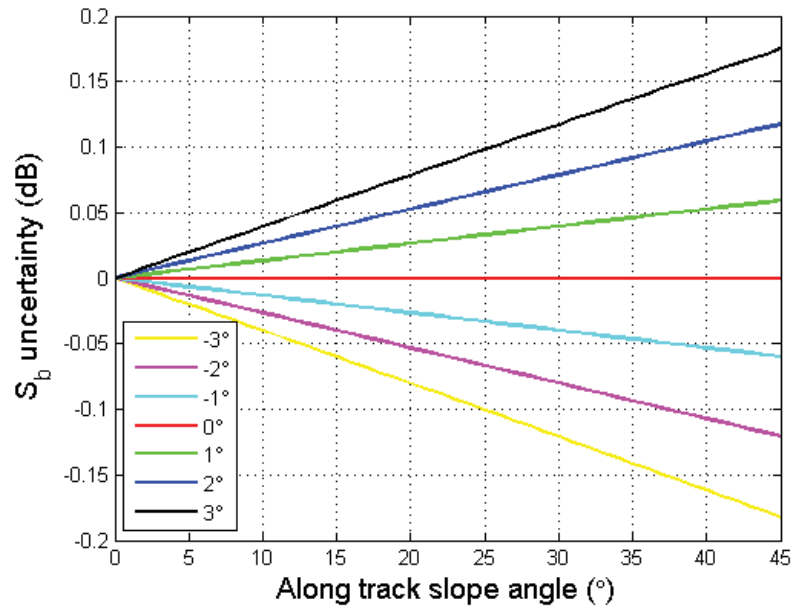
$$A \propto \frac{1}{\cos \gamma} \Rightarrow \frac{\delta A}{A} = \tan \gamma \delta \gamma \quad [\text{Eq. 27}]$$

579 The uncertainty in S_b can then be estimated as:

$$\delta_{A,\gamma} S_b = 10 \log_{10} \left(1 + \frac{\delta A}{A} \right) = 10 \log_{10} (1 + \tan \gamma \delta \gamma) \quad [\text{Eq. 28}]$$

580 For $\delta\gamma$ ranging from -3° to 3° Fig. 7 shows that the S_b uncertainty is insignificant for small
 581 uncertainties in the incidence angle (1 or 2°) and/or terrains with smooth topography (along-track
 582 slope angles up to 10° to 15°); even for steep areas with higher uncertainties in the topography,
 583 the S_b uncertainty remains within a few tenths of a dB.

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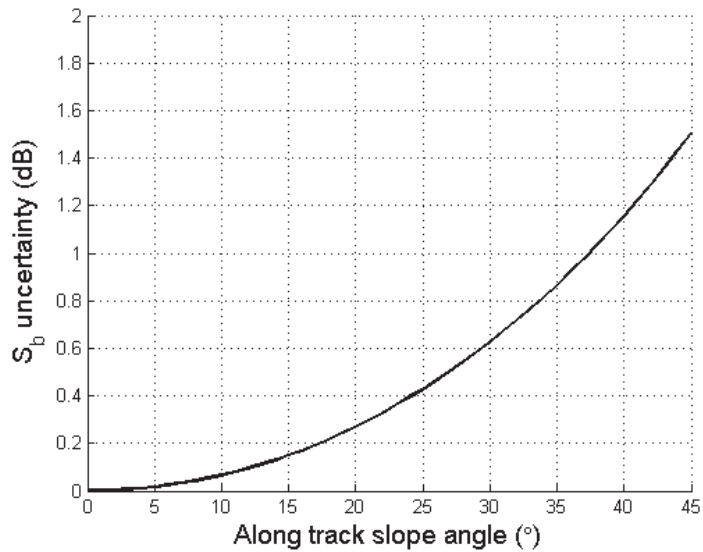
Figure 7: Uncertainty in S_b estimation due to uncertainty in along-track slope.

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If the along-track slope angle effect is not accounted for (as is often the case), the uncertainty is then directly given by the $1/\cos \gamma$ term (Fig. 8), where γ represents the slope angle. Here again, the S_b uncertainty may be negligible for smooth terrains (< 0.1 dB for incidence angles up to 15°) but increases significantly for steeper slopes (> 0.5 dB for slopes 30° to 45°). Note these results for along-track angles are valid for short- and long-pulse regimes, since both regimes have the same dependence on $1/\cos \gamma$ (see [Eq. 19, Eq. 20]).

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Figure 8: Uncertainty in S_b estimation if the along-track slope (0° to 45°) is ignored.

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In summary, for the compensation of the insonified area, the impact of inaccuracies in the sounder characteristics remains limited and can reasonably be kept small or negligible. The S_b dependence on incidence angle has a far more significant impact, potentially reaching several dB, depending on the beam angle and local seafloor slope. While computing the insonified area, the across-track slope angle plays the major role, while the along-track angle impact remains limited. Completely ignoring the seafloor slope (both across- and along-track) when estimating the footprint extent, logically leads to the largest uncertainties. Fortunately, commercially available backscatter processing software tools have started to address such compensations (e.g., QPS 2014). However, even when accounting for the local slopes, uncertainties of a few degrees may remain and the relations provided above can be used to assess their uncertainty contributions.

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3.5. Summary of the major uncertainty components

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Based on the elementary analysis presented above, the impacts of the main sources of backscatter uncertainty are summarized in Table 4. Each of the causes of S_b uncertainty is broken down into “random” or “bias” components. “Random” uncertainties are caused by noise or intrinsic fluctuations (e.g., echo signal instabilities, or small-scale uncertainties in the bathymetry) and can be mitigated through *a posteriori* statistical processing. “Bias” or systematic uncertainties may be caused by variations in the MBES characteristics, by unaccounted changes in environmental conditions, or by insufficiencies in the processing procedures; they may systematically vary as a function of depth, seafloor slope and ship’s motion. They can (up to some point) be corrected *a posteriori*, although this implies complementary operations that may prove difficult (e.g., sonar calibration, re-computation of the DTM, improved information about the water column). The bias corrections, once applied, still

620 have some residual uncertainty that must then be included in the uncertainty budget. The
 621 following scale is proposed to classify the magnitude of the uncertainty:

- 622 • *Negligible (N)* : 0.01 to 0.1 dB
- 623 • *Small (S)*: 0.1 to 1 dB
- 624 • *Moderate (M)* : 1 to 3 dB
- 625 • *High (H)* : 3 to 6 dB
- 626 • *Prohibitive (P)* : beyond 6 dB

627 Table 4: Major sources of uncertainty for compensated echo-level, source level (*SL*), transmission losses (*TL*),
 628 insonified area (*A*), and seafloor incidence angle. See the code (*N-S-M-H-P*) definition in the text. Uncertainties are
 629 categorized as Bias or Random uncertainty based on their effect on the measurement.

Measurement component	First-order uncertainty sources	Bias	Random	Magnitude	Possible quality improvement
<i>Compensated Echo Level</i>	<i>Signal fluctuations</i>		✓	<i>M to H (5.57 dB std.dev. for a Rayleigh distrib.)</i>	<i>Decreased (S) by data averaging (at the expense of resolution)</i>
	<i>Noise Level</i>		✓	<i>S in most cases</i>	<i>Improve sonar performance</i>
	<i>Sonar parameters (without calibration)</i>	✓		<i>Unpredictable – up to P</i>	<i>Calibration</i>
	<i>Sonar parameters (after calibration)</i>		✓	<i>N to S</i>	<i>Calibration accuracy</i>
<i>Incidence angle</i>	<i>Seafloor slope (compensated)</i>	✓	✓	<i>N to M</i>	<i>Bathymetry DTM accuracy</i>
	<i>Seafloor slope (ignored)</i>	✓	✓	<i>N to P according to topography</i>	<i>DTM for slope compensation</i>
<i>Area (A)</i>	<i>Footprint model approximation</i>	✓		<i>N to S</i>	-
	<i>Incidence angle (refraction, seafloor slope)</i>	✓	✓	<i>S to M. Possibly H to P (if seafloor slope ignored)</i>	<i>Improved accuracy in SVP and DTM</i>
	<i>Sonar parameters</i>	✓		<i>S</i>	<i>Constructor's information</i>
	<i>Propagation range</i>		✓	<i>N</i>	-
<i>Transmission Loss (TL)</i>	<i>Absorption coefficient</i>	✓		<i>S to H</i>	<i>Water column absorption profile</i>
	<i>Propagation range</i>		✓	<i>N</i>	-
	<i>Frequency differences (ignored)</i>	✓		<i>N to M</i>	<i>Sector frequency accounted for</i>

	<i>Water column anomalies</i>	✓		<i>N to P</i>	<i>Water column properties</i>
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630 **4. Conclusions**

631 This work has attempted to identify and model the major causes and magnitudes of backscatter
632 uncertainties from MBES systems. Unraveling the complexities of backscatter measurements is a
633 considerable task, and the approach outlined here is far from complete; however, it is hoped that
634 it offers a framework from which further understanding of the sources and magnitude of
635 backscatter uncertainties can be derived.

636 The elementary uncertainty analysis proposed here identified the major components of the
637 uncertainty budget (Tab. 4):

- 638 • The uncertainty in fluctuating and unreferenced measured echo levels is due to both the
639 random character of the echo intensity (causing noise-like fluctuations to be processed
640 statistically) and the incomplete knowledge of the MBES calibration parameters (leading to
641 biases). The statistical uncertainty can be controlled by averaging a number of samples into a
642 mean echo level with the understanding that increasing this number degrades resolution and
643 thus a trade-off has to be made between resolution and uncertainty. In contrast, the
644 uncertainty stemming from inaccurate values of MBES characteristics can reach
645 unpredictable and unacceptable magnitudes if appropriate calibration operations have not
646 been conducted nor reference data collected. MBES manufacturers should play a key role in
647 addressing this issue by providing the information needed to better document and reduce this
648 fundamental component of uncertainty, which is difficult to detect in the field data and
649 whose accurate evaluation is rarely accessible to users.
- 650 • The uncertainty in seafloor incidence angle measurement is mostly affected by seafloor slope
651 uncertainty controlled by the resolution and accuracy of bathymetric data used for DTM
652 production (if used at all). Greater attention must be placed on the incorporation of bi-
653 dimensional slope compensation inside the backscatter data processing tools and on the
654 improvement of local slope determination from the bathymetry data. This uncertainty
655 obviously impacts the computation of the backscatter angular response. Moreover, if not
656 accounted for, slope is often the major cause of error in the insonified area computation. The
657 sounder characteristics are normally sufficiently well known for the impact of their
658 uncertainty to remain acceptable; this again falls under the manufacturer’s responsibility.
- 659 • The transmission loss uncertainty is almost exclusively due to the absorption coefficient
660 estimation, the inaccurate estimation of which can have a significant impact on the
661 backscatter level estimation; however the combination of the measurement of temperature
662 and salinity values over the full water column with appropriate procedures for compensation
663 can keep the impact of the absorption coefficient within acceptable limits. The impact of
664 local perturbations of the water column properties is not well-understood and deserves
665 further investigation, although the use of ocean atlas data or ocean models can help to

666 mitigate this problem. Unexpected phenomena such as bubble clouds sweeping the MBES
667 arrays cause specific issues that are impossible to quantify in advance; however their joint
668 impact on the objective quality of bathymetry data can help detect their presence and justify
669 to disregard corrupted data.

670 This study was conducted as an initial step in the identification of the fundamental causes and
671 estimation of order-of-magnitude levels of the uncertainties associated with the collection of
672 MBES backscatter data. It has shown that it is difficult to predict broadly applicable numerical
673 values, since many of the major uncertainty sources vary on a case-to-case basis. Future efforts
674 need to be directed towards better provision of sonar characteristics from the manufacturers,
675 improvement of MBES calibration methods, and quantification of their reliability and objective
676 uncertainty. A second area of investigation is the impact of unexpected perturbations of the
677 seawater column properties (e.g. bubble clouds). Both topics suggest the need for new well-
678 designed field experiments and would benefit greatly from collaborative efforts of the concerned
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685 **Disclaimer**

686 The scientific results and conclusions, as well as any views or opinions expressed herein, are
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690 **References**

691 Ainslie MA, McColm JG (1998) A simplified formula for viscous and chemical absorption in
692 sea water. *Journal of the Acoustical Society of America*, 103(3): 1671-1672
693 <https://doi.org/10.1121/1.421258>

694 Alevizos E, Snellen M, Simons D, Siemes K, Greinert J (2017) Multi-angle backscatter
695 classification and sub-bottom profiling for improved seafloor characterization, in Lamarche, G.,
696 and Lurton, X. (Eds) *Seafloor backscatter data from swath mapping echosounders: From
697 technological development to novel applications*, Marine Geophysical Research
698 <https://doi.org/10.1007/s11001-017-9325-4>

699 Anderson JT, Holliday DV, Kloser RJ, Reid D, Simard Y (Ed.) (2007). *Acoustic seabed
700 classification of marine physical and biological landscapes*. ICES Cooperative Research Report
701 No 286; 183 pp ISBN 87-7482-058-3

702 Augustin JM, Lurton X (2005) Image amplitude calibration and processing for seafloor mapping
703 sonars. *Oceans 2005 - Europe*, 1, 698-701 Vol 1
704 <https://doi.org/10.1109/OCEANSE.2005.1511799>

705 Beaudoin J, Hiebert J, Calder B, Imahori G (2009) Uncertainty Wedge Analysis: Quantifying the
706 impact of Sparse Sound Speed Profiling Regimes on Sounding Uncertainty. Center For Coastal
707 and Ocean Mapping Paper 453, U.S. Hydrographic Conference

708 Beaudoin J, Johnson P, Lurton X, Augustin JM (2012) R/V falkor multibeam echosounder
709 system review UNH-CCOM/JHC Technical Report 12-001 September 4, 2012 57p
710 http://mac.unols.org/sites/mac.unols.org/files/20120904_Falkor_EM710_EM302_report.pdf Last
711 accessed June 2017

712 Bjørnø L (2017) *Applied Underwater Acoustics*, Neighbors TH, Bradley D (Eds) First edition.
713 Amsterdam, Netherlands. Elsevier, ISBN-13: 978-0128112403

714 Brown CJ, Smith SJ, Lawton P, Anderson JT (2011) Benthic habitat mapping: A review of
715 progress towards improved understanding of the spatial ecology of the seafloor using acoustic
716 techniques. *Estuarine, Coastal and Shelf Science*, 92(3): 502-20
717 <https://doi.org/10.1016/j.ecss.2011.02.007>

718 Brown CJ, Schmidt V, Malik M, Bouffant N (2015) Backscatter measurement by bathymetric
719 echo sounders. Report Chapter. Lurton X, Lamarche G (Eds) (2015) Backscatter measurements
720 by seafloor-mapping sonars. Guidelines and Recommendations. 200p [http://geohab.org/wp-](http://geohab.org/wp-content/uploads/2013/02/BWSG-REPORT-MAY2015.pdf)
721 [content/uploads/2013/02/BWSG-REPORT-MAY2015.pdf](http://geohab.org/wp-content/uploads/2013/02/BWSG-REPORT-MAY2015.pdf) Last accessed June 2017

722 Calder B, Mayer L (2003) Automatic processing of high-rate, high-density multibeam
723 echosounder data. *Geochemistry, Geophysics, Geosystems*, 4(6): 1048:1-22.
724 <https://doi.org/10.1029/2002GC000486>

725 Chu D, Hufnagle LC (2006) Time varying gain (TVG) measurements of a multibeam echo
726 sounder for applications to quantitative acoustics. In *IEEE/OCEANS 2006* (pp 1-5). September
727 18-21, Boston, MA, USA <https://doi.org/10.1109/OCEANS.2006.306818>

728 Demer DA, Berger L, Bernasconi M, Bethke E, Boswell K, Chu D, Domokos R, et al. (2015)
729 Calibration of acoustic instruments. ICES Cooperative Research Report No. 326. 133 pp.

730 De Moustier C, Alexandrou D (1991) Angular dependence of 12-kHz seafloor acoustic
731 backscatter. *The Journal of the Acoustical Society of America*, 90(1): 522-531
732 <https://doi.org/10.1121/1.401278>

733 Diesing M, Mitchell P, Stephens D (2016) Image-based seabed classification: what can we learn
734 from terrestrial remote sensing?, *ICES Journal of Marine Science*, 73(10):2425–
735 2441 <https://doi.org/10.1093/icesjms/fsw118>

736 Dolan MF, Lucieer VL (2014) Variation and uncertainty in bathymetric slope calculations using
737 geographic information systems. *Marine Geodesy*, 37(2): 187-219
738 <https://doi.org/10.1080/01490419.2014.902888>

739 Doonan J, Coombs R, McClatchie S (2003) The absorption of sound in seawater in relation to
740 the estimation of deep-water fish biomass. *ICES Journal of Marine Science*, 60(5): 1047–1055
741 [https://doi.org/10.1016/S1054-3139\(03\)00120-6](https://doi.org/10.1016/S1054-3139(03)00120-6)

742 Dyer I (1970) Statistics of sound propagation in the ocean. *The Journal of the Acoustical Society*
743 *of America*, 48(1B): 337-345 <https://doi.org/10.1121/1.1912133>

744 Eleftherakis D, Berger L, Le Bouffant N, Pacault A, Augustin JM, Lurton X (2018) Backscatter
745 calibration of high-frequency multibeam echosounder using a reference single-beam system, on
746 natural seafloor, in Lamarche, G., and Lurton, X. (Eds) *Seafloor backscatter data from swath*
747 *mapping echosounders: From technological development to novel applications*, *Marine*
748 *Geophysical Research* (in press)

749 Fonseca L, Calder B, and Wetzler M (2006) Experiments for Multibeam Backscatter adjustments
750 on the NOAA ship Fairweather. In *IEEE/OCEANS 2006*. September 18-21, Boston, MA, USA
751 <https://doi.org/10.1109/OCEANS.2006.307085>

752 Fonseca L, Mayer L (2007) Remote estimation of surficial seafloor properties through the
753 application of Angular Range Analysis to multibeam sonar data. *Marine Geophysical*
754 *Researches*, 28(2), 119-126 <https://doi.org/10.1007/s11001-007-9019-4>

755 Fonseca L, Brown C, Calder B, Mayer L, Rzhhanov Y (2009) Angular range analysis of acoustic
756 themes from Stanton Banks Ireland: A link between visual interpretation and multibeam
757 echosounder angular signatures. *Applied Acoustics*, 70(10): 1298-1304
758 <https://doi.org/10.1016/j.apacoust.2008.09.008>

759 Foote KG, Chu D, Hammar TR, Baldwin KC, Mayer LA, Hufnagle Jr LC, Jech JM (2005)
760 Protocols for calibrating multibeam sonar. *The Journal of the Acoustical Society of America*,
761 117(4): 2013-2027 <https://doi.org/10.1121/1.1869073>

762 Francois RE, Garrison GR (1982) Sound absorption based on ocean measurements. Part II: Boric
763 acid contribution and equation for total absorption. *The Journal of the Acoustical Society of*
764 *America*, 72(6): 1879-1890 <https://doi.org/10.1121/1.388673>

765 Fusillo L, De Moustier C, Satriano JH, Zietz S (1996) In-situ far-field calibration of multibeam
766 sonar arrays for precise backscatter imagery. In *OCEANS MTS/IEEE September 23-26*. Fort
767 *Lauderdale FL USA* <https://doi.org/10.1109/OCEANS.1996.566716>

768 Gallaudet TC (2001) *Shallow Water Acoustic Backscatter and Reverberation Measurements*
769 *using a 68-kHz Cylindrical Array*. PhD Dissertation. University of California, San Diego. Chapter

770 3, Using Environmental Information to correct for errors in Bathymetry and Seafloor Acoustic
771 Backscattering Strength Imagery

772 Gavrilov AN, Parnum IM (2010) Fluctuations of seafloor backscatter data from multibeam sonar
773 systems. IEEE Journal of Oceanic Engineering, 35(2): 209-219
774 <https://doi.org/10.1109/JOE.2010.2041262>

775 Greenaway SF, Weber TC (2010) Test methodology for evaluation of linearity of multibeam
776 echosounder backscatter performance. In OCEANS 2010 September 20-23. Seattle WA USA
777 <https://doi.org/10.1109/OCEANS.2010.5664383>

778 Hammerstad E (2000) EM technical note: Backscattering and seabed image reflectivity.
779 Kongsberg Maritime AS, Horten Available online at:
780 [https://www.km.kongsberg.com/ks/web/nokbg0397.nsf/AllWeb/C2AE0703809C1FA5C1257B5](https://www.km.kongsberg.com/ks/web/nokbg0397.nsf/AllWeb/C2AE0703809C1FA5C1257B580044DD83/$file/EM_technical_note_web_BackscatteringSeabedImageReflectivity.pdf)
781 [80044DD83/\\$file/EM technical note web BackscatteringSeabedImageReflectivity.pdf](https://www.km.kongsberg.com/ks/web/nokbg0397.nsf/AllWeb/C2AE0703809C1FA5C1257B580044DD83/$file/EM_technical_note_web_BackscatteringSeabedImageReflectivity.pdf) Last
782 accessed January 2018

783 Hare R, Godin A, Mayer LA (1995) Accuracy estimation of Canadian swath (multibeam) and
784 sweep (multitransducer) sounding systems. Technical report, Canadian Hydrographic Service
785 and University of New Brunswick, Fredericton

786 Hare R (2001) Error budget analysis for US Naval Oceanographic Office (NAVOCEANO)
787 hydrographic survey systems. University of Southern Mississippi, Hydrographic Science
788 Research Center for the Naval Oceanographic Office

789 Hasan RC, Lerodiconou D, Laurenson L, Schimel A (2014) Integrating multibeam backscatter
790 angular response, mosaic and bathymetry data for benthic habitat mapping. PLOS one 9(5):
791 e97339 <https://doi.org/10.1371/journal.pone.0097339>

792 Hauser O, Downs R, Rice G, Greenaway S, Annis M, Eisenberg J, Malik M (2015) NOAA's
793 Multibeam Sonar Test Procedure Manual: Formalizing and documenting a procedure to ensure
794 that a new multibeam sonar is properly installed, integrated, and capable of meeting
795 hydrographic standards. US Hydrographic Conference, March 16-19, Washington DC USA

796 Heaton J, Weber T, Rice G, Lurton X (2013) Testing of an extended target for use in high
797 frequency sonar calibration. Proceedings of Meetings on Acoustics 19(1)
798 <https://doi.org/10.1121/1.4800927>

799 Hellequin, L, Boucher JM, Lurton X (2003) Processing of high-frequency multibeam echo
800 sounder data for seafloor characterization. IEEE Journal of Oceanic Engineering, 28(1): 78-89
801 <https://doi.org/10.1109/JOE.2002.808205>

802 Hiroji A (2016) Extracting sonar relative along-track and Across-track radiometric beam pattern
803 for multi-sector multi-swath multibeam sonars, PhD dissertation. The University of New
804 Brunswick

805 Hughes Clarke JE, Mayer LA, Wells DE (1996) Shallow-water imaging multibeam sonars: a
806 new tool for investigating seafloor processes in the coastal zone and on the continental
807 shelf. *Marine Geophysical Researches*, 18(6): 607-629 <https://doi.org/10.1007/BF00313877>

808 Hughes Clarke JE, Danforth BW, Valentine P (1997) Areal seabed classification using
809 backscatter angular response at 95 kHz. In NATO SACLANTCEN Undersea Research Centre
810 Conference on High Frequency Acoustics in Shallow Water (pp 243-250) June 30 – July 4,
811 Lerici, Italy

812 Hughes Clarke JE (2012) Optimal use of multibeam technology in the study of shelf
813 morphodynamics, in *Sediments, Morphology and Sedimentary Processes on Continental*
814 *Shelves: Advances in Technologies, Research and Applications* (eds Li MZ, Sherwood CR, and
815 Hill PR), John Wiley & Sons, Ltd, Chichester, West Sussex, UK
816 <http://doi.org/10.1002/9781118311172.ch1>

817 IHO (International Hydrographic Organization) (2008) IHO Standards for Hydrographic Surveys
818 Special publication no 44, 5th edition www.iho.int/iho_pubs/standard/S-44_5E.pdf Last accessed
819 June 2017

820 Jackson D, Richardson M (2007) *High-frequency seafloor acoustics*. New York NY USA:
821 Springer Science & Business Media <https://doi.org/10.1007/978-0-387-36945-7>

822 Johannesson KA, Mitson RB (1983) *Fisheries acoustics. A practical manual for aquatic*
823 *biomass estimation*. FAO Fish. Tech. Pap. vol.240. 249 pp.

824 Kim YS, Moore RK, Onstott RG (1982) *Scattering Coefficient Estimation: An Examination of*
825 *the Narrow-Beam Approximation* (No RSL-TR-331-23). Kansas University Remote Sensing Lab
826 /Center for Research Inc. Technical Report, August 38 pp

827 Lacharité M, Brown CJ, Gazzola V (2017) Multisource multibeam backscatter data:
828 developing a strategy for the production of benthic habitat maps using semi-automated seafloor
829 classification methods, in Lamarche, G., and Lurton, X. (Eds) *Seafloor backscatter data from*
830 *swath mapping echosounders: From technological development to novel applications*, *Marine*
831 *Geophysical Research* <https://doi.org/10.1007/s11001-017-9331-6>

832 Ladroit Y, Sintès C, Lurton X, Garello R (2012) Extended scatterers model for fast sonar signal
833 simulation. In *IEEE Oceans Conference*, May 21-24 Yeosu, Republic of Korea
834 <https://doi.org/10.1109/OCEANS-Yeosu.2012.6263487>

835 Ladroit Y, Lamarche G, Pallentin A (2018) Seafloor multibeam backscatter calibration
836 experiment: comparing 45°-tilted 38-kHz split-beam echosounder and 30-kHz multibeam data,
837 in Lamarche G, and Lurton, X. (Eds) Seafloor backscatter data from swath mapping
838 echosounders: From technological development to novel applications, Marine Geophysical
839 Research, <https://doi.org/10.1007/s11001-017-9340-5>

840 Lamarche G, Lurton X, Verdier A L, Augustin J M (2011) Quantitative characterisation of
841 seafloor substrate and bedforms using advanced processing of multibeam backscatter—
842 Application to Cook Strait, New Zealand. Continental Shelf Research, 31(2):S93-S109
843 <https://doi.org/10.1016/j.csr.2010.06.001>

844 Lamarche G, Lurton X (2017) Recommendations for improved and coherent acquisition and
845 processing of backscatter data from seafloor-mapping sonars, in Lamarche, G., and Lurton, X.
846 (Eds) Seafloor backscatter data from swath mapping echosounders: From technological
847 development to novel applications, Marine Geophysical Research
848 <https://doi.org/10.1007/s11001-017-9315-6>

849 Lanzoni C, Weber T (2011) A Method for Field Calibration of a Multibeam Echo Sounder. In
850 Oceans MTS/IEEE September 19-22 Kona HI, USA
851 <https://doi.org/10.23919/OCEANS.2011.6107075>

852 Lehaitre, M, Delauney L, Compère C (2008) Biofouling and underwater measurements, in Babin
853 M, Roesler CS, Cullen JJ (Eds) Real-time observation systems for ecosystem dynamics and
854 harmful algal blooms: Theory, instrumentation and modelling. Oceanographic Methodology
855 Series. UNESCO, Paris, 463-493

856 Llewellyn, KC (2006) Corrections for beam pattern residuals in backscatter imagery from the
857 Kongsberg Simrad EM300 Multibeam Echosounder. MS Thesis. University of New Brunswick
858 Ocean Mapping Group

859 Long DG, Skouson GB (1996) Calibration of spaceborne scatterometers using tropical rain
860 forests. IEEE Transactions on Geoscience and Remote Sensing, 34(2): 413-424
861 <https://doi.org/10.1109/36.485119>

862 Lucieer V, Huang Z, Siwabessy J (2015) Analysing Uncertainty in Multibeam Bathymetric Data
863 and the Impact on Derived Seafloor Attributes, Marine Geodesy 39(1): 32-52
864 <https://doi.org/10.1080/01490419.2015.1121173>

865 Lucieer, V., Roche, M., Degrendele, K., Malik, M., Dolan, M. and Lamarche, G. (2017) User
866 expectations for multibeam echo sounders backscatter strength data - Looking back into the
867 future, in Lamarche, G., and Lurton, X. (Eds) Seafloor backscatter data from swath mapping
868 echosounders: From technological development to novel applications, Marine Geophysical
869 Research <https://doi.org/10.1007/s11001-017-9316-5>

870• Lurton X (2010) An Introduction To Underwater Acoustics – Principles and Applications,
871 Second Edition, Springer-Verlag, Berlin ISBN 978-3-540-78480-7
872

873 Lurton X, Augustin JM (2010) A Measurement Quality Factor for Swath Bathymetry Sounders.
874 IEEE Journal Of Oceanic Engineering, 35(4): 852-862
875 <https://doi.org/10.1109/JOE.2010.2064391>

876 Lurton X, Lamarche G (eds) (2015) Backscatter measurements by seafloor-mapping sonars.
877 Guidelines and Recommendations. 200p. [http://geohab.org/wp-](http://geohab.org/wp-content/uploads/2014/05/BSWGREPORT-MAY2015.pdf)
878 [content/uploads/2014/05/BSWGREPORT-MAY2015.pdf](http://geohab.org/wp-content/uploads/2014/05/BSWGREPORT-MAY2015.pdf) Last accessed June 2017

879 Lurton X, Eleftherakis D, Augustin JM (2017) Analysis of seafloor backscatter strength
880 dependence on the survey azimuth using multibeam echosounder data, in Lamarche, G., and
881 Lurton, X. (Eds) Seafloor backscatter data from swath mapping echosounders: From
882 technological development to novel applications, Marine Geophysical Research
883 <https://doi.org/10.1007/s11001-017-9318-3>

884 Matsumoto H, Dziak RP, Fox CG (1993) Estimation of seafloor microtopographic roughness
885 through modeling of acoustic backscatter data recorded by multibeam sonar systems. The
886 Journal of the Acoustical Society of America, 94(5): 2776-2787
887 <https://doi.org/10.1121/1.407361>

888 Mayer LA (2006) Frontiers in seafloor mapping and visualization. Marine Geophysical
889 Researches, 27(1): 7-17 <https://doi.org/10.1007/s11001-005-0267-x>

890 Mandell J (1964) The statistical analysis of experimental data. Dover Publications, Inc., New
891 York, USA

892 Parnum IM, Gavrilov AN (2011) High-frequency multibeam echo-sounder measurements of
893 seafloor backscatter in shallow water: Part 1–Data acquisition and processing. Underwater
894 Technology, 30(1): 3-12 <https://doi.org/10.3723/ut.30.003>

895 Peritsky MM (1973) Statistical estimation of mean signal strength in a Rayleigh fading
896 environment. IEEE transactions on communication, 21(11): 1207-1213
897 <https://doi.org/10.1109/TCOM.1973.1091577>

898 Prigent C, Aires F, Jimenez C, Papa F, Roger J (2015) Multiangle backscattering observations of
899 continental surfaces in Ku-Band (13 GHz) from satellites: understanding the signals, particularly
900 in arid regions. IEEE Transactions on Geoscience and Remote Sensing, 53(3): 1364-1373
901 <https://doi.org/10.1109/TGRS.2014.2338913>

902 QPS Technical note (2014) Technical note 1: Correcting backscatter for seafloor 3D incidence.
903 Available online at:

904 <https://confluence.qps.nl/display/KBE/Technical+Note+1%3A+Correcting+backscatter+for+seafloor+3D+incidence> Last accessed June 2017
905

906 Reed T, Hussong D (1989) Digital image processing techniques for enhancement and
907 classification of SeaMARC II side scan sonar imagery. Journal of Geophysical Research 94(B6):
908 7469-90 <http://doi.org/10.1029/JB094iB06p07469>

909 Rice G, Greenaway S, Weber T, Beaudoin J (2012) Methods for Collecting and Using
910 Backscatter Field Calibration Information for the Reson 7000 Series Multibeam. Center for
911 Coastal and Ocean Mapping. Paper 840. <http://scholars.unh.edu/ccom/840> Last accessed June
912 2017

913 Rice G, Cooper R, Degrendele K, Gutierrez F, Le Bouffant N, Roche M (2015) Chapter 5 -
914 Acquisition: best practice guide. In: Lurton X, Lamarche G (eds) Backscatter measurements by
915 seafloor-mapping sonars - Guidelines and Recommendations. Geohab report, 79-132
916 <http://geohab.org/wp-content/uploads/2013/02/BWSG-REPORT-MAY2015.pdf> Last accessed
917 June 2017

918 Rice G, Malik M (2015) NOAA Ship Nancy Foster EM 710 Acceptance Testing with
919 Hydrographic Systems and Technology Programs Multibeam sonar acceptance procedures, 9-13
920 March http://mac.unols.org/sites/mac.unols.org/files/NF_EM710_acceptance.pdf Last accessed
921 June 2017

922 Rzhanov Y, Fonseca L, Mayer L (2012) Construction of seafloor thematic maps from multibeam
923 acoustic backscatter angular response data. Computers & Geosciences, 41: 181-187
924 <https://doi.org/10.1016/j.cageo.2011.09.001>

925 Schimel A, Beaudoin J, Gaillot A, Keith G, Le Bas T, Parnum I, Schmidt V (2015) Chap 6:
926 Processing backscatter data: from datagrams to Angular responses and mosaics. In: Lurton X,
927 Lamarche G (eds) Backscatter measurements by seafloor-mapping sonars - Guidelines and
928 Recommendations. Geohab report, 133-164 [http://geohab.org/wp-](http://geohab.org/wp-content/uploads/2013/02/BWSG-REPORT-MAY2015.pdf)
929 [content/uploads/2013/02/BWSG-REPORT-MAY2015.pdf](http://geohab.org/wp-content/uploads/2013/02/BWSG-REPORT-MAY2015.pdf) Last accessed June 2017

930 Sea-Bird Electronics Inc. (2010) Application note no 6. Available online
931 <http://www.seabird.com/document/an06-determination-sound-velocity-ctd-data>. Last accessed:
932 February 2017

933 Simons DG, Snellen M (2009) A Bayesian approach to seafloor classification using multi-beam
934 echo-sounder backscatter data. Applied Acoustics, 70(10): 1258-1268
935 <https://doi.org/10.1016/j.apacoust.2008.07.013>

936 Stanic S, Kennedy E (1992) Fluctuations of high-frequency shallow-water seafloor
937 reverberation. The Journal of the Acoustical Society of America, 91(4): 1967-1973
938 <https://doi.org/10.1121/1.403680>

- 939 Teng, Y (2011) Sector-specific beam pattern compensation for multi-sector and multi-swath
 940 multibeam sonars. MScE Thesis, Department of Geodesy and Geomatics Engineering,
 941 University of New Brunswick
- 942 Ulaby FT, Allen CT, Fung AK (1983) Method for retrieving the true backscattering coefficient
 943 from measurements with a real antenna. IEEE Transactions on Geoscience and Remote Sensing,
 944 GE-21(3): 308-313 <https://doi.org/10.1109/TGRS.1983.350558>
- 945 Urick RI (1983) Principles of underwater acoustics. McGraw-Hill, New York ISBN 0932146627
- 946 Wang Q, Gogineni SP (1991) A numerical procedure for recovering true scattering coefficients
 947 from measurements with wide-beam antennas. IEEE Transactions on Geoscience and Remote
 948 Sensing, 29(5): 778-783 <https://doi.org/10.1109/36.83993>
- 949 Welton B (2014) A Field Method for Backscatter Calibration Applied to NOAA's Reson 7125
 950 Multibeam Echo-Sounders. MS Thesis. University of New Hampshire, Durham, NH, US
- 951 Zhu SJ, Tang GA, Xiong LY, Zhang G (2014) Uncertainty of slope length derived from digital
 952 elevation models of the Loess Plateau, China. Journal of Mountain Science 11(5): 1169-1181
 953 <https://doi.org/10.1007/s11629-013-2788-0>

954 APPENDIX A

955 Statistical uncertainty in measured *EL*

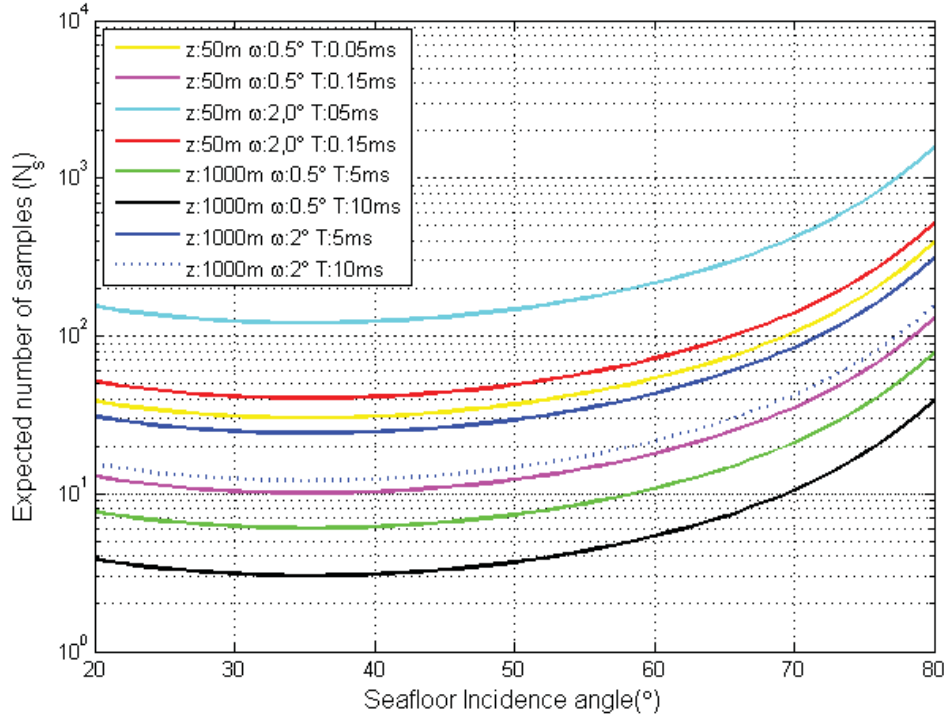
956 The statistical fluctuation of the *EL* is an inherent property of backscattered signals and therefore
 957 an unavoidable source of random uncertainty. However, confidence in the mean echo level
 958 reliability can be improved by increasing the number of samples used in averaging. In MBES
 959 data, this is done most often by averaging across-track and along-track samples. However, this
 960 should only be done for homogeneous seafloor as the mean angular response can be corrupted at
 961 the transition between two seafloor types. Mosaic segmentation into areas showing similar
 962 backscatter can help in selecting regions of the same seafloor type over which the samples can be
 963 averaged (Rzhanov et al. 2012). The number of samples available for each beam is controlled by
 964 the across-track footprint extent, so the largest number of samples is obtained for the outer-most
 965 beams. Assuming that the time series is being sampled at a high enough rate compared with the
 966 pulse duration, the number of statistically-independent samples N_s inside a beam is computed as
 967 the ratio of the length of the receive beam footprint in the across-track direction and the projected
 968 pulse duration (Simons and Snellen 2009):

$$N_s(\theta) \approx \left(\frac{z\omega}{\cos^2 \theta} \right) / \left(\frac{cT}{2 \sin \theta} \right) \quad [\text{Eq. A1}]$$

969 where z is the water depth, ω the R_x across-track beamwidth, c the sound speed, T the pulse
 970 length and θ the incidence angle. Eq. [A1] holds for long-pulse regime, excluding the angles

971 around nadir. Obviously, the benefit of averaging over several samples exists only when $N_s > 1$.
 972 Fig. A1 presents the number of statistically independent samples for a MBES with $\omega = 0.5^\circ$ and
 973 2° ; and $z = 50$ m (with $T = 0.05$ ms and 0.15 ms) and 1000 m (with $T = 5$ ms and 10 ms). N_s
 974 increases with decreasing T and increasing ω .

975



976

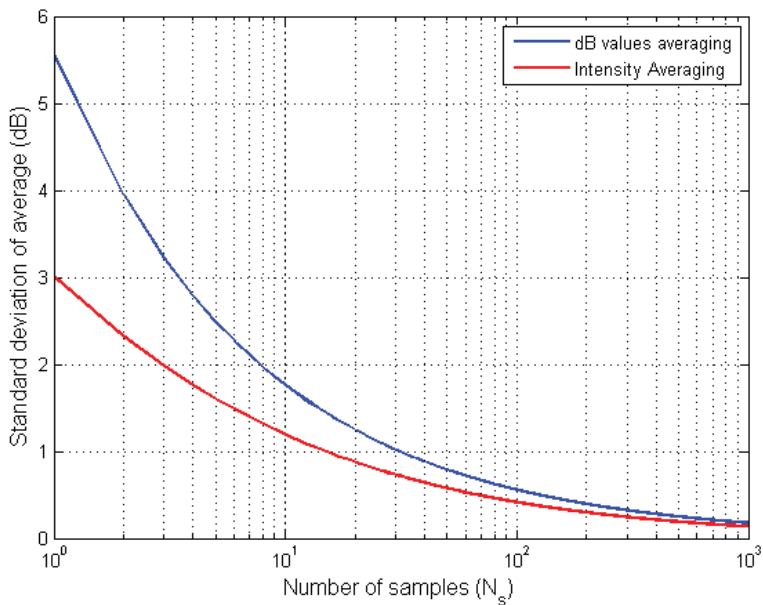
977 Figure A1: Estimated number [Eq. A1] of statistically independent samples for each beam for a multibeam
 978 echosounder at water depths 50 m and 1000 m; beamwidths of 0.5° and 2° ; and pulse lengths (0.15; 0.5;
 979 5 and 10 ms).

980 The standard deviation of N averaged independent samples is given as:

$$\sigma_{\bar{x}} = \frac{\sigma_x}{\sqrt{N}} \quad [\text{Eq. A2}]$$

981 where $\sigma_{\bar{x}}$ and σ_x are the standard deviations of averaged and individual samples respectively. Eq.
 982 [A2] is valid provided that the N averaged values are statistically independent, are derived from a
 983 same population, and have the same variance (Mandell 1964). Assuming the standard deviation
 984 of individual samples is 5.57 dB (Rayleigh distribution) and averaging over the dB values, more
 985 than 30 individual samples are required to achieve a 1 dB standard deviation (Fig. A2). If the
 986 envelope squared amplitudes (i.e. intensity) in natural units is considered for the averaging
 987 (which is a preferable way to do it), the dB value of the standard deviation referenced to the

988 mean is $10 \log_{10}(1 + 1/\sqrt{N}) \approx 4.34/\sqrt{N}$ dB (Bjørnø 2017 p. 527). In this case, to reduce the
 989 standard deviation to 1 dB, only ~ 20 samples are required (Fig. A2). Although the uncertainty is
 990 lowered by averaging over larger number of samples, the spatial resolution is adversely affected
 991 which may or may not be important depending on the type of application (compare high
 992 resolution mapping, with large scale mapping).



993
 994 Figure A2: Estimated number of statistically independent samples to be averaged in order to obtain a given
 995 standard deviation (in dB). The initial distribution is Rayleigh, with a standard deviation of 5.57 dB.