Global scale measurement of Ocean Bubble Depth with Space Lidar.

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Visible and microwave satellite measurements can provide the global whitecap fraction. The bubble clouds are three-dimensional structures, and a space-based lidar can provide complementary observations of the bubble depth. Here, we use lidar measurements of the Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observation (CALIPSO) satellite to quantify global bubble depth from the depolarization. The relationship between CALIPSO bubble depth and wind speed from the Advanced Microwave Scanning Radiometer for EOS (AMSR-E) and AMSR2 is similar to a recently derived relationship based on buoy measurements. The CALIPSO-based bubble depth data show global distributions and seasonal variations consistent with the high wind speed (> 7 m/s) but with some variance. We also found similarities between the CALIPSO bubble depth and the whitecap fraction from AMSR2 and WindSat. Our findings support the use of spaceborne lidar measurements for advancing the understanding of the 3D bubble properties, and the ocean physics at high wind speeds.

1. Introduction

Measurements of the whitecap fraction W—defined as the fraction of the ocean surface covered by whitecaps (sea foam)—are usually extracted from photographs and video images collected from ships and aircraft (Monahan, 1971; Asher and Wanninkhof, 1998; Callaghan and White, 2009; Kleiss and Melville, 2011). More recently, the measurement of whitecaps fraction using passive remote sensing has been demonstrated (Anguelova and Bettenhausen, 2019).

The whitecap fraction W quantifies wave breaking with air entrainment. Therefore, W is a suitable forcing parameter for developing parameterizations of surface fluxes in models that represent the enhancement of the air-sea interactions due to breaking waves, including heat transfer (Andreas et al., 2015), gas exchange (Wanninkhof et al., 2009), and sea spray production (Veron, 2015). Large-scale wave breaking affects the wave field evolution; thus, its realistic and accurate representation in wave models is a necessity. In this sense, the capability to measure W from space is a breakthrough for global observations of air-sea processes.

While oceanic whitecaps are the most visible and direct surface expression of wave breaking, they also mark areas of bubble plumes formed beneath the breaking waves. The bubble plumes are three-dimensional structures with properties described with a range of variables such as void fraction, size distributions, penetration depth, surface area, and bubble plume volumes, often averaged over many wave periods. This complexity implies that quantifying all the statistics of these bubble plumes is essential to obtain robust parameterizations of the fluxes at the ocean-atmosphere interface. Recent studies attest to this by showing that the statistics of bubble plumes are strongly correlated with total wave-breaking dissipation (Schwendeman and Thomson, 2015; Callaghan et al., 2016; Callaghan, 2018; Derakhti et al., 2020, 2023). Even if the whitecap fraction and the bubble depths are related (Derakhti et al. 2023), the considerable variation of sea states and breaking wave scales requires additional research to validate such relationships across

48 the global ocean (Callaghan 2018). For this reason, the capability to determine the bubble depth
49 from space combined with the existing capability to determine the whitecap fraction would be a
50 breakthrough for global physical oceanography. It would allow us to confirm the existence of a
51 general whitecaps-to-bubble depth relationship and, in any case, to measure the energy
52 dissipation rate associated with breaking waves (Callaghan 2018).
53
54 Lidar systems have been deployed on ships and aircrafts to characterize ocean optical properties,
55 phytoplankton, zooplankton, and fish stocks (Dickey et al., 2011; Hoge et al., 1988; Churnside et
56 al., 2001; Churnside and Thorne, 2005; Reese et al., 2011). As active sensors, lidar
57 measurements have distinct advantages over passive remote sensing for ocean observations in
58 that they can provide information on the vertical structure of ocean properties.
59
60 Regarding the spatial structure of bubble properties, lidar measurements cannot match the
61 coverage of the passive systems. However, space lidar can provide the average bubble depth,
62 which, in conjunction with passive measurements, provides independent assessments of the
63 bubble volume and the breaking wave energy dissipation rate.
64
65 Currently, there are no ocean space lidar systems. However, the National Aeronautics and Space
66 Administration (NASA) and the Centre National d'Etudes Spatiales launched the Cloud-Aerosol
67 Lidar and Infrared Pathfinder Satellite Observation (CALIPSO) satellite in 2006 as part of the A-
68 train Earth Observing Sensor suite (Winker et al., 2009). The primary instrument on CALIPSO is
69 the Cloud-Aerosol Lidar with Orthogonal Polarization (CALIOP) sensor, and it collected global
70 lidar measurements from 2006 to 2023. Because of its polarization characterization capabilities,
71 the CALIPSO mission offered a unique opportunity for the first global evaluation of bubble
72 properties from a space lidar. Here, we focus on retrieving the bubble depth using CALIOP's 532
73 nm polarization channels.
74
75 A lidar with depolarization can provide unique information on two bubble properties: the bubble
76 depth (Josset et al. 2024) and the void fraction (Churnside et al. 2010, Josset et al. 2024). We
77 focus here on the bubble depth because its link with the integrated depolarization is direct (see
78 addendum, Method M1).
79
80 The CALIPSO capability for water column profiling is limited because the lidar sensor is
81 designed for atmospheric research and has a coarse in-water vertical resolution of 22.4 m.
82 Therefore, our analysis focuses on the vertically integrated depolarization from 60 m above to
83 300 m below the ocean surface. As an initial validation of our approach, we compare CALIPSO-
84 based bubble depth data with previously derived wind speed relationships.
85
86 **2. Data and Methods**
87 2.1. CALIPSO lidar data Analysis
88 The addendum (Method M1, sections a and b) contains the details of the CALIPSO lidar data
89 analysis and the derivation of the associated uncertainties.
90
91 In support of developing satellite lidar retrievals of bubble depth, Naval Research Laboratory
92 (NRL) deployed a shipboard lidar system (Gould et al. 2019, Wang et al. 2022) during the
93 National Science Foundation (NSF) Breaking Bubbles deployment. For the shipboard lidar, the

94 track is well defined (Fig. 1, left), and the period is limited to the 2019 cruise in the Gulf of
95 Alaska (Breaking Bubbles) between December 4th (Dutch Harbor) and December 23rd (Seattle)
96 (Derakthi et al. 2023). For CALIPSO, to capture sufficient data statistics, we initially limited the
97 coverage to the Gulf of Alaska surrounding the ship track during December 2019 (pink rectangle
98 in Fig. 1, left).

99

100 The lidar system acquired 113 hours of ocean backscatter profiles during storm conditions,
101 providing high-resolution data to derive a relationship between the bubble depth and lidar
102 depolarization. This dataset can be, in turn, combined with the high-resolution co-polarized
103 observations to determine the bubble void fraction (Churnside et al. 2010, Josset et al. 2024).
104 The shipboard lidar high-resolution observations can also be integrated and thus provide the link
105 between the bubble depth and integrated depolarization observations. Any low-resolution lidar
106 system, such as the CALIPSO lidar, can use this relationship.

107

108 Assessment of ocean bubble properties from a lidar co-polarization channel is challenging in
109 actual ocean conditions (Josset et al. 2024) because of the difficulty in discriminating between
110 the bubble plume and the natural ocean background in this channel. The difficulty increases for a
111 space lidar because the signal-to-noise ratio and the vertical resolution are low.

112

113 As wind speed increases, however, the ocean signal measured by CALIPSO's integrated
114 depolarization is almost entirely due to backscatter from the underwater bubbles. Based on our
115 shipboard lidar measurements taken in the Gulf of Alaska, we found that there is a simple
116 statistical relationship between the shipboard lidar integrated depolarization and the bubble depth
117 derived directly from the lidar high-resolution profiling capabilities (Josset et al. 2024).

118

119 Therefore, we used the depolarization ratio, defined as the surface integrated cross-polarized
120 backscatter divided by the integrated co-polarized backscatter, to retrieve the bubble depth. This
121 ratio is independent of atmospheric transmittance and is very accurately calibrated. To account
122 for the lack of bubbles at low wind speeds, we included only a subset of the data in the analyses,
123 which was more representative of breaking wave conditions. Specifically, we removed the
124 retrievals under the conditions of very high surface depolarization (> 0.1, to remove sea ice), low
125 wind (< 7 m/s), bubble depth ≤ 0 m (so that the bubbleless ocean does not dominate the statistic),
126 and scenes with liquid water clouds, ice clouds, and stratospheric features (see Method M1).

127

128 The CALIPSO data are the Synergized Optical Depth of Aerosols (SODA) product on the
129 ICARE website (https://www.icare.univ-lille.fr/soda/). This product has pre-processed the
130 collocation between CALIPSO and the Advanced Microwave Scanning Radiometer (AMSR)
131 unified data, including AMSR for Earth Observing System (AMSR-E, 2002 to 2011) and
132 AMSR2 (2012 to present). This allowed the pairing of the CALIPSO surface-integrated
133 attenuated backscatter with wind speed retrievals from AMSR-E/AMSR2 measurements (Wentz
134 et al. 2014).

135

136 For the current analysis, and as described in Method M1, the relationship between bubble depth
137 and depolarization δ_R is:

138

139
$$\text{Bubble Depth} = -20.18\delta_R^{-0.255} + 58.47 \text{ m}$$

140

141 The relationship between depolarization and bubble depth depends on the amount of multiple
142 scattering. CALIPSO is keeping a higher amount of multiple scattered light than the shipboard
143 lidar, and the distribution of depolarization is shifted towards higher values (Method M1, Fig.
144 M.2). For the data in the Gulf of Alaska, the median depolarization of the shipboard observations
145 is 0.0196 ± 0.0054 and 0.0329 ± 0.0205 for CALIPSO.

146

147 The anticipated error budget is described in Method M1. This high depolarization distribution is
148 likely the cause of the high bias that we determined. The use of a rescaling factor for the
149 depolarization to bubble depth relationship could increase the accuracy of the algorithm.

150

151 2.2. Whitecap fraction satellite retrievals
152 Satellite retrievals of whitecap fraction W are based on the high emissivity of the sea foam at
153 microwave frequencies measured by radiometers in terms of brightness temperature T_B
154 (Anguelova and Bettenhausen, 2019). The whitecap retrieval algorithm utilizes a radiative
155 transfer model to simulate the emissivity of foam-free and foam-covered ocean areas at the top
156 of the atmosphere. Geophysical variables, including wind vector (speed U_{10} and direction ϕ), sea
157 surface temperature (T), columnar water vapor (V), and columnar cloud liquid water (L), serve as
158 inputs to the models comprising the whitecap algorithm. The input data (U_{10}, ϕ, T, V, L) can be
159 taken either from a geophysical model or from satellites. Comparison of different versions of the
160 whitecap retrieval algorithm have shown that the optimal approach is to use T_B and input
161 variables (U_{10}, ϕ, T, V, L) from the same spaceborne sensor to minimize errors due to time-space
162 collocations.

163

164 Physically based whitecap retrievals (as opposed to an initial empirical version) were first
165 developed within the framework of the WindSat mission (Gaiser et al., 2004; Bettenhausen et al.,
166 2006). Recently, the whitecap retrieval algorithm was adapted for use with AMSR2 observations
167 of T_B (Anguelova et al., 2022). Modifications were necessary because AMSR2 sensor tracks the
168 WindSat frequency channels (from 6 to 37 GHz) closely but differs in the incidence angles and
169 calibration.

170

171 The whitecap algorithm produces W retrievals at different microwave frequencies (e.g., 10, 18
172 and 37 GHz) and two polarizations (horizontal H and vertical V, equivalents of the co-
173 polarization and cross-polarization of lidar observations - Method M1). The W values at different
174 frequencies are useful approximations for the thickness of the whitecaps. For example, W at 10
175 GHz represent predominantly thick foam layers formed during the active wave breaking with
176 bubble plumes below featuring large penetration depths and wide range of bubble sizes. In
177 contrast, W at 37 GHz includes both thick active whitecaps and thin layers of decaying foam left
178 behind the breaking waves. The polarization differences can be used as a measure for the
179 whitecap sensitivity to wind speed forcing — stronger at H polarization and weaker for V
180 polarization.

181

182 In this study, we have used W retrievals from AMSR2 and WindSat for 5-22 December 2019
183 obtained with the respective whitecap algorithms for each sensor. WindSat T_B data are stored at
184 NRL, while AMSR2 T_B data were taken from the Japan Aerospace Exploration Agency (JAXA).
185 We have used AMSR2 Level-1R product, which spatially matches the center positions and

186 footprint sizes for each frequency channel (Maeda et al., 2015). Before running the whitecap
187 algorithm, the geophysical input data (U_{10}, ϕ, T, V, L) were retrieved for each sensor. The
188 WindSat W retrievals are for a footprint of approximately 25 km $\times$ 35 km; these were
189 subsequently grided at $1/4°\times1/4°$ grid cell. The AMSR2 W retrievals are used at their footprint
190 resolution of 14 km $\times$ 22 km (no gridding was applied). For both instruments, the W values used
191 are averages of the W retrievals at 18 GHz, H and V polarizations.

192
193

194 **3. Results and Discussion**
195 3.1. Comparison of the shipboard and space lidar bubble depth data in the Gulf of Alaska
196 We compared the occurrences of the lidar bubble depth retrievals (NRL shipboard lidar and
197 CALIPSO) as a function of wind speed (Fig. 1, c). Over the 19-day period of breaking bubble
198 field measurements over the whole Gulf of Alaska, the R/V Sikuliaq's track (Fig. 1, a and b)
199 transected stormy ocean environments. Thus, the shipboard bubble depth values range widely,
200 from 0.14 to 30.93 m, with a median of 1.68 m and a standard deviation of 2.84 m (Fig. 1, c). For
201 CALIPSO, the bubble depth range has a minimum of 1.08 m, a maximum of 22.17 m, a median
202 of 9.01 m, and a standard deviation of 5.68 m.

203

204 Fig. 1 shows that the bubble depth of CALIPSO increases as the wind increases, as expected.
205 However, the horizontal and vertical spatial resolutions of the ship-based measurements are far
206 finer than that achieved with an 80 m diameter CALIPSO footprint. In other words, the NRL
207 shipboard lidar provides several profiles of bubble depth for each bubble cloud. This statistically
208 increases the occurrence of data with a low bubble depth for any wind speed. The maximum or
209 average value of the bubble depth is the quantity increasing as a function of wind speed.
210 CALIPSO does not provide such high-resolution observations and measures one depth for each
211 laser shot for an 80 m diameter cylinder (Method M1, Fig. M.4). This quantity correlates more
212 directly with wind speed than high-resolution observations. Therefore, it is possible to obtain a
213 better correlation with wind speed from the shipboard lidar by averaging the bubble depth values
214 within a given time interval.

215

216 Even with the differences in observing geometries and resolution, interesting similarities exist
217 between the bubble depth statistic of the shipboard and spaceborne lidar data. Half of the bubble
218 depth statistic is below 9 m, and the number of bubble observations above 20 m decreases
219 sharply.

220

221 There is a positive bias of 4.5 m in the CALIPSO bubble depth. We describe the derivation of
222 this bias in Method M1. A bias correction would bring the CALIPSO data closer to the shipboard
223 lidar data for a bubble depth of around 15 m and above. However, this correction would increase
224 the difference for depths lower than 10 m. This inconsistency is an indication that the constant
225 bias derived in Method M1 could be a simplification of the uncertainty. A positive bias around
226 4.5 m for CALIPSO bubble depth above 15 m and a negative bias of the same value for depth
227 below 15 m would bring the two histograms closer. That being said, the observations of the
228 shipboard lidar do not cover as wide of a meteorological variability as CALIPSO. The shipboard
229 dataset is much more limited in time and space. The swell was extremely steep for some
230 observations of the shipboard lidar. This may change the relationship between wind speed and

231   bubble depth. Therefore, modifying the algorithm based solely on this comparison seems
232   premature.

233

234   Interestingly, the linear increase of CALIPSO bubble depth with wind speed is similar to the
235   relationship found by Vagle et al. (2010). This is notable given that the algorithm is only based
236   on depolarization and does not use the wind speed dependence explicitly.

237

238   3.2. Regional and global comparison of the space lidar bubble depth and wind speed
239   Figure 2a shows the CALIPSO bubble depth as a function of wind speed for all data collected in
240   the Gulf of Alaska during December 2019. We extended the dataset temporal scale by using data
241   from June 2006 to June 2020 (Fig. 2b). We then extended to the spatial scale by using data for
242   the same time period but for the global oceans (Fig. 2c). The observed relationship between the
243   bubble depth and wind speed in Fig. 2a seems to follow relatively well the various relationships
244   presented in Cifuentes-Lorenzen et al. (2023). The best fit is with the mean plume depth of
245   Derakhti et al. (2023), where there is a good agreement with the CALIPSO bubble depths above
246   25 m and a positive bias for the depths below 25 m. The variations in the bubble depth to wind
247   speed relationships used in our comparison are expected and can be related to the ocean's
248   complexity. That being said, even if the agreement with Derakhti et al. (2023) is slightly better
249   when we limit the data to the Gulf of Alaska and December 2019 (Fig. 2a), the trend for depths
250   lower than 25 m does not change much when we extend the temporal (Fig. 2b) or the spatial
251   scales (Fig. 2c); i.e., the bubble depth from CALIPSO stay biased high in all cases.

252

253   Results from these satellite evaluations demonstrate the capability of CALIPSO to quantitatively
254   derive the bubble depth values within the range of variability associated with echo sounders or
255   shipboard lidar data. The data are similar to those of Cifuentes-Lorenzen et al. (2023), even if
256   they are primarily biased high. Considering the novelty of the CALIPSO lidar bubble depth
257   retrieval, it would make sense that the difference with the observations of Derakhti et al. (2023)
258   comes from the inaccuracy of our first algorithm. However, the results are promising, and this
259   justifies additional discussion of the CALIPSO bubble depth data at different scales.

260

261

262   3.3. Comparison of CALIPSO bubble depth and AMSR2/WindSat whitecap fraction in the Gulf
263   of Alaska
264   Breaking waves create bubbles, and this typically correlates with higher wind speed. CALIPSO
265   depolarization increases as wind speed increases. Accordingly, the global open-ocean bubble
266   depth structures should be qualitatively similar to the wind speed and whitecaps fraction
267   measured from satellites.

268

269   In Fig. 3, we show the similarities between the whitecap fraction from AMSR2 and WindSat in
270   the Gulf of Alaska in December 2019 and the bubble depth of CALIPSO in the same area. As we
271   can see, the contours of the higher whitecap fraction (Fig. 3 c and d) show similarities with the
272   highest values of the bubble depth.

273

274   Both the bubble depth and whitecap fraction are lower in the eastern and southern parts of the
275   Gulf of Alaska. These variables increase in the northern and western parts of the Gulf. There are

no data for AMSR2 for the northwestern part of this map, but the higher bubble depth in this area corresponds well to the WindSat higher whitecap fraction.

3.4. Global CALIPSO bubble depth

At the global scale, we expect to observe strong seasonal cycles in bubble depth in the high latitudes (roaring sixties), where the higher wind speed significantly enhances breaking wave processes. This spatial and seasonal variability in wave breaking should be apparent in global patterns of bubble depth.

Combining all CALIPSO bubble depth data for June 2006–2020 analysis period yields a global climatology that exhibits all the anticipated major breaking wave features (Fig. 4). Elevated bubble depth values in the Southern Ocean reflect the region's high wind speed while lower average values are found in the equator and tropics for all seasons.

Patchy seasonal variations in bubble depth in the Greenland, Iceland, and the United Kingdom gap are also reflected in the CALIPSO bubble depth data and correspond to varying sources of breaking waves.

Climatologies of bubble depth data for the Boreal summer (June–August) (Fig. 4c and 4g) and Boreal winter (December–February) (Fig. 4a and 4e) further illustrate the strong seasonality of high-latitude bubble depth and, again, demonstrate the feasibility of characterizing underwater bubble depth and their variability with a space-based lidar.

4. Conclusions

The results presented here demonstrate the quantitative measurement of bubble depth with a space-based lidar. CALIPSO bubble depth retrievals allow independent assessments of missing energy and provide a globally comprehensive data set for algorithm development of bubble depth to whitecap fraction algorithm, thus addressing a key knowledge gap in ocean physics. The next step is to combine the space lidar bubble depth with AMSR2 whitecap fraction to directly measure the three-dimensional bubble properties globally.

Open Research

The SODA, AMSR-E/AMSR2 and CALIPSO data used in this research are publicly available (after registration to https://www.icare.univ-lille.fr/login/?prourl=/asd-content/archive/, go to https://www.icare.univ-lille.fr/data-access/data-archive-access/?dir=CALIOP/SODA-333m.v2.30/ and https://www.icare.univ-lille.fr/data-access/data-archive-access/?dir=CALIOP/SODA-333m.v2.31/). AMSR2 brightness temperature data used for whitecap fraction retrievals were downloaded from JAXA (https://www.eorc.jaxa.jp/AMSR/datacatalog/tb/). All the other data used in this study are archived at (https://data.mendeley.com/datasets/scjcjyvjhn/1 , DOI: 10.17632/scjcjyvjhn.1).

References

Andreas, E. L, Mahrt, L., & Vickers, D. (2015). An improved bulk air–sea surface flux algorithm, including spray-mediated transfer. Quarterly Journal of the Royal Meteorological Society, 141(687), 642–654. https://doi.org/10.1002/qj.2424

Anguelova, M. D., & Bettenhausen, M. H. (2019). Whitecap fraction from satellite measurements: Algorithm description. *Journal of Geophysical Research: Oceans*, 124, 1827–1857. https://doi.org/10.1029/2018JC014630.

Anguelova, M. D., M. H. Bettenhausen, E. J. Hyer, and C. P. Camacho (2022). Demonstrate the Use of Satellite Whitecap Fraction Retrievals for Air-Sea Interaction Products in Navy Operational Models. NRL Memorandum Report (https://apps.dtic.mil/sti/citations/AD1184122)

Asher, W. E., and R. Wanninkhof (1998), The effect of bubble-mediated gas transfer on purposeful dual gaseous tracer experiments, *J. Geophys. Res*., 103(C5), 10,555–10,560, doi:10.1029/98JC00245.

Bettenhausen, M. H., Smith, C. K., Bevilacqua, R. M., Wang, N.-Y., Gaiser, P. W., & Cox, S. (2006). A nonlinear optimization algorithm for WindSat wind vector retrievals. *Transactions on Geoscience and Remote Sensing*, 44(3), 597–610. https://doi.org/10.1109/TGRS.2005.862504

Callaghan, A. H., & White, M. (2009). Automated processing of sea surface images for the determination of whitecap coverage. *Journal Atmospheric and Oceanic Technology*, 26(2), 383–394.

Callaghan, A. H., G. B. Deane, and M. D. Stokes (2016). Laboratory air-entraining breaking waves: Imaging visible foam signatures to estimate energy dissipation. *Geophys. Res. Let*., 43, 11–320, doi:10.1002/2016GL071226.

Callaghan, A. H. (2018). On the relationship between the energy dissipation rate of surface877 breaking waves and oceanic whitecap coverage, *J. Phys. Oceanogr*., 48, 2609–2626, doi:10.1175/JPO-D-17-0124.1.

Churnside, J. H., and R. E. Thorne (2005). Comparison of airborne lidar measurements with 420 kHz echo-sounder measurements of zooplankton, *Appl. Opt.*, 44, 5504–5511.

Churnside, J. H., K. Sawada, and T. Okumura (2001). A comparison of airborne lidar and echo sounder performance in fisheries. *J. Mar. Acoust. Soc. Jpn.*, 28, 175–183.

Churnside, J. H. (2010). Lidar signature from bubbles in the sea. *OPTICS EXPRESS* Vol. 18, No. 8/

Cifuentes-Lorenzen, A., Zappa, C. J., Randolph, K., & Edson, J. B. (2023). Scaling the bubble penetration depth in the ocean. *Journal of Geophysical Research: Oceans*, 128, e2022JC019582. https://doi.org/10.1029/2022JC019582

Derakhti, M., J. Thomson, and J. T. Kirby (2020). Sparse sampling of intermittent turbulence generated by breaking surface waves. *J Phys. Oceanogr.*, 50(4), 867–885, doi:913 10.1175/JPO-D-19-0138.1.

Gaiser, P. W., St Germain, K. M., Twarog, E. M., Poe, G. A., Purdy, W., Richardson, D., et al. (2004). The WindSat spaceborne polarimetric microwave radiometer: Sensor description and early orbit performance. *Transactions on Geoscience and Remote Sensing*, 42(11), 2347–2361. https://doi.org/10.1109/TGRS.2004.836867

Maeda, T., Y. Taniguchi and K. Imaoka, (2015). GCOM-W1 AMSR2 Level 1R Product: Dataset of Brightness Temperature Modified Using the Antenna Pattern Matching Technique. *Transactions on Geoscience and Remote Sensing*, vol. 54, no. 2, pp. 770-782, Feb. 2016, doi: 10.1109/TGRS.2015.2465170.

Morteza, Derakhti, Jim Thomson, Christopher S Bassett, et al. (2023). Statistics of bubble plumes generated by breaking surface waves. *ESS Open Archive.* February 27,. DOI: 10.22541/essoar.167751591.11265648/v1

Dickey, T. D., G. W. Kattawar, and K. J. Voss (2011). Shedding new light on light in the ocean. *Phys. Today*, 64, 44–49.

Gould, Jr., R. W., D. Josset, S. Anderson, W. Goode, R. N. Conmy, B. Schaeffer, S. Pearce, T. Mudge, J. Bartlett, D. Lemon, D. Billenness, O. Garcia (2019). Estimating Oil Slick Thickness with LiDAR Remote Sensing Technology. Bureau of Safety and Environmental Enforcement (BSEE) Oil Spill Response Research Branch ; https://www.bsee.gov/sites/bsee.gov/files/research-reports//1091aa.pdf.

Hoge, F. E., C. W. Wright, W. B. Krabill, R. R. Buntzen, G. D. Gilbert, R. N. Swift, J. K. Yungel, and R. E. Berry (1988). Airborne lidar detection of subsurface oceanic scattering layers. *Appl. Opt.*, 27, 3969–3977.

Josset, D., S. Cayula, B. Concannon, S. Sova, A. Weidemann. On the bubble-bubbleless ocean continuum and its meaning for the lidar equation: Lidar measurement of underwater bubble properties during storm conditions. Published at https://arxiv.org/abs/2404.10033.

Kleiss, J. M., and W. K. Melville (2011). The analysis of sea surface imagery for whitecap kinematics, *Journal of Atmospheric and Oceanic Technology*, 28(2), 219–243.

Monahan, E. C., (1971). Oceanic whitecaps. *J. Phys. Oceanogr.*, 1, 139–144, https://doi.org/10.1175/1520-0485(1971)001,0139:OW.2.0.CO;2.

Reese, D. C., R. T. O'Malley, R. D. Brodeur, and J. H. Churnside (2011). Epipelagic fish distributions in relation to thermal fronts in a coastal upwelling system using high-resolution remote-sensing techniques. ICES *J. Mar. Sci.*, 68, 1865–1874.

Schwendeman, M., and J. Thomson (2015). Observations of whitecap coverage and the relation to wind stress, wave slope, and turbulent dissipation. *J. Geophys. Res.: Oceans*, 120, 8346–8363, doi:10.1002/2015JC011196.

Veron, F. (2015). Ocean spray. Annual Review of Fluid Mechanics, 47(1), 507–538. https://doi.org/10.1146/annurev-fluid-010814-014651.

Wang, D.; Josset, D.; Savelyev, I.; Anguelova, M.; Cayula, S.; Abelev, A. (2022) An Experimental Study on Measuring Breaking-Wave Bubbles with LiDAR Remote Sensing. *Remote Sens.*, *14*, 1680. https://doi.org/10.3390/rs14071680.

Wanninkhof, R., Asher, W. E., Ho, D. T., Sweeney, C., & McGillis, W. R. (2009). Advances in quantifying air-sea gas exchange and environmental forcing. Annual Review of Marine Science, 1(1), 213–244. https://doi.org/10.1146/annurev.marine.010908.163742

Wentz, F.J., T. Meissner, C. Gentemann, K.A. Hilburn, J. Scott, (2014). Remote Sensing Systems GCOM-W1 AMSR2 Daily Environmental Suite on 0.25 deg grid, Version V.v, [indicate subset if used]. Remote Sensing Systems, Santa Rosa, CA. Available online at www.remss.com/missions/amsr.

Winker, D. M., M. A. Vaughan, A. Omar, Y. X. Hu, K. A. Powell, Z. Y. Liu, W. H. Hunt, and S. A. Young (2009). Overview of the CALIPSO Mission and CALIOP data processing algorithms. *J. Atmos. Oceanic Technol.*, 26, 2310–2323.
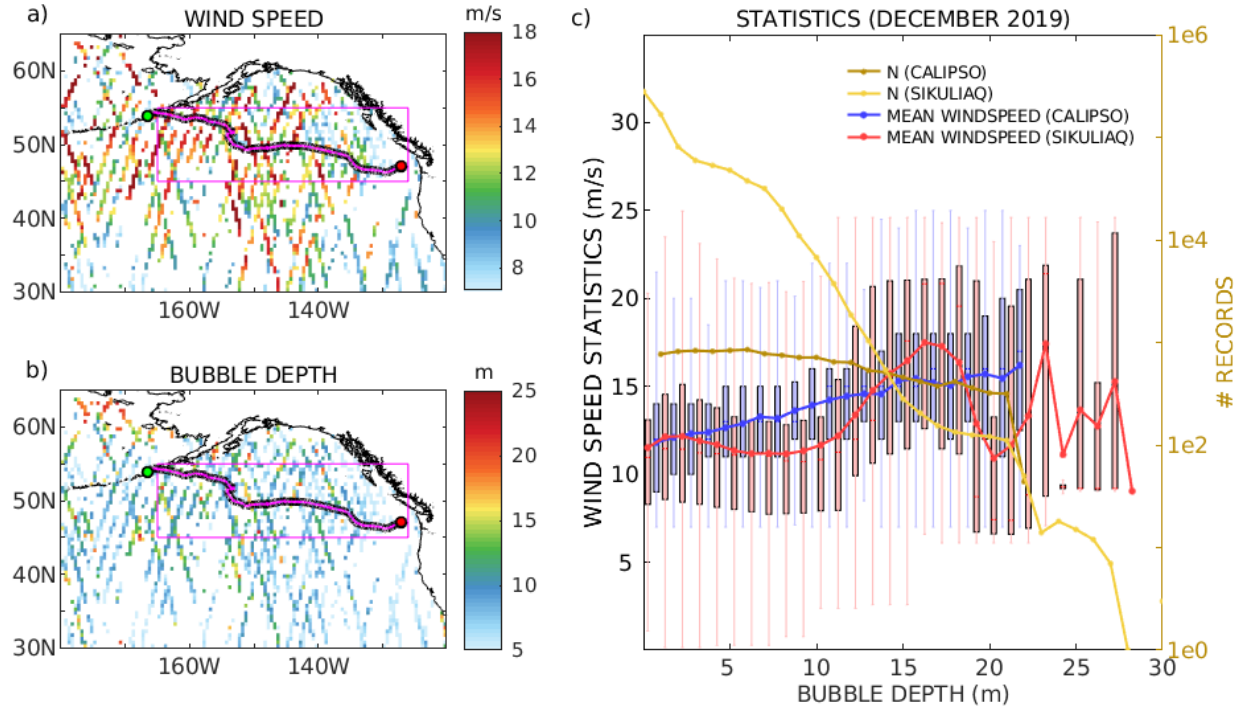
454
455 Fig. 1 a) AMSR-E/AMSR2 wind speed collocated with the CALIPSO track in December 2019 during the breaking
456 bubble experiment. b) Same as a) for the CALIPSO bubble depth. The trajectory of the R/V Sikuliaq is shown in
457 magenta. c) Boxplot of bubble depth from the NRL shipboard lidar (red) and CALIPSO bubble depth (blue) in
458 December 2019 in the Gulf of Alaska. The vertical box is the interquartile range (IQR), and shows the first (Q1),
459 second (red bar) and third (Q3) quantiles. The whiskers on each boxplot show the limits between (Q1-1.5 X IQR and
460 Q3+1.5 X IQR). The number of occurrences is also indicated for the NRL shipboard lidar (yellow) and CALIPSO
461 (dark yellow) with an y-axis on the right.
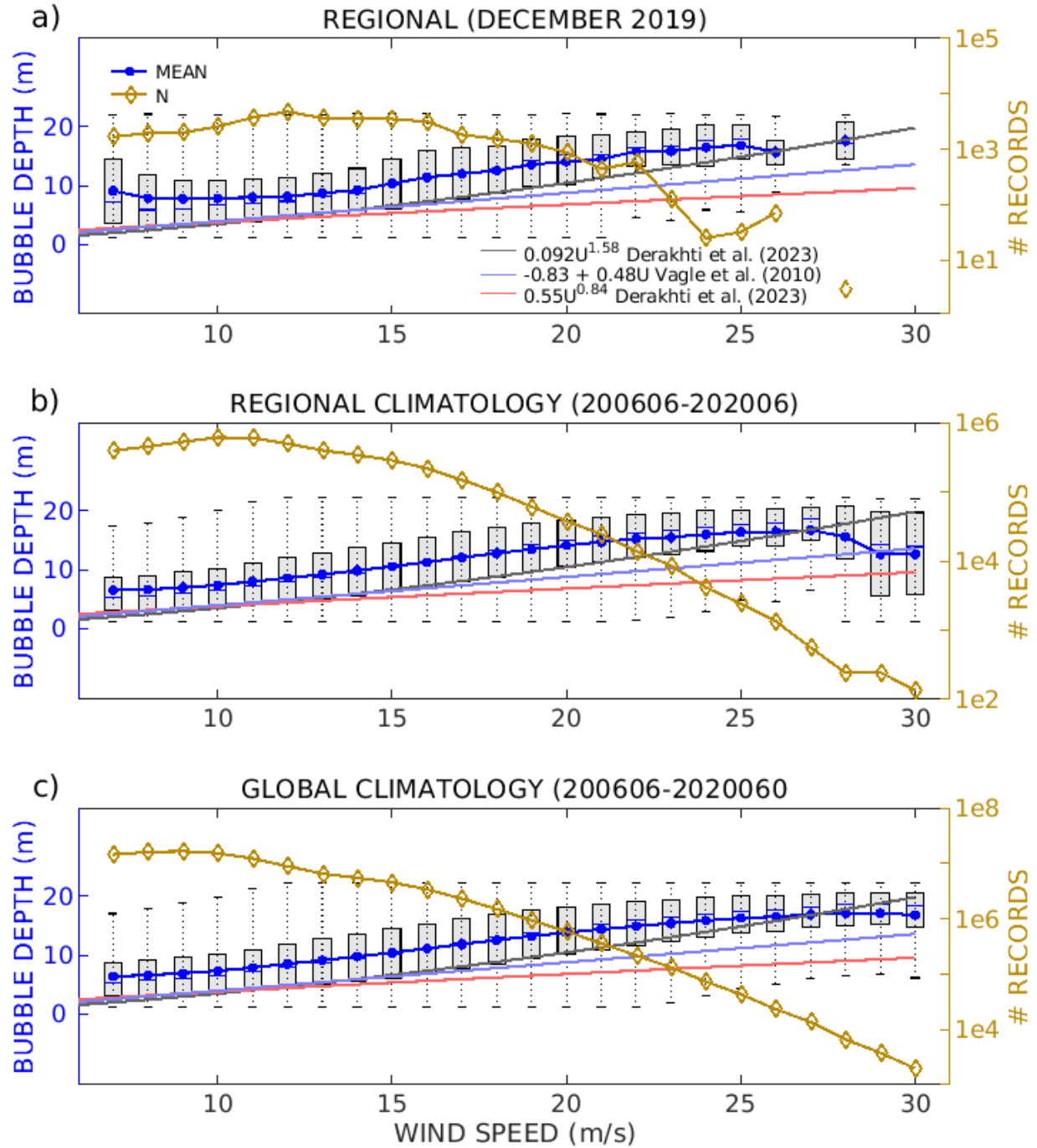462
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467 Fig. 2. Bubble depth as a function of wind speed. a) December 2019 in the Gulf of Alaska (blue); b) 2006-2020
468 climatology in the Gulf of Alaska; c) 2006 – 2020 climatology at global scale. The 3 curves are the three
469 relationships discussed in Cifuentes-Lorenzen (2023). In all panels N is the number of data points (yellow curve
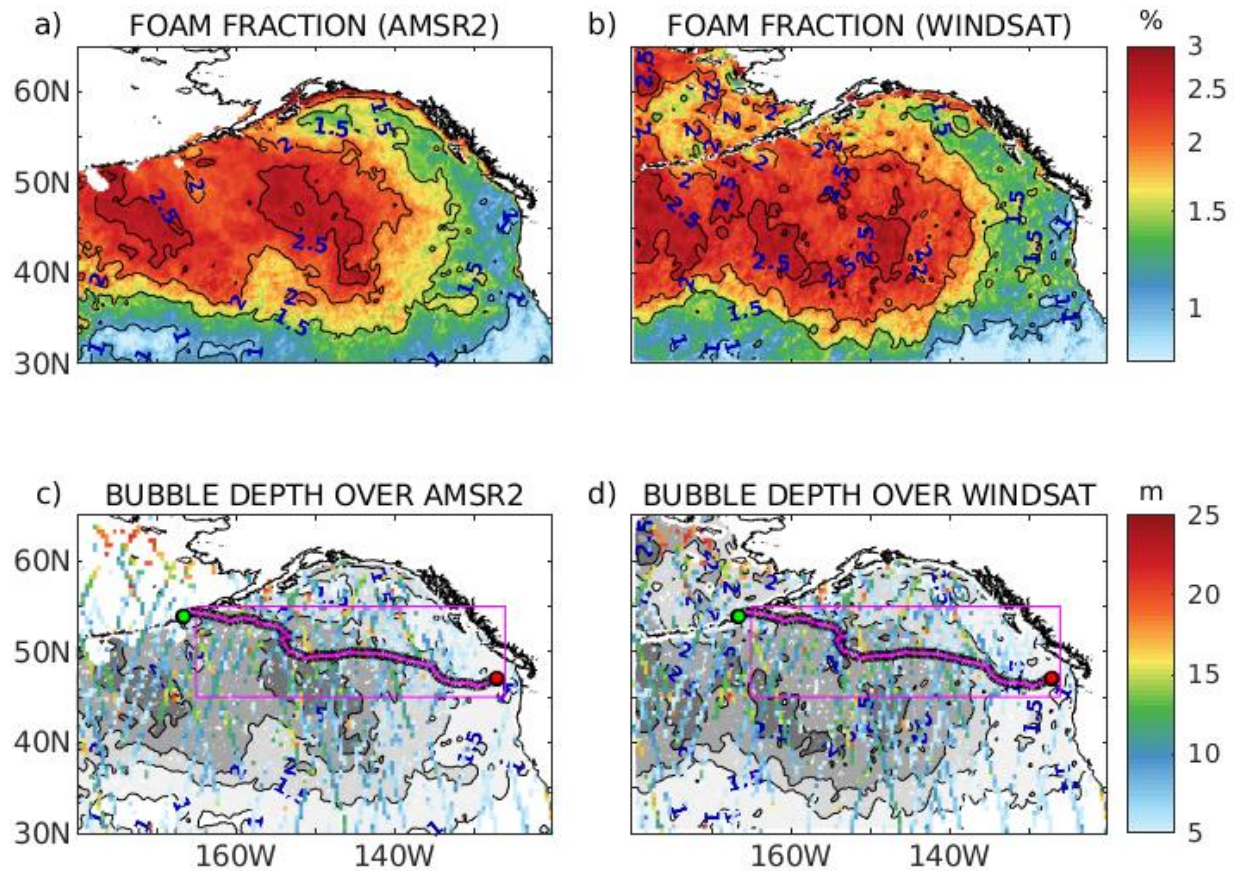470 with diamonds).
471

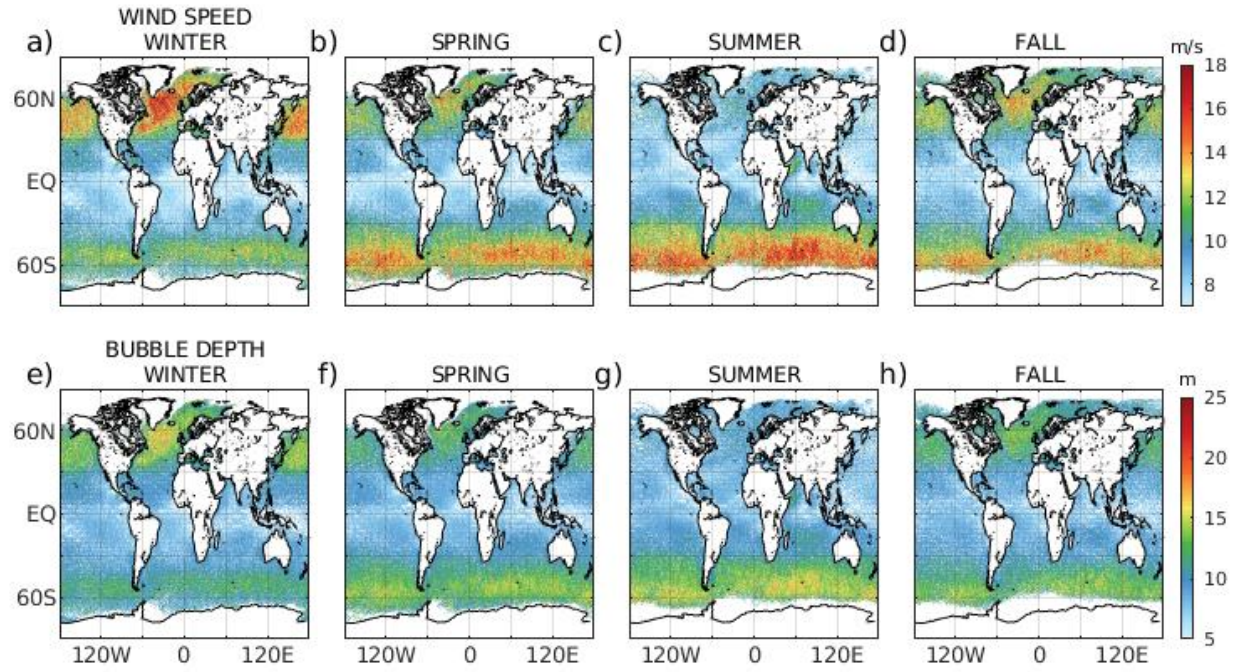Fig. 3. a) Whitecap fraction from AMSR2 and b) WindSat. c) CALIPSO bubble depth with the countour of the whitecap fraction for AMSR2 and d) WindSat.

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479   Fig. 4. Seasonal maps of CALIPSO bubbles depths.