

1 Sparse sampling of intermittent turbulence generated by breaking surface 2 waves

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ABSTRACT

9 We examine how Eulerian statistics of wave breaking and associated turbu-
10 lence dissipation rates in a field of intermittent events compare with those ob-
11 tained from sparse Lagrangian sampling by surface following drifters. We use
12 a polydisperse two-fluid model with large-eddy simulation (LES) resolution
13 and volume-of-fluid surface reconstruction (VOF) to simulate the generation
14 and evolution of turbulence and bubbles beneath short-crested wave break-
15 ing events in deep water. Bubble contributions to dissipation and momentum
16 transfer between the water and air phases are considered. Eulerian statis-
17 tics are obtained from the numerical results which are available on a fixed
18 grid. Next, we sample the LES/VOF model results with a large number of
19 virtual surface-following drifters that are initially distributed in the numerical
20 domain, regularly or irregularly, before each breaking event. Time-averaged
21 Lagrangian statistics are obtained using the time-series sampled by the virtual
22 drifters. We show that convergence of statistics occurs for signals that have
23 minimum length of approximately 1000-3000 wave periods with randomly
24 spaced observations in time and space relative to three-dimensional break-
25 ing events. We further show important effects of (i) extent of measurements
26 over depth and (ii) obscuration of velocity measurements due to entrained
27 bubbles, which are the two typical challenges in most of the available in situ
28 observations of upper ocean wave breaking turbulence. An empirical correc-
29 tion factor is developed and applied to the previous observations of Thomson
30 et al. (2016). Applying the new correction factor to the observations notice-
31 ably improves the inferred energy balance of wind input rates and turbulence
32 dissipation rates. Finally, both our simulation results and the corrected obser-
33 vations suggested that the total wave breaking dissipation rates have a nearly
34 linear relation with active whitecap coverage.

35 **1. Introduction**

36 Many previous studies have shown that turbulence dissipation rates in the ocean surface layer
37 are elevated in the presence of breaking waves (e.g., Agrawal et al. 1992; Terray et al. 1996;
38 Gemmrich and Farmer 2004; Gemmrich 2010; Thomson 2012; Sutherland and Melville 2015).
39 This turbulence is important as an input of energy from the wind to the ocean (Gemmrich et al.
40 1994), a sink of energy for the surface waves (Melville 1996), and a driver of air-sea gas exchange
41 (Zappa et al. 2007). The turbulence is complicated by two-phase flow, in which bubbles are active
42 particles (e.g., Rapp and Melville 1990; Lamarre and Melville 1991; Derakhti and Kirby 2014a;
43 Deike et al. 2017a). Another challenge is the intermittent nature of the forcing, with individual
44 waves breaking as a result of random phase interference patterns and modulational instability
45 (Babanin 2011).

46 Direct measurements of the turbulence beneath breaking surface waves are rare. Recent exam-
47 ples have employed a surface-following reference frame (e.g., Gemmrich 2010; Thomson 2012;
48 Sutherland and Melville 2015; Zippel et al. 2018), which is a natural choice for the observations
49 but a challenge to reconcile with the fixed (Eulerian) reference frames common in numerical mod-
50 els. Furthermore, the observations generally are sparse in space and time, such that it has been
51 difficult to ensure robust statistics. Published observations of turbulence in the ocean near-surface
52 layer generally find that 1) turbulence levels greatly exceed those predicted by law-of-the-wall
53 shear scaling, and 2) this wave-enhanced layer is limited to a depth of approximately one signif-
54 icant wave height in a fixed reference frame (or 1 to 2 m in a wave-following reference frame)
55 (Esters et al. 2018). As most of these observations use acoustic Doppler methods to obtain turbu-
56 lent fluid velocities, the data in the most active portion of the breaking waves are often occluded

57 by bubbles. Thus, existing observations likely represent an incomplete average of the surface
58 conditions, which lack the maxima occurring in space and time.

59 Numerical models and laboratory experiments have been essential in filling the gaps, *e.g.*, quan-
60 tifying turbulence-bubble interaction in bubbly flows beneath breaking waves and providing a
61 high-resolution spatio-temporal variation of turbulence dissipation rates during active breaking.
62 The early studies of Rapp and Melville (1990) and Lamarre and Melville (1991) established time
63 and length scales for the turbulence from two-dimensional focused wave packets, including the
64 importance of bubbles in the setting of the total dissipation. More recently, Wang and Wijesekera
65 (2018) conducted a large-scale laboratory experiment with three-dimensional (3-D), *i.e.*, short-
66 crested, breaking crests in a modulated wave train. Their measurements showed that values of
67 near-surface turbulence dissipation rates during an active breaking event is two to three orders of
68 magnitude larger than those before and after the wave breaking. The recent numerical efforts of
69 Derakhti and Kirby (2014a, 2016) and Deike et al. (2016, 2017a) resolve the breaking of indi-
70 vidual waves and the associated turbulence and bubble dynamics. Derakhti and Kirby (2014a)
71 results showed that high dissipation rates occurs preferentially in regions with high void fraction
72 within bubble plumes. Furthermore, their simulation results predicted that bubble-induced dissipa-
73 tion accounts for approximately 50% of the total wave-breaking-induced turbulence dissipation
74 regardless of breaker type and intensity. Here, bubble-induced dissipation refers to the enhance-
75 ment of turbulence dissipation due to the subgrid-scale (SGS) turbulent motions generated by the
76 dispersed bubbles (see Derakhti and Kirby 2014a, §4.3.1 for more details).

77 The present work is motivated by the study of Thomson et al. (2016), in which turbulence
78 dissipation rates were estimated using Doppler velocity profiles within the upper meter of the
79 wave-following surface. That study concluded that strong turbulence is isolated to a very thin layer
80 (< 1 m), but that orbital motions advect the turbulence over vertical scales of at least one significant

wave height. The main focus of Thomson et al. (2016) was evaluating the energetic balance at the surface, with the conclusion that the observed energy dissipation rates were insufficient to balance the energy input rates using several different formulations.

Here, we revisit the topics of Thomson et al. (2016) by sampling a high-fidelity numerical model in the Lagrangian mode of the surface-following observations. We use a polydisperse two-fluid model (Derakhti and Kirby 2014a) with large-eddy simulation (LES) resolution and volume-of-fluid surface reconstruction (VOF) to simulate the generation and evolution of turbulence and bubbles beneath 3-D short-crested wave breaking events in deep water (§2). We first scale the model domain to match the observed whitecap coverage values, and we scale the model wave heights to match the wind-wave (*i.e.*, equilibrium) portion of the observed spectrum (*i.e.*, neglecting swell). We then determine the effects of sparse sampling and intermittent breaking, as well as the effects of data occlusion by bubbles and limitations in the vertical extent of the observed profiles (§3). In §4, we comment on the apparent discrepancy between the observed wind-input energy fluxes and total turbulence dissipation rates reported by Thomson et al. (2016). Examination of potential Lagrangian sampling bias related to a partially trapped drifter in convergence zones in the turbulence observations is left for future work.

2. Methods

In this section, we first present the model governing equations for continuity of mass and momentum of liquid and gas phases of a polydisperse two-fluid mixture, as described in Derakhti and Kirby (2014a). The model set-up including details of the incident wave conditions and the scaling of the model domain to match observations of whitecap coverage are then described. Finally, we explain our methodology to convert the model results to surface following virtual drifters.

103 Demonstrations of model convergence and performance, including detailed comparisons of free
 104 surface evolution, bubble void fraction, integral properties of the bubble plume, organized and
 105 turbulent velocity fields and total wave-breaking-induced energy dissipation, for various deep-
 106 and shallow-breaking waves may be found in (Derakhti and Kirby 2014a,b, 2016; Derakhti et al.
 107 2018, 2019).

108 *a. Mathematical formulations*

109 The computations here are performed using the LES/VOF Navier-Stokes solver TRUCHAS
 110 (Francois et al. 2006) with extensions of a polydisperse bubble phase and various turbulence clo-
 111 sures (Carrica et al. 1999; Ma et al. 2011; Derakhti and Kirby 2014a). Details of the mathematical
 112 formulations and numerical method may be found in Derakhti and Kirby (2014a, §2).

113 The filtered governing equations for continuity of mass and momentum of the liquid phase are
 114 given by (Derakhti and Kirby 2014a):

$$\frac{\partial \alpha \rho}{\partial t} + \frac{\partial \alpha \rho \tilde{u}_j}{\partial x_j} = 0, \quad (1)$$

$$\frac{\partial \alpha \rho \tilde{u}_i}{\partial t} + \frac{\partial \alpha \rho \tilde{u}_i \tilde{u}_j}{\partial x_j} = \frac{\partial \Pi_{ij}}{\partial x_j} + \alpha \rho g \delta_{3i} + \mathbf{M}^{gl}, \quad (2)$$

115
 116 where $(i, j) = 1, 2, 3$; ρ is a constant liquid density; α and \tilde{u}_i are the volume fraction and the filtered
 117 velocity in the i direction of the liquid phase respectively; δ_{ij} is the Kronecker delta function; g is
 118 the gravitational acceleration; and $\Pi_{ij} = \alpha(-\tilde{p} \delta_{ij} + \tilde{\sigma}_{ij} - \rho \tau_{ij})$ with \tilde{p} the filtered pressure, which
 119 is identical in each phase due to the neglect of interfacial surface tension, $\tilde{\sigma}_{ij}$ viscous stress and
 120 τ_{ij} the SGS stress estimated using an eddy viscosity assumption and the Dynamic Smagorinsky
 121 model, which includes liquid/bubble interaction effects (for more details see Derakhti and Kirby
 122 2014a, §2.4). Finally, \mathbf{M}^{gl} are the momentum transfers between liquid and gas phases, including
 123 the filtered virtual mass, lift, and drag forces (Derakhti and Kirby 2014a).

124 Using the same filtering process as in the liquid phase, the equations for the bubble number
 125 density and continuity of momentum for each bubble size class with a diameter d_k^b , $k = 1, \dots, N_G$,
 126 are then given by (Derakhti and Kirby 2014a):

$$\frac{\partial N_k^b}{\partial t} + \frac{\partial \tilde{u}_{k,j}^b N_k^b}{\partial x_j} = R_k^b, \quad (3)$$

$$0 = -\frac{\partial \alpha_k^b \tilde{p}}{\partial x_j} \delta_{ij} + \alpha_k^b \rho^b g_i + \mathbf{M}_k^{lg}, \quad (4)$$

127 where m_k^b , N_k^b , $\alpha_k^b = m_k^b N_k^b / \rho^b$, and $\tilde{u}_{k,j}^b$ are the mass, number density, volume fraction, and filtered
 128 velocity in the j direction of the k th bubble size class; ρ^b is the bubble density; and R_k^b includes the
 129 source due to air entrainment in the interfacial cells (Derakhti and Kirby 2014a, §2.3), intergroup
 130 mass transfer, and SGS diffusion terms. Finally, \mathbf{M}_k^{lg} represents the total momentum transfer
 131 between liquid and the k th bubble size class, and satisfies $\mathbf{M}^{gl} + \sum_{k=1}^{N_G} \mathbf{M}_k^{lg} = 0$. In (4), we neglect
 132 the inertia and shear stress terms in the gas phase following Carrica et al. (1999) and Derakhti and
 133 Kirby (2014a).

135 *b. Model set-up*

136 Our numerical experiments are carried out in a virtual wave tank of unperturbed constant depth
 137 h , extending a length L_x in the x direction, and $\pm L_y/2$ in the transverse y direction. The vertical
 138 direction z in the fixed reference frame is positive upward and measured from the still water level.
 139 The virtual wave tank is sufficiently deep to avoid any depth-limited wave breaking, such that the
 140 experiments remain focused on whitecaps.

141 All simulations are performed with the model initialized with quiescent conditions. An incident
 142 wave packet is then generated at the model upstream boundary ($x = 0$). The input focused wave
 143 packet was composed of $N = 10$ sinusoidal components of steepness $a_n k_n$, $n = 1, \dots, N$, where
 144 a_n and k_n are the amplitude and wave number of the n th frequency component. The steepness of

145 individual wave components is taken to be constant across the spectrum, or $a_1 k_1 = a_2 k_2 = \dots =$
 146 $a_N k_N = S_g / N$ with $S_g = \sum_{n=1}^N a_n k_n$ taken to be a measure of the wave train global steepness. Based
 147 on linear theory, the free surface elevation for the 3-D short-crested focused packets (Wu and Nepf
 148 2002; Derakhti et al. 2018; Kirby and Derakhti 2019) at the wavemaker is given by

$$\eta(0, y, t) = \sum_{n=1}^N a_n \cos[2\pi f_n(t - t_f) + \frac{k_n x_f}{\cos \theta(y)}], \quad (5)$$

149 where f_n is the frequency of the n th component, x_f and t_f are the predefined, linear theory esti-
 150 mates of location and time of the focal point respectively, and $\theta(y)$ is the angle of incidence of
 151 each wave component at various transverse locations with $\cos \theta(y) = x_f / \sqrt{x_f^2 + y^2}$. The discrete
 152 frequencies f_n were uniformly spaced over the band $\Delta f = f_N - f_1$ with a central frequency de-
 153 fined by $f_c = (f_N + f_1)/2$. Increasing the global steepness S_g and/or decreasing $\Delta f/f_c$ increases
 154 the total wave energy loss due to the resulting breaking event(s) in the virtual tank. Finally, liquid
 155 velocities for each wave component are calculated using linear theory and then superimposed at
 156 the wavemaker.

157 We define the spectrally-weighted frequency of the wave field f_s as

$$f_s = \frac{\int f E(f) df}{\int E(f) df}, \quad (6)$$

158 where E [m^2/s] is the power spectral density of the wave field. The characteristic wave length L_s
 159 and period T_s are then calculated based on f_s and using the linear dispersion relation (as in Tian
 160 et al. 2010; Derakhti and Kirby 2016). The reference x -location x^* is taken as the location at which
 161 the first forward-moving jet of considered breaking events in a numerical case hits the undisturbed
 162 free surface, and is normalized by L_s . Further, we define $y^* = y/L_s$ and $z^* = z/L_s$.

163 Each numerical case is defined by setting the geometry of the virtual tank and the input wave
 164 packet. Here, three representative cases are considered, with all relevant parameters summarized

165 in Table 1. In all cases, most of the wave components in the input packets are characterized as
166 deep water waves.

167 Figure 1a shows the temporal variation of the normalized free surface elevations at the center of
168 the tank and slightly upstream of the break point, $(x^*, y^*) = (-0.1, 0)$, for the case T1. The results
169 indicate that the current wavemaker setup (Eq. 5) results in a repeatable sequence of waves in the
170 incident packet with a period of $T_g = N/\Delta f$. In all cases, the observed main breaking events in
171 the virtual tank occur approximately every T_g . As shown in Figure 1a,d,e, however, the incident
172 waves and the x -location of the main breaking event within each T_g are not the same; this may be
173 partially because of seiching in the virtual tank and reflections from the numerical boundaries.

174 We only consider the model results for $t > t_0$ for all the analyses presented in this paper, where
175 $t_0 > 12T_g > 200$ s is a time after which the background turbulence levels reach a quasi-steady state.
176 For each case, we define the main breaking event Em ($m = 1, 2, \dots, N_E$) as the most energetic
177 breaking event that occurs in $m - 1 < (t - t_0)/T_g < m$. The total number of considered main
178 breaking events N_E for the cases T1, T2, and T3 are 9, 10, and 6 respectively. Thus considering an
179 output sampling rate of f_{out} , the time series of all Eulerian variables predicted by the model have
180 $N_E T_g f_{out}$ data points, where f_{out} was 20 Hz for T1 and T2, and 25 Hz for T3.

181 *c. Matching the model and observed conditions*

182 We need to choose a number of well-defined parameters to present both the wave breaking
183 forcing and model results in a non-dimensional form, such that they can be appropriately scaled to
184 field conditions. Here our goal is to have the wave spectrum E and the fractional area of breaking
185 crests of the simulated cases as consistent as possible with those observed in the field. The latter
186 is usually referred to as the active part of the whitecap coverage W of visible breaking crests,
187 hereafter referred to as W_A , which is a space- and time-averaged quantity calculated over a given

188 domain. There is a growing body of literature documenting a direct relationship between W_A
 189 and the total wave breaking energy dissipation in the upper ocean (Callaghan et al. 2016, 2017;
 190 Callaghan 2018; Anguelova and Hwang 2016). We also need a characteristic breaking wave height
 191 to scale the vertical profiles of wave-breaking-related dynamical measures, such as the turbulence
 192 dissipation rates.

193 Figure 1b shows the normalized power spectral density $E^* = EH_{eq}^{-2}f_s$ at $(x^*, y^*) = (-0.1, 0)$ for
 194 all three simulated cases, the vertical dashed lines show the frequency range of wave components
 195 in the input packet for each case. Figure 1c shows examples of observed E^* for various values of
 196 whitecap coverage W and wind speed U_{10} in the vicinity of OWS-P at 50° N, 145° W provided by
 197 Schwendeman and Thomson (2015); Thomson et al. (2016). Here we define H_{eq} as a characteristic
 198 breaking wave height given by

$$H_{eq} = 4 \sqrt{\int_{f_s}^{2f_s} E(f) df}. \quad (7)$$

199 The results demonstrate that both the simulated and observed E^* have a self-similar shape in
 200 the range $f_s < f < 2f_s$, which is usually called an equilibrium range of a wave spectrum (Phillips
 201 1985). Second, the shape of the simulated wave spectrum for $f > f_s$ is similar to the observations.
 202 The f^{-4} dependence in the simulated spectra $E(f)$ is achieved by using the constant steepness
 203 spectrum ($a_i k_i = \text{const.}$) in which $E(f) \sim a_i^2 \sim k_i^{-2}$ and $k_i^{-2} \sim f^{-4}$ from the linear dispersion rela-
 204 tion for deep water waves. However, in the field spectra there is much more energy at frequencies
 205 $f < f_s$ due to the presence of swell. Many field observations of the speed of visible whitecaps have
 206 shown that the dominant speed of breaking waves is about half the phase speed of waves at the
 207 peak of the energy spectrum (Melville and Matusov 2002; Gemmrich et al. 2008; Thomson and
 208 Jessup 2009; Kleiss and Melville 2010; Schwendeman et al. 2014; Sutherland and Melville 2013).
 209 Assuming linear dispersion, this means that the frequency associated with breaking waves f_b is
 210 noticeably larger than the peak frequency f_p , and that f_b is usually in the equilibrium range. Thus

211 H_{eq} will be an appropriate choice for a characteristic height of visible breaking waves. Further,
 212 since H_{eq} is not sensitive to the amount of energy of swell waves, which is completely absent in our
 213 numerical simulations, it is preferable for the purpose of model-field observations comparisons of
 214 vertical scaling of turbulence dissipation rates. Here, H_{eq} prior to breaking onset is 0.24m, 0.20m,
 215 and 0.14m in the simulated cases T1, T2, and T3 respectively, with $H_{eq}/H_s \approx 0.8$ because the main
 216 part of the wave energy is distributed in the range $f_s < f < 2f_s$ (see Figure 1b).

217 Figure 1c shows that H_{eq} as defined in Eq. (7) varies between 0.45 and 0.7 of the correspond-
 218 ing significant wave height H_s for the field conditions with $6 < U_{10} < 16$ m/s. Considering the
 219 Pierson-Moskowitz spectrum, we obtain $f_s = 1.3f_p$ and $H_{eq}/H_s = 0.57$ which is consistent with
 220 the averaged value of $H_{eq}/H_s \approx 0.6$ obtained from the observations of Schwendeman and Thom-
 221 son (2015) and Thomson et al. (2016) in the North Pacific.

222 We define the active breaking index $I(x, y, t)$ as

$$I(x, y, t) = \mathcal{H}[\alpha_{sa}^b(x, y, t) - \alpha_{th}^b], \quad (8)$$

223 where \mathcal{H} is the Heaviside step function, α_{sa}^b is the vertical average of the bubble void fraction
 224 over the surface layer of depth $d_s = 0.4H_{eq}$, and α_{th}^b is a threshold value. Then, we obtain the
 225 fraction of the active breaking crests, or active whitecap coverage, W_A of the breaking events occur
 226 between time t_1 and t_2 and over the area \mathcal{A} as

$$W_A = \frac{\int_{t_1}^{t_2} \int_{-y_1}^{y_1} \int_{x_1}^{x_2} I \, dx \, dy \, dt}{\mathcal{A}(t_2 - t_1)}, \quad (9)$$

227 where x_1, x_2, y_1 indicate the horizontal extent of the averaging area $\mathcal{A} = 2y_1(x_2 - x_1)$ and are
 228 set such that it includes the breaking crests. Unless stated otherwise, we set $t_1 = t_0$ and $t_2 =$
 229 $t_0 + N_E T_g$ to do the time averaging over all available breaking events in the virtual tank after $t = t_0$.
 230 In the field, W_A (and W) is obtained from image processing of visible whitecaps and should be
 231 independent of the selected field of view. In Eq. (9), however, W_A depends on \mathcal{A} . As we will

232 explain later in this paper, choosing $\alpha_{th}^b = 0.02$ provides estimates of W_A which are consistent with
233 the observed values of whitecap coverage in the open ocean (Schwendeman and Thomson 2015).
234 We note that if d_s varies between $0.2H_{eq}$ and $0.6H_{eq}$ the corresponding W_A values vary less than
235 30% of the W_A value estimated using $d_s = 0.4H_{eq}$ for the simulated cases. Finally, the temporal
236 variation of the instantaneous W_A values using Eq. (9) for the various individual simulated breakers
237 (not shown) are consistent with the data reported in Figure 1 of Callaghan et al. (2016).

238 *d. Conversion of the model results to virtual drifters*

239 In this paper, our main goal is to examine potential sampling biases and convergence of statistics
240 of the field observations of intermittent wave breaking turbulence collected by surface following
241 platforms (e.g., SWIFT drifters) using our high resolution numerical simulations. To do this, we
242 need to sample our model results, which are available at fixed Eulerian grid points, in a manner
243 which is similar to how a physical drifter (Figure 1f) obtains samples in the field.

244 We first introduce a number of virtual drifters that move with the free surface and local liquid
245 velocity in the computational domain. Then, we interpolate the model Eulerian results onto vertical
246 line segments that are attached to the virtual drifters and extend from the instantaneous free surface
247 $z = \eta$ to $z = \eta - l_{vd}$. In the surface-following reference frame $z_{sf} = z - \eta$, all the interpolated
248 results will be in the range $-l_{vd} < z_{sf} < 0$.

249 For each breaking event Em ($m = 1, 2, \dots, N_E$), a total number of 231 virtual drifters are released
250 at $t = t_0 + (m - 1)T_g$, which is well before the onset of the main breaking event Em , and remain
251 in the water for a time T_g . We consider both uniform and random initial spacing of the virtual
252 drifters to make sure that the resultant statistics are independent of the initial deployment of the
253 virtual drifters. Figure 1d,e show two snapshots of the instantaneous locations of the virtual drifters
254 (markers) released uniformly at $x^* = -0.2, -0.5 < y^* < 0.5$ for the breaking events E3 and E4 of

255 the case T1. Figure 1f shows a snapshot of a physical drifter in the field in the vicinity of an active
256 breaking crest propagating towards the drifter.

257 The horizontal location of each virtual drifter is updated using the vertical average of the water
258 horizontal velocity components over the surface layer of depth $0.2H_{eq}$. Panels (g) and (h) show
259 the corresponding horizontal displacements of some of the virtual drifters released in a uniform
260 grid and during the events E3 and E4 of the case T1 respectively. Panel (i) shows an example of
261 the horizontal displacement of a SWIFT drifter in the field. In these frames, each color segment
262 represents the horizontal displacement during a fixed time, equal to T_s in the model results and
263 $T_p/2$ in the observations. Both simulated and observed results indicate that a drifter trapped in an
264 active breaking crest may experience horizontal displacements that are significantly greater than
265 when it is riding on a non-breaking crest. This is consistent with the recent work of Deike et al.
266 (2017b); Pizzo et al. (2019).

267 3. Results

268 A glossary of all variables used hereafter is given in Table A1 in the appendix. In our model re-
269 sults, the rate of transfer of energy from the resolved motions to the SGS motions is $\varepsilon_{sgs} = v_{sgs} |\mathcal{S}|^2$
270 where v_{sgs} is the SGS eddy viscosity, $|\mathcal{S}|^2 = 2\mathcal{S}_{ij}\mathcal{S}_{ij}$, and $\mathcal{S}_{ij} = \frac{1}{2}(\partial\tilde{u}_i/\partial x_j + \partial\tilde{u}_j/\partial x_i)$ is the
271 resolved rate of strain. ε_{sgs} includes both shear- and bubble-induced dissipation (Derakhti and
272 Kirby 2014a). At high Reynolds number, with the filter width much larger than the Kolmogorov
273 length scale, the viscous dissipation rate is typically much smaller than ε_{sgs} , and thus ε_{sgs} approx-
274 imates the turbulent kinetic energy (TKE) dissipation rate ε , which is commonly considered as a
275 characteristic indicator of intensity of turbulence.

276 Figure 2 shows two snapshots of 3-D variation of ε_{sgs} during active breaking period and $\approx 2T_s$
277 after the breaking-onset for the breaking event E3 of the case T1. Consistent with previous wave

breaking simulations (see, for example, Derakhti and Kirby 2014a, Figures 11 and 12), our numerical results indicate that wave-breaking-induced ε_{sgs} has a strong temporal and spatial variation, with local values of ε_{sgs} varying from $O(1) \text{ m}^2\text{s}^{-3}$ (or W/kg) down to the background levels, and with large values of ε_{sgs} concentrated near the wave crest and in regions of high void fraction (bubble void fractions are not shown here). The latter is consistent with the recent laboratory measurements of turbulence dissipation rates ε within wave breaking crests by Deane et al. (2016). They reported large values of $\varepsilon > 1 \text{ W/kg}$ during the acoustically active phase of wave breaking in which air is actively entrained and fragmented into bubbles.

As summarized in §1, in most practical applications the long-time-average (*e.g.*, over many wave periods) of TKE dissipation rates over a relatively large surface area, $O(100L_p \times 100L_p)$, is of interest. In this section, we first examine how the Eulerian averages of ε_{sgs} compare with those obtained from surface following virtual drifters. Then we comment on the convergence of statistics obtained from the virtual drifters. Last, we examine the effect of incomplete sampling of ε_{sgs} by the virtual drifters due to limited vertical field of view and occlusion due to the entrained bubbles.

293 *a. Lagrangian vs Eulerian averaging of ε_{sgs}*

Figure 3a shows the spatio-temporal variation of the horizontal average of ε_{sgs} in a surface-following reference frame, $\check{\varepsilon}_{sgs}(t, z_{sf})$, for the breaking event E3 of the case T1. Here, the averaging is performed over a surface that is parallel to $\eta(x, y, t)$ at a distance $|z_{sf}|$ below the free surface with the horizontal area of $2L_s \times L_s$. The values of $\check{\varepsilon}_{sgs}$ before the breaking onset ($t/T_s < 133$) and those after $4T_s$ after the breaking onset ($t/T_s > 137$) are approximately comparable, demonstrating that the elevated turbulence dissipation rates due to a wave breaking event return to background levels after a few local wave periods. Figure 3b demonstrates the same trend in the depth in-

301 tegrated values of $\check{\varepsilon}_{sgs}$. Here the first two peaks are corresponding to the main breaking wave
 302 and the successive smaller breaking wave that occur in $133 < t/T_s < 134$ and $134 < t/T_s < 135$
 303 respectively.

304 Figure 3c,e show the spatio-temporal variation of ε_{sgs} sampled by the two virtual drifters shown
 305 in Figure 2 during E3 of the case T1. Their corresponding time-averaged profiles, over $T_g \approx$
 306 $9.5T_s$, in the surface-following reference frame are shown by the thick black lines in the panels
 307 (d) and (f), where the background thin gray lines represent the results from all available virtual
 308 drifters released before the beginning of the breaking event T1-E3. The considerable variation
 309 in $\bar{\varepsilon}_{sgs}$ obtained from virtual drifters indicates that, in the presence of surface wave breaking, the
 310 Lagrangian-based values of dissipation rates averaged over approximately 10 wave periods still
 311 have a considerable variation depending on the time fraction that a drifter is positioned within the
 312 localized turbulence patches generated during the active breaking process. We will examine the
 313 convergence of Lagrangian-averaged dissipation rates in the next section.

314 Figure 4a shows examples of the ensemble-averaged profiles of time-averaged dissipation rate
 315 $\bar{\varepsilon}^L$, which are obtained from the Lagrangian virtual drifters (or time-averaging over many wave
 316 periods at least 1000 wave periods) as

$$\bar{\varepsilon}^L = \frac{\sum_{i=1}^{i=231N_E} \bar{\varepsilon}_{sgs}^i}{231N_E}. \quad (10)$$

317 Here, $\bar{\varepsilon}_{sgs}^i$ is the time-averaged dissipation rate over time T_g sampled by the i^{th} ($i = 1, \dots, 231N_E$)
 318 virtual drifter (Figure 3d,f). Figure 4a also shows examples of the ensemble-averaged profiles of
 319 the time-horizontal-averaged of the Eulerian model results $\bar{\varepsilon}^E$ in the surface-following reference
 320 frame given by

$$\bar{\varepsilon}^E = \frac{\sum_{m=1}^{m=N_E} \bar{\varepsilon}_{sgs}^m}{N_E}, \quad (11)$$

321 where $\bar{\epsilon}_{sgs}^m$ is the time-horizontal-averaged dissipation rate over T_g of the breaking event E_m ($m =$
322 $1, \dots, N_E$). Further, the corresponding profiles of $\bar{\bar{\epsilon}}^E$ in the fixed reference frame is presented in
323 the panel (b).

324 The vertical structure of the long-time-averaged turbulence dissipation rate $\bar{\bar{\epsilon}}$ predicted by the
325 model (Figure 4) has a number of important features. First our results indicate that $\bar{\bar{\epsilon}}$ below a
326 surface breaking layer with depth H_{eq} ($\approx 0.6H_s$) is proportional to $(-z_{sf}/H_{eq})^{-2}$ in the surface-
327 following reference frame and $(-z/H_{eq})^{-2}$ in the fixed reference frame. This is consistent with
328 the model proposed by Terray et al. (1996) (T96) in the fixed reference frame below the surface
329 breaking layer with depth that T96 took to be $z > 0.6H_s$, and with more recent observations in a
330 surface-following reference frame (Gemmrich 2010; Zippel et al. 2018; Sutherland and Melville
331 2015).

332 In contrast to T96, but consistent with the same recent field observations (Gemmrich 2010;
333 Sutherland and Melville 2015; Thomson et al. 2016; Zippel et al. 2018), $\bar{\bar{\epsilon}}$ is not constant in the
334 surface breaking layer ($z < 0.6H_s$). Our model results indicate that in the upper half of the surface
335 breaking layer in the surface-following reference frame, $|z_{sf}|/H_{eq} < 1/2$, $\bar{\bar{\epsilon}}$ is well described with
336 $\bar{\bar{\epsilon}}(z_{sf} = 0) \exp(\beta z_{sf}/H_{eq})$ where β is a decreasing function of the active whitecap coverage W_A . In
337 other words, $\bar{\bar{\epsilon}}$ will be approximately constant for extreme wave breaking forcing with a relatively
338 large value of W_A ($\gg 0.01$) for $|z_{sf}|/H_{eq} < 1/2$. Figure 4a also shows that $\bar{\bar{\epsilon}}$ still follows the
339 exponential decay in the lower half of the surface breaking layer $1/2 < |z_{sf}|/H_{eq} < 1$ with β
340 increasing from approximately 8 for small W_A to 9 for large W_A values for the simulated cases.

341 Furthermore, in the fixed reference frame and within the surface breaking layer, $\bar{\bar{\epsilon}}$ is described
342 with $\bar{\bar{\epsilon}}(z = 0) \exp(\beta z/H_{eq})$ where β is again a decreasing function of W_A but varies between 3
343 and 4 for the simulated cases. Our model results demonstrate that the Eulerian maximum of $\bar{\bar{\epsilon}}$ is
344 above $z = 0$, where $\bar{\bar{\epsilon}}$ has persistently large values in $0 < z/H_{eq} < 1$ compared with those in other

345 z elevations. We note that Thomson et al. (2016) have reported a comparable vertical structure for
346 $\bar{\bar{\varepsilon}}$ but with the Eulerian maximum slightly below $z = 0$ and the observed values of $\bar{\bar{\varepsilon}}$ in the fixed
347 reference frame immediately started to decrease after reaching their maximum. We suspect that
348 these differences between observed and simulated results are mainly due to the limitation of the
349 experimental method in high bubble void fraction regions, and thus, the exclusion of high TKE
350 dissipation rate values in the corresponding data. we will comment on further effects of this issue
351 in §3.c below.

352 In addition to the vertical distribution of $\bar{\bar{\varepsilon}}$, the depth-integrated turbulence dissipation rate $\int \bar{\bar{\varepsilon}} dz$
353 ($= \int \bar{\bar{\varepsilon}} dz_{sf}$) is of significant interest, and as reviewed in §2.c, its relationship with the whitecap
354 coverage W (or W_A) remains an active area of research with important impacts for wave prediction
355 modeling, air-sea interaction, and ocean engineering communities. Figure 5a shows the variation
356 of $\int \bar{\bar{\varepsilon}} dz$ with the active whitecap coverage W_A (Eq. 9) for our model results. As reviewed by
357 Brumer et al. (2017), recent observations reveal that W is an increasing function, with the scatter
358 of data, of the wind speed U_{10} ; varies between 10^{-3} and 0.1 for $7 < U_{10} < 20$ m/s. Further,
359 available data of active whitecap coverage $W_A = \lambda W$ ($\lambda < 1$) indicates that λ is a decreasing
360 function, with the scatter of data, of U_{10} ; varies between 0.5 ± 0.25 and 0.2 ± 0.1 for $5 < U_{10} < 20$
361 m/s (Scanlon and Ward 2016, Figure6.b). Thus the resulting values of $8 \times 10^{-4} < W_A < 0.021$
362 shown in 5a, after dividing by the correction factor λ , are comparable with the range of observed
363 W values for $7 < U_{10} < 20$ m/s. Thus we conclude that our definition of W_A (Eq. 9) may be
364 interpreted as an active part of the whitecap coverage values reported in the literature, and that
365 the energetics of wave breaking forcing in the simulated cases may be comparable to most of the
366 available field observations including Schwendeman and Thomson (2015); Thomson et al. (2016).

$$\int [\bar{\varepsilon} - \bar{\varepsilon}_0] dz_{sf} = a W_A^b, \quad (12)$$

368 where ε_0 is the background turbulence dissipation rate due to mechanisms other than visi-
 369 ble wave breaking, $\varepsilon_0 \approx 0$ in our simulations. Using least squared curve fitting, we obtain
 370 $a = 0.077(-0.026, +0.039)$ and $b = 0.94(\pm 0.07)$ (with $R^2 = 0.98$), where coefficients in paren-
 371 theses represent 95% confidence intervals. This is an interesting and important result, and we will
 372 further comment on it in §4.

373 Figure 5b shows that a relatively high fraction of total dissipation rate occurs above the mean sea
 374 level. This fraction is still noticeably high, $\approx 80\%$, even for small W_A values of about 0.001. This
 375 is consistent with the field observations of Gemmrich (2010) showing that most of the breaking
 376 turbulence is concentrated very close to the surface, especially in the wave crest. This is also con-
 377 sistent with the laboratory study of Deane et al. (2016), who found that relatively high dissipation
 378 rate values are concentrated in the crest region of the breaking waves. In particular, Deane et al.
 379 (2016) find that the majority of energy dissipation occurs within bubble plumes, where turbulent
 380 dissipates saturate around 10^0 or $10^1 \text{ m}^2 \text{s}^{-3}$.

381 Finally, the results shown in Figure 4a and Figure 5 demonstrate that the Lagrangian statistics of
 382 intermittent wave breaking turbulence, obtained from the sampled data by freely drifting platforms,
 383 are representative of the corresponding Eulerian statistics when the length of the Lagrangian data
 384 is very large compared with the local wave breaking period. In the next section, we examine how
 385 such Lagrangian statistics converge as a function of the length of data.

386 *b. Convergence of statistics*

387 In this section, we examine how the Lagrangian statistics of dissipation rates obtained from n
388 randomly selected virtual drifters $\bar{\varepsilon}_n^L$ compare to the associated statistics using all available virtual
389 drifters $\bar{\varepsilon}^L$ (Eq. 10). We note that the normalized length of the considered data \mathcal{L}/T_s used to
390 obtain $\bar{\varepsilon}_n^L$ and $\bar{\varepsilon}^L$ are $\approx 9.5n$ and $\approx 9.5 \times 231N_E$ respectively.

391 Figure 6 shows $\bar{\varepsilon}_n^L$ (solid lines) for three n values and two different W_A as well as the corre-
392 sponding ensemble-averaged profiles using all the virtual drifters $\bar{\varepsilon}^L$ (dashed lines). The random
393 selection of n virtual drifters for each n value were repeated 100 times (gray lines). For better
394 visibility, five examples of these profiles are plotted as individual black curves. As expected for
395 both cases, $\bar{\varepsilon}_n^L$ (solid lines) converges to $\bar{\varepsilon}^L$ (dashed lines) with increasing n or the length of signal.
396 However, the convergence of statistics with increasing n occurs more rapidly for the case with a
397 larger W_A value (or more active breaking) compared with the case with a relatively smaller W_A
398 value.

399 Figure 7a shows the variation of the normalized RMSE of $\int \bar{\varepsilon}_n^L dz_{sf}$ (with mean value $\int \bar{\varepsilon}^L dz_{sf}$)
400 with normalized signal length $\mathcal{L}/T_s \approx 9.5n$ for the two cases shown in Figure 6. The results
401 show that this error may be fairly reasonably estimated using the Gaussian distribution formula-
402 tion (dashed lines). Further, Figure 7b shows the variation of the normalized standard error for four
403 values of $\mathcal{L}/T_s = 100, 500, 1000$ and 3000 for all W_A values. Our results suggest that the required
404 length of data to obtain stable Lagrangian statistics (with normalized RMSE $< 25\%$) of intermit-
405 tent wave-breaking-induced turbulence using freely drifting platforms should be at least 1000 and
406 3000 characteristic wave breaking periods for medium to high sea states (e.g., $W_A > 5 \times 10^{-3}$) and
407 low sea states (e.g., $W_A < 5 \times 10^{-3}$) respectively. In other words, more Lagrangian sampled data

408 is needed to obtain stable statistics of wave breaking turbulence as the probability of breaking or
409 W_A decreases.

410 Assuming the characteristic wave breaking period of sea waves as $T_p/2$ (see §2.c), the minimum
411 required length of data to perform averaging will be $\mathcal{L}_{min} = 500T_p$ and $1500T_p$ for $W_A > 5 \times 10^{-3}$
412 and $W_A < 5 \times 10^{-3}$ respectively. In other words, \mathcal{L}_{min} varies between 1 to 3 hours depending on a
413 particular sea state. Brumer et al. (2017) and Callaghan and White (2009) found that whitecap data
414 collected over approximately 20 to 30 minutes are needed to reduce uncertainty related to average
415 whitecap coverage values. We note that \mathcal{L}_{min} may be obtained from one or multiple drifters,
416 with a sampling time of each drifter should be smaller than the time-scale during which the wind
417 forcing condition can be assumed as constant. We will show in §4 that considering \mathcal{L}_{min} noticeably
418 decreases the amount of scatter in the previous observations of the total TKE dissipation rates.

419 *c. Effects of occlusion due to the entrained bubbles and truncated vertical sampling*

420 We know from previous numerical (Derakhti and Kirby 2014a) and laboratory (Blenkinsopp
421 and Chaplin 2007) studies that the most active region of turbulence generation and dissipation
422 include relatively large air bubble void fractions. Figure 8a shows the distribution of the number
423 of the simulated data points sampled by the virtual drifters across dissipation rate and bubble void
424 fraction bins, that are uniformly spaced in log scale, for the breaking event E3 of the case T1. The
425 results indicate that $\alpha^b > 1\%$ in a noticeable portion of regions with relatively high ε_{sgs} values.
426 Figure 8b shows examples of the variation of the fraction of the total dissipation within the regions
427 with $\alpha^b < \alpha_0^b$ against α_0^b . In other words, although the high dissipation regions occur in a short
428 portion of time ($\sim W_A$) but their contribution to the total dissipation is large. The results shown in
429 Figure 8 reveal that approximately half of the total dissipation occur in regions with $\alpha^b > 1\%$.

430 Void fractions above 1% significantly decrease the quality of the data collected by acoustic
 431 Doppler methods by decreasing the correlation of coherent pulses (Mori et al. 2007). As a result,
 432 a large portion of high dissipation rate values (Figure 8b) in the observed data are occluded by
 433 bubbles. Figure 9a shows the comparison between the vertical profiles of averaged dissipation
 434 rates obtained by (solid lines) a regular ensemble-time-averaging defined in (10) and (dashed lines)
 435 a conditional averaging over data points with $\alpha^b < 1\%$ for the two cases with different W_A values.
 436 Although the effect of the occlusion due to bubbles is limited to the breaking surface layer $|z_{sf}| <$
 437 $0.6H_{eq}$, such data occlusion results in a considerable under-prediction of the total wave breaking
 438 dissipation rates in field observations using acoustic Doppler methods. Further, a limited vertical
 439 extend of sampled data by drifters causes the underestimation of the total dissipation rates as well.
 440 If we assume that a drifter can only sample the TKE dissipation rates in regions with $\alpha^b < \alpha_0^b$
 441 and up to a depth $|z_0|$, then the incomplete observed wave-breaking-induced TKE dissipation rates,
 442 $\int_{z_0}^0 [\bar{\varepsilon} - \bar{\varepsilon}_0]_{\alpha^b < \alpha_0^b} dz_{sf}$, will be always less than their corresponding true values $\int [\bar{\varepsilon} - \bar{\varepsilon}_0] dz_{sf}$ (Eq.
 443 12).

444 Figure 9b shows examples of the variation of a correction factor $\mathcal{C} > 1$ defined as

$$\mathcal{C} = \frac{\int [\bar{\varepsilon} - \bar{\varepsilon}_0] dz_{sf}}{\int_{z_0}^0 [\bar{\varepsilon} - \bar{\varepsilon}_0]_{\alpha^b < \alpha_0^b} dz_{sf}}, \quad (13)$$

445 with the depth and the active whitecap coverage W_A for $\alpha_0^b = 1\%$. As expected, \mathcal{C} is an increasing
 446 function of W_A and a decreasing function of $|z_0|$. However, \mathcal{C} has a very weak correlation with $|z_0|$
 447 for $|z_0| > 0.6H_{eq}$ (or below a breaking surface layer). Based on our numerical results, we obtain a
 448 simple relationship to predict \mathcal{C} as a function of W_A and $|z_0|/H_{eq}$:

$$\mathcal{C} \approx \mathcal{C}_b Z^\gamma \quad (14)$$

449 where $Z = \text{Min}(|z_0|/H_{eq}, 0.6) + 0.4$,

$$\mathcal{C}_b = c_1 \log_{10} W_A + c_2, \quad (15)$$

450 and

$$\gamma = d_1 \log_{10} W_A + d_2. \quad (16)$$

451 Here the empirical coefficients c_1, c_2, d_1 , and d_2 are obtained for a particular choice of α_0^b and by
452 using the least squared curve fitting. Table 2 documents these parameters for $\alpha_0^b = 1/3, 1$, and 3%.

453 The two line segments with markers shown in Figure 9b represent the corresponding fits given
454 in Eqs.(14-16) (here $\alpha_0^b = 1\%$) for the smallest and largest W_A values of the simulated cases. We
455 conclude that the empirical formulations Eqs. (14-16) provide a fairly reasonable approximation
456 for the correction factor \mathcal{C} for the range of W_A considered here.

457 4. Discussion

458 These results (§3.b and §3.c) improve interpretation of observed long-time-averaged total wave-
459 breaking-induced TKE dissipation rates, $\int [\bar{\epsilon} - \bar{\epsilon}_0] dz_{sf}$. However, they are purely based on the
460 idealized wave forcing and boundary conditions. Further, there is some uncertainty in generalizing
461 our results to a broader range of field observations, *e.g.*, using the correction factor \mathcal{C} (Eqs. 14-16).

462 Proper application of the model results to field data relies on the connection between the model
463 active whitecap coverage defined in Eq. (9) and field estimates of active whitecap coverage using
464 image processing techniques.

465 Here, we examine the variation of the corrected total wave-breaking-induced TKE dissipation
466 rates

$$\int [\bar{\epsilon} - \bar{\epsilon}_0] dz_{sf} \approx \mathcal{C} \int_{z_0}^0 [\bar{\epsilon} - \bar{\epsilon}_0]_{\text{Obs}} dz_{sf} \quad (17)$$

467 against the corresponding observed whitecap coverage W provided by Schwendeman and Thom-
468 son (2015) and the rate of wind energy input examined by Thomson et al. (2016). Schwende-
469 man and Thomson (2015) provided 119 data points including the whitecap coverage W estimated
470 from shipboard video systems, the total observed TKE dissipation rate $\int_{z_0}^0 [\bar{\epsilon} - \bar{\epsilon}_0]_{\text{Obs}} dz_{sf}$ with

471 $|z_0| = 0.42\text{m}$, wave spectra parameters, wind speed U_{10} and air-side friction velocity u_* . Here
 472 $\int_{z_0}^0 [\bar{\varepsilon} - \bar{\varepsilon}_0]_{\text{Obs}} dz_{sf}$ is the observed TKE dissipation rate values above a background level and is
 473 collected from the freely drifting SWIFT drifters (Thomson 2012). Following the results pre-
 474 sented in §3.b, we also consider clustering of the observed data to obtain averaged values over at
 475 least one hour, in addition to applying the correction \mathcal{C} on the observed data.

476 To estimate \mathcal{C} , we need to estimate the active whitecap coverage W_A from the observations of W
 477 provided by Schwendeman and Thomson (2015), where $W_A = \lambda W$ and λ ranges from 0.5 ± 0.25
 478 to 0.2 ± 0.1 in moderate to high sea states (Scanlon and Ward 2016, Figure 6b). We use a fit,
 479 provided by Callaghan (2018), to the bin-averaged λ measurements of Scanlon and Ward (2016)
 480 given by

$$\lambda = \frac{1}{1 + 8.65 [0.001U_{10}^2 + 0.02]^{0.69}}, \quad (18)$$

481 in which the estimated λ values decreases from ≈ 0.5 for $U_{10} = 5 \text{ m/s}$ to ≈ 0.15 for $U_{10} = 23 \text{ m/s}$
 482 (for further details see Callaghan 2018, Figure 5a and the related text therein).

483 a. Observed TKE dissipation rates versus active whitecap coverage

484 Figure 10 shows the variation of $\int_{z_0}^0 [\bar{\varepsilon} - \bar{\varepsilon}_0]_{\text{Obs}} dz_{sf}$ (denoted by small circles) and $\mathcal{C} \int_{z_0}^0 [\bar{\varepsilon} -$
 485 $\bar{\varepsilon}_0]_{\text{Obs}} dz_{sf}$ (denoted by small diamonds) with W (panel a) and with $W_A = \lambda W$ (panel b) for all
 486 available data points, where ε_0 represents turbulence dissipation by any mechanisms other than
 487 visible whitecaps, e.g., micro-breakers, Langmuir circulations, internal breaking waves, shear
 488 production, etc. In general, the background turbulence dissipation rate per unit surface area (or
 489 $\int [\bar{\varepsilon}_0]_{\text{Obs}} dz_{sf}$) is a sea state dependent quantity, and may vary between $O(0.01) \text{ W/m}^2$ (Hwang
 490 and Sletten 2008) and $O(0.1) \text{ W/m}^2$ (Gemmrich 2010; Sutherland and Melville 2015). Here, we
 491 take $\int_{z_0}^0 [\bar{\varepsilon}_0]_{\text{Obs}} dz_{sf} = 3.2 \times 10^{-4} [\text{m}^3 \text{s}^{-3}] (\approx 0.3 \text{ W/m}^2)$ which is approximately 0.9 of the mini-
 492 mum observed $\int_{z_0}^0 [\bar{\varepsilon}]_{\text{Obs}} dz_{sf}$ value reported by Schwendeman and Thomson (2015). Applying our

493 correction factor \mathcal{C} (Eqs. 14-16) significantly improves the results, such that the variation of the
494 total TKE dissipation rates increases approximately one order of magnitude by increasing W from
495 $\approx 10^{-3}$ to $\approx 3 \times 10^{-2}$. Such strong correlation between $\int [\bar{\epsilon} - \bar{\epsilon}_0] dz_{sf}$ and W , or W_A , is consis-
496 tent with our simulation results presented in Figure 5a and with previous semi-empirical and field
497 studies (Hwang and Sletten 2008; Anguelova and Hwang 2016; Callaghan 2018).

498 Based on the results shown in §3.b, part of the scatter in the data shown in Figure 10 (small
499 symbols) may be related to an insufficient record length \mathcal{L} (here $\mathcal{L} = 512s \approx 40 - 80T_p$) used to
500 perform the averaging, especially for W values smaller than 5×10^{-3} . We first sort the data points
501 corresponding to developing ($c_p \leq U_{10}$) and developed ($c_p > U_{10}$) sea states into U_{10} bins with a
502 spacing of 1 m/s. Then we remove the data points in which the measured W values vary from the
503 threshold power law fit provided by Callaghan et al. (2008) by more than a factor of 3. Finally, the
504 associated average values of different parameters are calculated using the remaining data points at
505 each bin with enough data points such that $\mathcal{L} > 400\hat{T}_p$, hereafter $(\hat{\cdot})$ represent clustered averaging
506 within each U_{10} bin.

507 Performing the clustered averaging described above on the data set of Schwendeman and Thom-
508 son (2015) results in seven clustered data points with $\hat{U}_{10} = 8.2, 8.8, 10.0, 11.2, 15.0$ and 15.9 m/s,
509 all characterized as a developed sea state. The corrected and raw clustered averaged values are
510 denoted by large symbols in Figure 10, where the size and color of the corrected data points repre-
511 sent their relative \mathcal{L}/\hat{T}_p and the wave age values respectively. Further, the range of the observed
512 W values for each bin is denoted by the vertical line segment in Figure 10a. The horizontal line
513 segments represent sensitivity of the results with respect to $0.1 < \lambda < 0.75$. Finally, the small
514 gray diamonds show the sensitivity of the corrected clustered-averaged total dissipation rates with
515 respect to the variation of the resulting correction factor \mathcal{C} for the choice of $\alpha_0^b = 1/3\%$ and 3% .

516 Using least-squared curve fitting, and assuming a power-law relationship

$$\int [\bar{\varepsilon} - \bar{\varepsilon}_0] dz_{sf} = a'_1 W^{b'_1} = a_1 W_A^{b_1}, \quad (19)$$

517 we obtain $a'_1 = 0.026(-0.012, +0.022)$ and $b'_1 = 0.77(\pm 0.12)$ (with $R^2 = 0.98$) and $a_1 =$
518 $0.24(-0.17, +0.58)$ and $b_1 = 0.98(\pm 0.2)$ (with $R^2 = 0.97$), where coefficients in parentheses
519 represent 95% confidence intervals. Both our simulation results (Figure 5a) and the corrected
520 clustered-averaged observations (Figure 10b) suggest that the total wave-breaking-induced TKE
521 dissipation rates $\int [\bar{\varepsilon} - \bar{\varepsilon}_0] dz_{sf}$ have a power law dependence with W_A with the exponent slightly
522 less than 1. Thus, the empirical relationship between total dissipation and whitecap coverage is
523 approximately linear.

524 Dynamical explanations for the whitecap coverage dependence are proposed by Callaghan
525 (2018), who scales dissipation rates with the volume of bubble plumes caused by breaking waves
526 (and thereby the active whitecap coverage and bubble plume penetration depth). In particular,
527 results with a fixed averaged bubble penetration depth in Figure 8 of Callaghan (2018) show a
528 similar dependence in comparison to Figures 5a and 10b in the present work. Anguelova and
529 Hwang (2016) also demonstrate a relation between active whitecap coverage and total wave break-
530 ing dissipation rates. We further comment on this in the next section. Quantification of averaged
531 penetration depth of bubble plumes relative to active whitecap areas is left for future study.

532 *b. Dissipation scaling and the distribution of breaking crests*

533 Many previous studies (e.g., Melville and Matusov 2002; Thomson and Jessup 2009; Romero
534 et al. 2012; Sutherland and Melville 2013) have applied the Phillips (1985) framework to estimate
535 an energy dissipation rate using the fifth moment of the breaking crest-length distribution $\Lambda(c)$,

536 where c is the crest speed, as

$$\int [\bar{\mathcal{E}} - \bar{\mathcal{E}}_0] dz_{sf} = \frac{\rho b}{g} \int c^5 \Lambda(c) dc. \quad (20)$$

537 Although the present study lacks measurements of $\Lambda(c)$, the results herein are still relevant to
538 the scaling of the breaking dissipation rate and the breaking strength parameter b . Beginning
539 with the laboratory work of Drazen et al. (2008), the emerging literature suggests a dependence
540 $b \sim (Ak)^{5/2}$, where Ak is the wave steepness given by amplitude A and wavenumber k . Romero
541 et al. (2012) extended this from the steepness of wave packets in the laboratory to the spectral
542 steepness, such that dissipation could be prescribed in a spectral wave model. Zappa et al. (2016)
543 recently reviewed the published results on the breaking strength parameter b .

544 Although we do not evaluate a spectral dissipation rate or the breaking strength with the present
545 analysis, we can attempt to reconcile the $\Lambda(c)$ framework with the relationship between the total
546 dissipation rate and active whitecap coverage. Kleiss and Melville (2010) relate active whitecap
547 coverage W_A to the first moment of $\Lambda(c)$ and a timescale for the persistence of the breaking crest
548 τ ,

$$W_A = \int c \tau \Lambda(c) dc. \quad (21)$$

549 If the timescale τ is proportional to breaking wave period T , then dispersion implies it is propor-
550 tional to phase speed and the effective relation is active whitecap coverage and the second moment,
551 $W_A = \int c^2 \Lambda(c) dc$. Kleiss and Melville (2010) evaluated these formulations, along with the zeroth
552 moment of $\Lambda(c)$, and find that all have strong linear relations ($R^2 > 0.96$) to the observed active
553 whitecap coverage (their Figure 7). If these lower moments of $\Lambda(c)$ are all similarly related to W_A ,
554 we can expect higher moments to be related as well. Generally, for a narrow distribution, higher
555 moments have a quasi-linear relationship to lower moments, because the tail of the distribution
556 is sufficiently small as to have minimal effect. Given the canonical tail of $\Lambda(c) \sim c^{-6}$ (Melville

557 and Matusov 2002), the distribution is indeed narrow and the net dependence of total dissipation
558 is closer to c^{-1} .

559 The implications for spectral dissipation remain to be determined, but it is thus at least empiri-
560 cally consistent for both active whitecap coverage W_A (Figures 5a and 10b) and $\int c^5 \Lambda(c) dc$ to be
561 related to the total wave breaking dissipation rate.

562 *c. Observed TKE dissipation versus wind energy input rates*

563 In an equilibrium sea state, the rate of wind energy input per unit area to the upper ocean F
564 ($\text{m}^3 \text{s}^{-3}$ or W/kg) is balanced mainly by the wave breaking energy dissipation. Figure 11 demon-
565 strates the significance of applying our correction factor \mathcal{C} to the data from Thomson et al. (2016)
566 in observing this expected equilibrium balance. In addition, the results show that clustering of
567 the individual data points (to achieve long enough record lengths \mathcal{L}) noticeably reduces the scat-
568 ter. Here, we use the formulation $F = c_e u_*^2 \rho^a / \rho$, where ρ^a is a constant air density. Other for-
569 mulations for the wind input rate give similar results (see Thomson et al. (2016)). The vertical
570 and horizontal line segments represent sensitivity of the results with respect to $2 < c_e < 3$ and
571 $0.1 < \lambda < 0.75$ respectively. As in Figure 10, the small gray diamonds show the sensitivity of the
572 corrected clustered-averaged total dissipation rates with respect to the variation of the resulting
573 correction factor \mathcal{C} for the choice of $\alpha_0^b = 1/3\%$ and 3% .

574 Given equilibrium conditions in which wind input and breaking dissipation rates balance, it is
575 not surprising that whitecap coverage has a nearly linear relationship to dissipation rate. Both
576 whitecap coverage and wind input have been regularly related to the cube of the wind speed (e.g.,
577 Brumer et al. (2017)) or the cube of wind friction velocity (e.g., Craig and Banner (1994)). The
578 implied empirical dependence between these parameters is thus linear, with dynamic interpretation
579 still an open question.

580 **5. Summary**

581 A high-resolution two-fluid LES/VOF numerical model (Derakhti and Kirby 2014a) represent-
582 ing breaking waves and turbulence is used to show that robust estimates of average turbulence
583 dissipation rates are possible from sparse Lagrangian sampling in a surface-following reference
584 frame (as done with field observations). Bubbles are treated as a multi-component continuum,
585 with different components representing different bubble diameters. Turbulence is modeled using
586 LES with a dynamic Smagorinsky closure. Bubble contributions to dissipation and momentum
587 transfer between the water and air phases are considered. Numerical simulations are run for many
588 wave periods to build up quasi-steady background turbulence levels, with breaking events occur-
589 ring approximately every $10T$, where T is the wave period. We sample the LES/VOF model
590 results with a large number of virtual surface-following drifters that are initially distributed in the
591 numerical domain, regularly or irregularly, before each breaking event. Time-averaged Lagrangian
592 statistics are obtained using the time-series sampled by the virtual drifters.

593 Convergence of statistics occurs for signals that have minimum length of approximately $1000T$
594 with randomly spaced observations in time and space relative to 3-D breaking events. This result
595 holds over a wide range of relative breaking activity, which is scaled in the model domain to match
596 field observations of whitecap coverage. The average dissipation rates have a nearly linear relation
597 to whitecap coverage. The model results also indicated that the high turbulence dissipation rates
598 are correlated with bubble plumes (and thus high void fractions). Using a canonical cutoff of
599 0.01 void fraction ($\alpha^b = 1\%$) for field observations of turbulence, an empirical correction factor
600 $\mathcal{C} = \mathcal{C}(W_A, |z_0|/H_{eq})$ is developed and applied to the previous observations of Thomson et al.
601 (2016); where W_A is the active whitecap coverage, $|z_0|$ is the extent of measurements over depth,
602 and H_{eq} is a characteristic breaking wave height.

603 Applying the correction factor to observations significantly alters the estimations of average
604 turbulence dissipation rates sampled by surface following drifters, especially in high sea states,
605 and thus, improves the inferred energy balance of wind input rates and turbulence dissipation
606 rates. Finally, both our simulation results and the corrected observations suggested that the total
607 wave breaking dissipation rates have a nearly linear relation with active whitecap coverage.

608 We emphasize that the proposed correction factor is based purely on numerical simulations of a
609 limited number of idealized wave breaking events, in which a number of relevant processes such as
610 direct wind forcing have been ignored. In the absence of new field methods for direct observation
611 of turbulence inside bubble plumes, applying the proposed correction factor to the open ocean
612 conditions must be made cautiously. More field observations of near-surface turbulence and bubble
613 plumes are needed, especially in high sea states.

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616 from www.apl.washington.edu/swift and whitecap coverage data are available from
617 <http://hdl.handle.net/1773/42596>.

618 APPENDIX

619 **A Summary of All Mathematical Variables and Symbols Used in §3-§5**

620 Table A1 summarizes the symbols, definitions, and units for the variables used in the Results,
621 Discussion and Summary sections.

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761 **LIST OF TABLES**

762 Table 1.	Input parameters for the simulated short-crested (3-D) focused wave packets. 763 In all three cases $N = 10$, $\Delta f/f_c = 0.75$, $T_g/T_s \approx 9.5$, $x_f/L_s \approx 1.6 \sim 1.9$, $y_f = 0$, 764 and $t_f = 15.0$ s. Definitions of all parameters presented here are given in §2.b.	37
765 Table 2.	The empirical coefficients in Eqs. (15) and (16) for three values of α_0^b (%) 766 obtained using the least squared curve fitting. Coefficients in parentheses rep- 767 resent 95% confidence intervals.	38
768 Table A1.	Summary of mathematical variables and symbols used in §3-§5. Here [-] indicates that the 769 corresponding variable is dimensionless. The order of the symbols are consistent with the 770 order of their first appearance in §3-§5.	39

Case no.	S_g	f_c	T_s	L_s	h/L_s	L_x/L_s	$L_y/2L_s$
		(1/s)	(s)	(m)			
T1	0.44	0.7	1.8	5.1	0.59	5	0.8
T2	0.32	0.7	1.8	5.1	0.59	5	0.8
T3	0.40	0.9	1.4	3.0	0.42	4	1.0

771 TABLE 1. Input parameters for the simulated short-crested (3-D) focused wave packets. In all three cases
 772 $N = 10$, $\Delta f/f_c = 0.75$, $T_g/T_s \approx 9.5$, $x_f/L_s \approx 1.6 \sim 1.9$, $y_f = 0$, and $t_f = 15.0$ s. Definitions of all parameters
 773 presented here are given in §2.b.

α_0^b	c_1	c_2	d_1	d_2
1/3	0.67 (± 0.21)	4.30 (± 0.50)	-0.89 (± 0.15)	-3.00 (± 0.3)
1	0.48 (± 0.12)	3.15 (± 0.26)	-0.79 (± 0.13)	-2.90 (± 0.3)
3	0.26 (± 0.12)	2.20 (± 0.25)	-0.73 (± 0.12)	-2.84 (± 0.3)

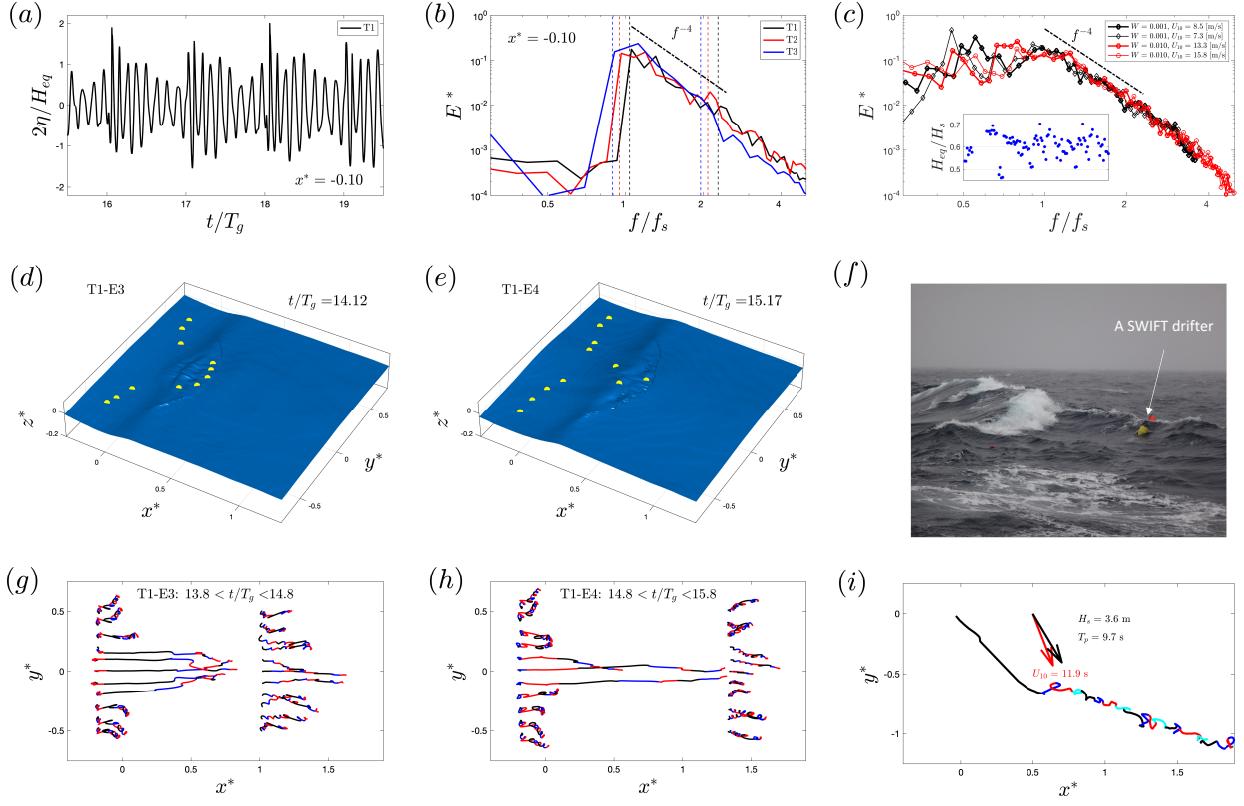
774 TABLE 2. The empirical coefficients in Eqs. (15) and (16) for three values of α_0^b (%) obtained using the least
 775 squared curve fitting. Coefficients in parentheses represent 95% confidence intervals.

Table A1. Summary of mathematical variables and symbols used in §3-§5. Here [-] indicates that the corresponding variable is dimensionless. The order of the symbols are consistent with the order of their first appearance in §3-§5.

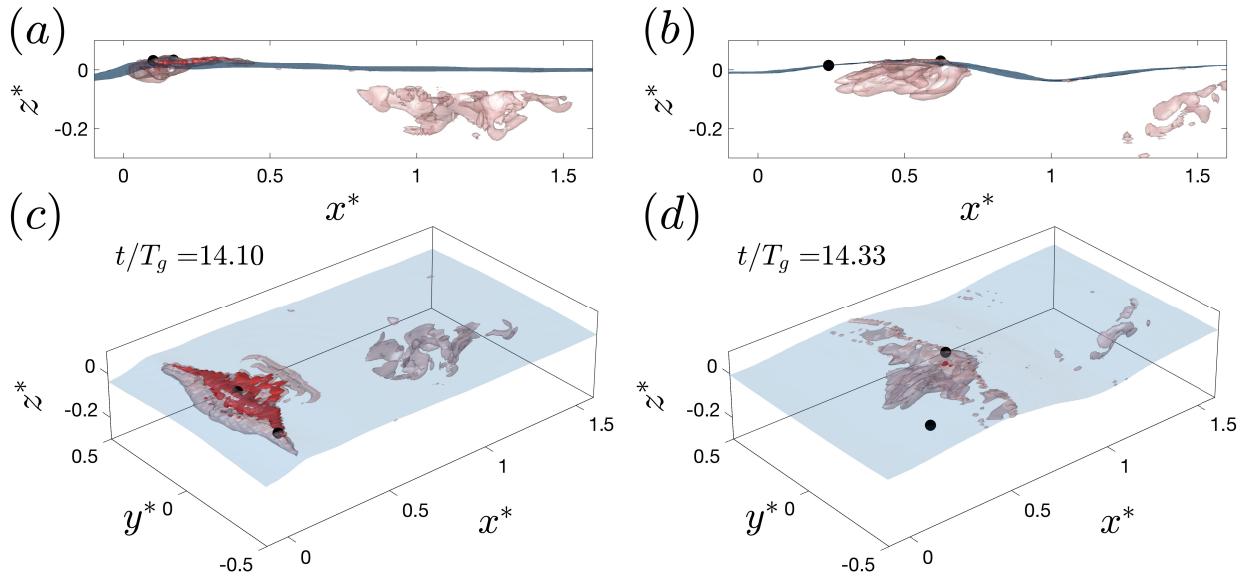
Symbol	Definition	Units
ϵ	TKE dissipation rate	[m ² s ⁻³]
ν_{sgs}	SGS eddy viscosity	[m ² s ⁻¹]
ϵ_{sgs}	Energy transfer rate from the resolved to SGS motions	[m ² s ⁻³]
$\check{\epsilon}_{sgs}$	Horizontal average of ϵ_{sgs} in a surface-following reference frame	[m ² s ⁻³]
$\bar{\epsilon}_{sgs}$	Time-averaged ϵ_{sgs} over a few wave periods obtained from one Lagrangian virtual drifter	[m ² s ⁻³]
$\bar{\bar{\epsilon}}^L$	Time-averaged ϵ_{sgs} over many wave periods obtained from all the Lagrangian virtual drifters	[m ² s ⁻³]
$\bar{\bar{\epsilon}}_n^L$	Time-averaged ϵ_{sgs} obtained from n randomly selected Lagrangian virtual drifters	[m ² s ⁻³]
$\bar{\bar{\epsilon}}^E$	Time-averaged $\check{\epsilon}_{sgs}$ over many wave periods	[m ² s ⁻³]
f_s	Spectrally-weighted frequency of a wave field (Eq. 6)	[s ⁻¹]
f_p	Peak frequency of a wave field	[s ⁻¹]
T_s	$= 1/f_s$	[s]
T_p	$= 1/f_p$	[s]
T_g	Wave packet period	[s]
L_s	Characteristic wave length calculated based on f_s	[m]
L_p	Characteristic wave length calculated based on f_p	[s]
η	Free surface elevation	[m]
z_{sf}	Vertical elevation in a surface-following reference frame	[m]
H_s	Significant wave height	[m]
H_{eq}	Characteristic wave breaking height (Eq. 7)	[m]
W	Whitecap coverage	[-]
W_A	Active whitecap coverage	[-]
λ	$= W_A/W$	[-]
U_{10}	10-m wind speed	[ms ⁻¹]
$\int \bar{\bar{\epsilon}} dz_{sf}$	$= \int \bar{\bar{\epsilon}} dz$, total energy dissipation rate	[m ³ s ⁻³]
$\int \bar{\bar{\epsilon}}_0 dz_{sf}$	Total energy dissipation rate due to background processes	[m ³ s ⁻³]
\mathcal{L}	Record length	[s]
α^b	Bubble void fraction	[-]
α_0^b	Bubble void fraction threshold above which the data is assumed to be occluded by bubbles	[-]
\mathcal{C}	The correction factor defined in Eq. 13	[-]
\hat{T}_p	Clustered averaged T_p within U_{10} bins	[s]
\hat{U}_{10}	Clustered averaged U_{10}	[ms ⁻¹]
\hat{u}_*	Air side friction velocity	[ms ⁻¹]
c_p	$= L_p/T_p$	[ms ⁻¹]
c	Wave phase speed	[ms ⁻¹]
b	Breaking strength parameter	[-]
Ak	Wave steepness	[-]
$\Lambda(c)$	Distribution of the breaking wave crest lengths per unit sea surface area per unit increment in velocity c	[m ⁻² s]
F	Wind energy input rate	[m ³ s ⁻³]

776 LIST OF FIGURES

- 778 **Fig. 1.** Comparison between model and field conditions. (a) Temporal variation of the normalized
 779 free surface elevations $2\eta/H_{eq}$ for the case T1 and (b) its normalized power spectral density
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 789 wave component. In (g, h) and (i), each color segment represents the horizontal displace-
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- 790 **Fig. 2.** Side-view (a, b) and 3-D view (c, d) of two snapshots of the spatial variation of the tur-
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 794 The waves propagate in the positive x direction. The patches of sub-surface ε_{sgs} in $x^* > 0.75$
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 797 case T1. (a) Phase-resolved horizontal-averaged turbulence dissipation rate $\bar{\varepsilon}_{sgs}$ in the
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 801 for all available virtual drifters released at before the beginning of the breaking event T1-E3
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 803 follow the law-of-the-wall vertical scaling ($\sim 1/z$). 44
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 818 lines in (a) show the normalized standard error estimation using the Gaussian distribution. 48



843 FIG. 1. Comparison between model and field conditions. (a) Temporal variation of the normalized free
 844 surface elevations $2\eta/Heq$ for the case T1 and (b) its normalized power spectral density $E^* = EH_{eq}^{-2}f_s$ for
 845 all three cases at $(x^*, y^*) = (-0.1, 0)$. (d, e) Snapshots of the spatial variation of the normalized free surface
 846 elevations and of the virtual drifters initially released at $x^* = -0.2$ for the breaking events E3 and E4 of the case
 847 T1. (g, h) The normalized horizontal displacements of the virtual drifters shown in (d, e). (c, f, i) Examples
 848 of E^* , Heq/hs , a SWIFT drifter, and the horizontal displacement of a SWIFT drifter from the previous field
 849 observations in the vicinity of OWS-P at 50° N, 145° W (Thomson et al. 2016). Vertical dashed lines in (b)
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 853 displacement during T_s and $T_p/2$ respectively.



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 857 ($x^* = 0, y^* = -0.3$) and ($x^* = 0, y^* = 0$). The waves propagate in the positive x direction. The patches of
 858 sub-surface ϵ_{sgs} in $x^* > 0.75$ correspond to the preceding breaking event.

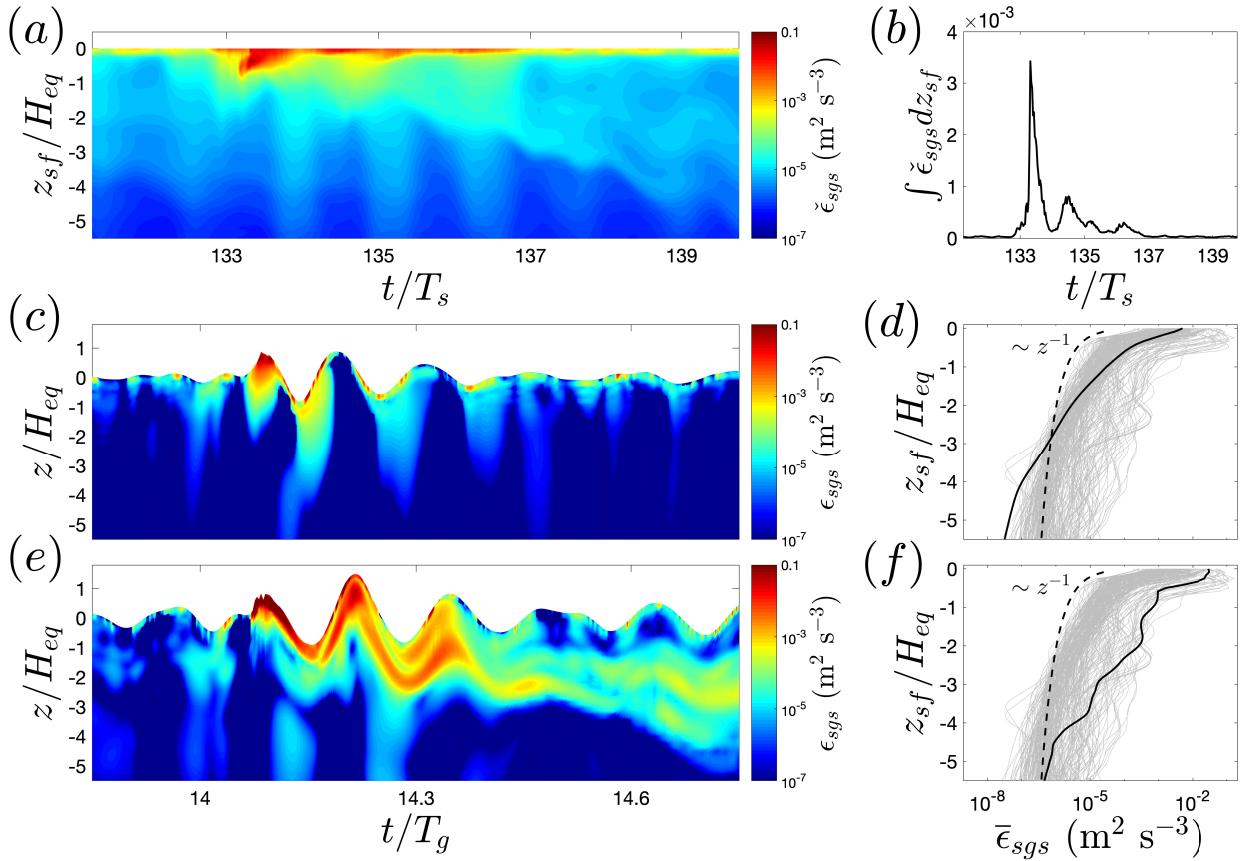
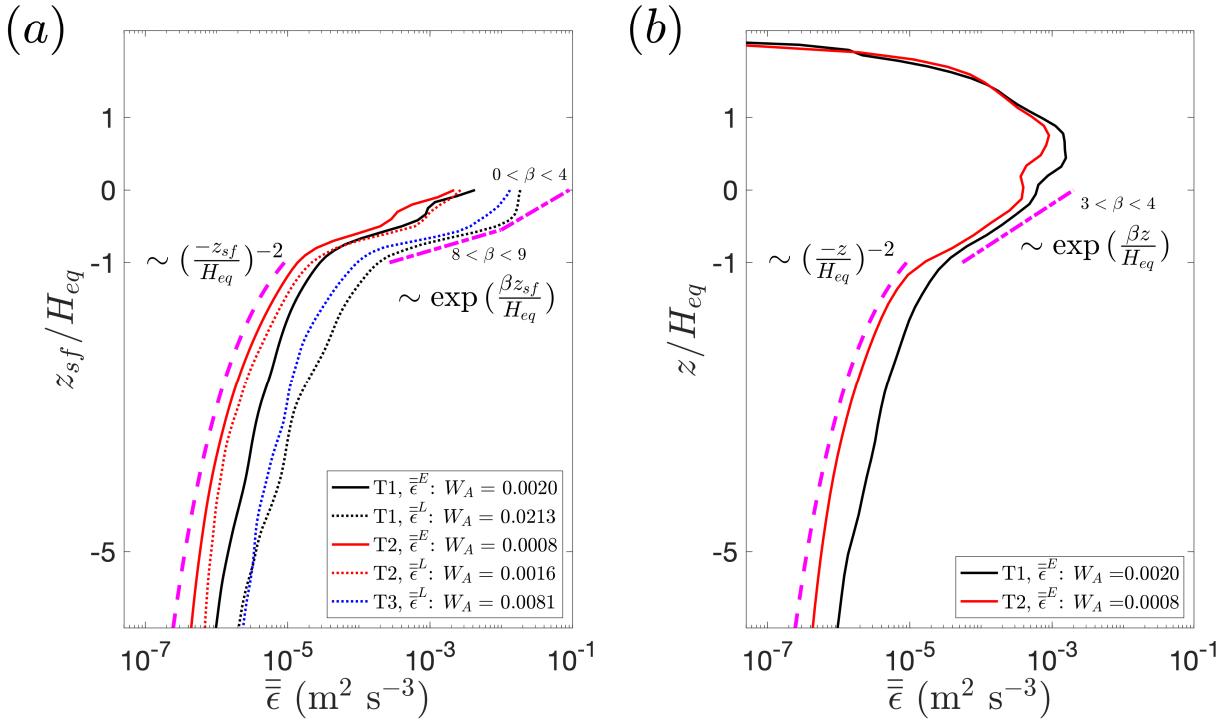
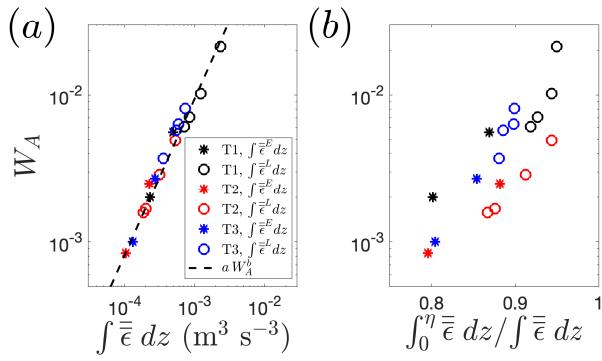


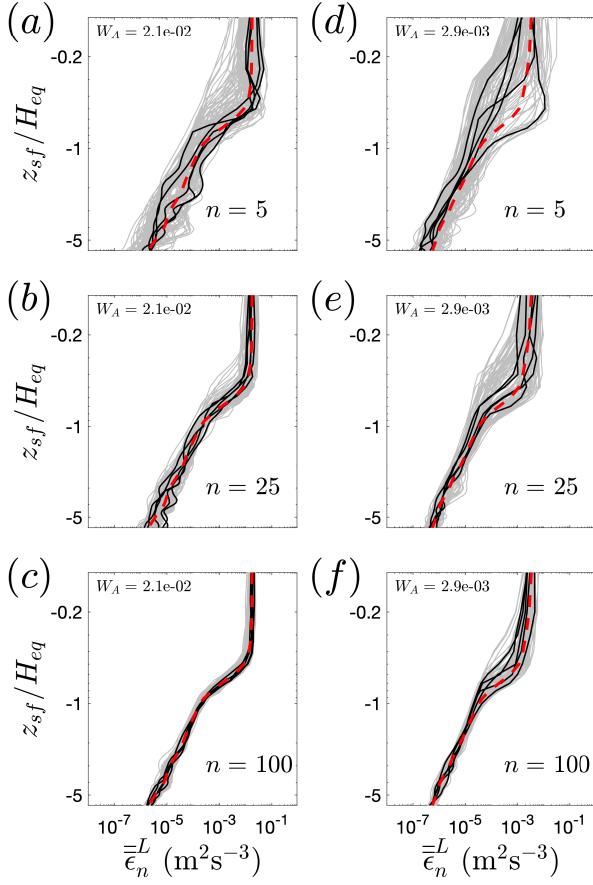
FIG. 3. Various measures of the turbulence dissipation rate ϵ_{sgs} for the breaking event E3 of the case T1. (a) Phase-resolved horizontal-averaged turbulence dissipation rate $\tilde{\epsilon}_{sgs}$ in the surface following reference frame, (b) temporal variation of the vertically integrated $\tilde{\epsilon}_{sgs}$, (c, e) phase-resolved $\tilde{\epsilon}_{sgs}$ sampled by the two virtual drifters shown in Figure 2, and (d, f) their corresponding time-averaged vertical profiles, $\bar{\epsilon}_{sgs}$ (thick black lines) as well as the results for all available virtual drifters released at before the beginning of the breaking event T1-E3 (thin gray lines). The dashed lines in panels (d) and (f) demonstrate that $\bar{\epsilon}_{sgs}$ does not follow the law-of-the-wall vertical scaling ($\sim 1/z$).



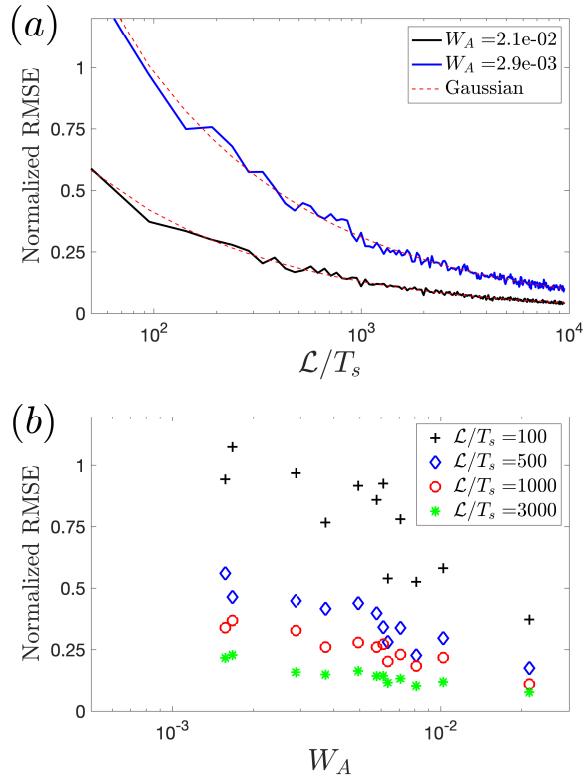
866 FIG. 4. Model results of the vertical profile of the ensemble-time-averaged turbulence dissipation rates $\bar{\epsilon}$ in
867 (a) the surface-following reference frame, and (b) the fixed reference frame. Here H_{eq} the equilibrium wave
868 height, W_A the active whitecap coverage, $\bar{\epsilon}^L$, and $\bar{\epsilon}^E$ are defined in Eqs. (7), (9), (10), and (11), respectively.



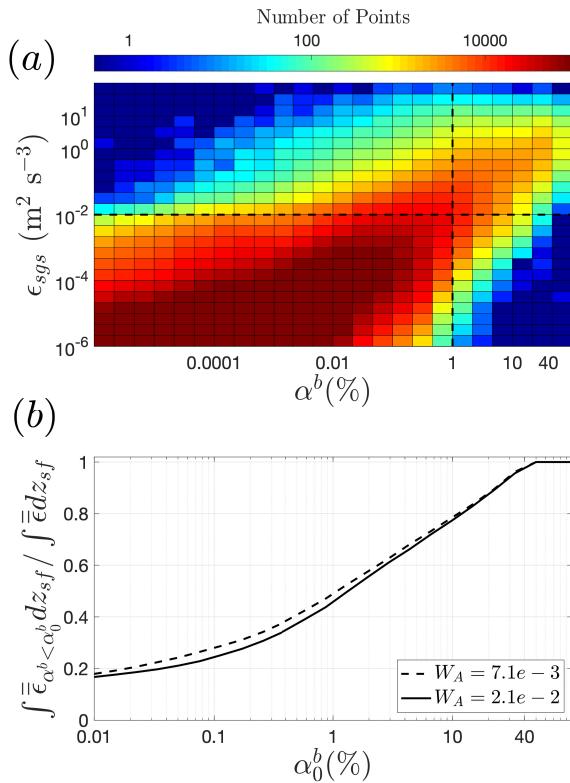
869 FIG. 5. Model results of the variation of (a) the vertically integrated long-time-averaged dissipation rates, and
 870 (b) the fraction of total dissipation above still water level $z = 0$ with the active whitecap coverage W_A (Eq. 9).



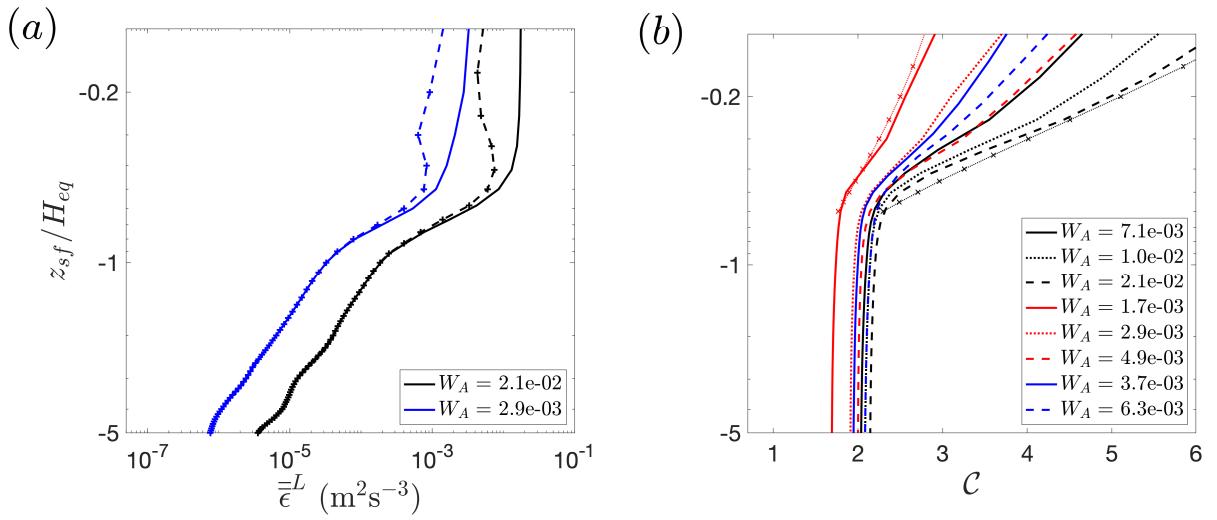
871 FIG. 6. Variation of the Lagrangian statistics of turbulence dissipation rates obtained from n randomly selected
 872 virtual drifters $\bar{\epsilon}_n^L$ (solid lines) with the signal length $\mathcal{L} \approx 9.5nT_s$ for cases with (a – c) $W_A = 2.1 \times 10^{-2}$, and
 873 (d – f) $W_A = 2.9 \times 10^{-3}$. For better visibility, five examples of $\bar{\epsilon}_n^L$ profiles are plotted in each frame as individual
 874 black curves. The red dashed lines show $\bar{\epsilon}^L$, which represents the average of all virtual drifters (Eq. 10).



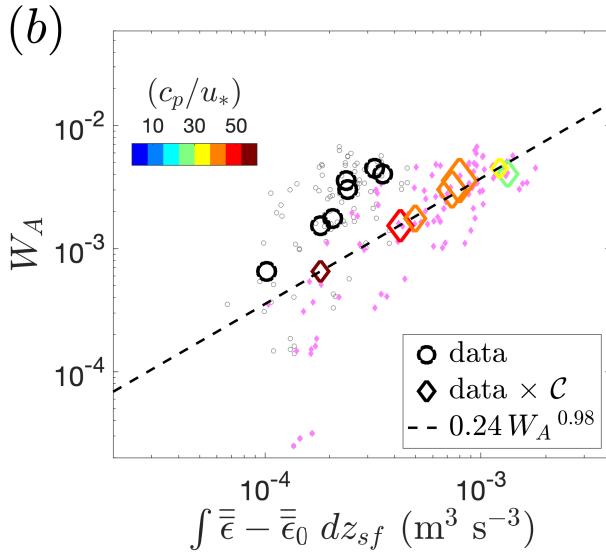
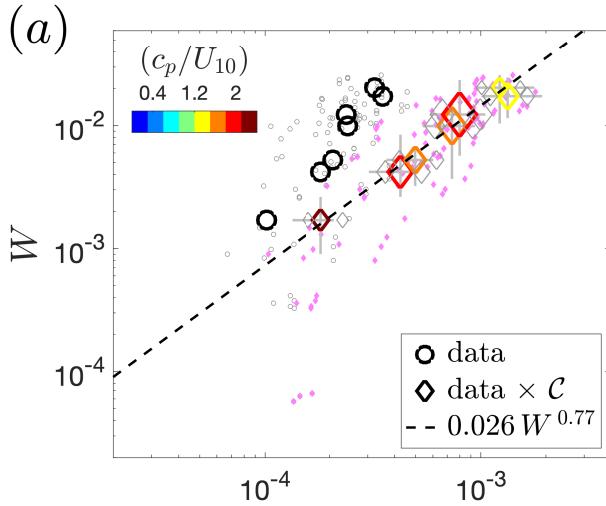
875 FIG. 7. Variation of the normalized standard error of total turbulence dissipation rate estimates (a) with
 876 the signal length $\mathcal{L}/T_s \approx 9.5n$, and (b) with active whitecap coverage W_A . The dashed lines in (a) show the
 877 normalized standard error estimation using the Gaussian distribution.



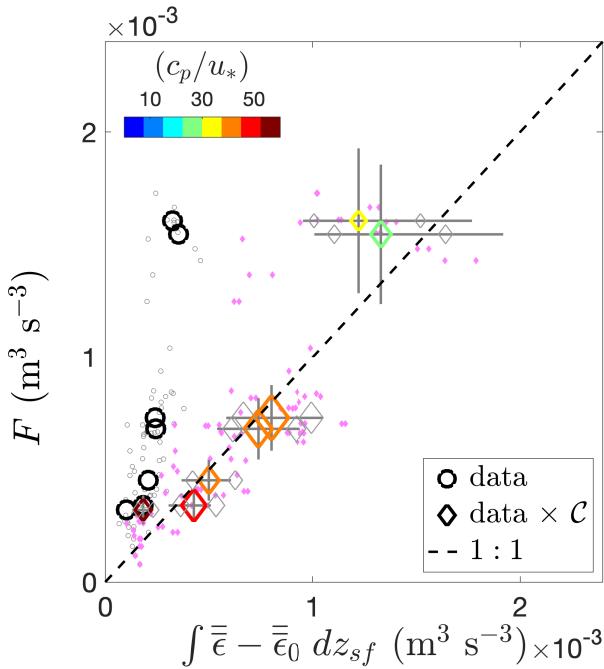
878 FIG. 8. (a) Example of a 2-D histogram of the model results of the local turbulence dissipation rates ϵ_{sgs} and
 879 bubble void fraction, and (b) two examples of the variation of the fraction of the total dissipation within the
 880 regions with $\alpha^b < \alpha_0^b$ against α_0^b . In (a), the vertical and horizontal dashed lines show $\alpha^b = 1\%$ and $\epsilon_{sgs} = 0.01$
 881 [$\text{m}^2 \text{s}^{-3}$], respectively.



882 FIG. 9. Incomplete sampling of turbulence dissipation rates. (a) Effect of the bubble occlusion on the La-
 883 grangian averages by considering (solid lines) all model results $\bar{\epsilon}^L$ (Eq. 10), and (dashed lines) only data points
 884 in which $\alpha^b < 1\%$. (b) Variation of the correction factor \mathcal{C} (Eq. 13) with depth and the active whitecap coverage
 885 W_A . In (b), the two line segments with markers show the fit defined in Eq. 14 for the smallest and largest W_A
 886 values of the simulated cases.



887 FIG. 10. Variation of the total wave-breaking-induced TKE dissipation rates with (a) total whitecap coverage
 888 W , and (b) active whitecap coverage $W_A = \lambda W$ (Eq. 18). Circles show the raw data provided by Schwendeman
 889 and Thomson (2015), and diamonds show the corrected results by applying the correction factor \mathcal{C} defined
 890 in Eqs. (14-16, $\alpha_0^b = 1\%$). Small and large symbols represent the results considering an averaging time of
 891 $\mathcal{L} = 512s \approx 40 - 80T_p$ and those obtained by a conditional clustered-averaging explained in the text with
 892 $\mathcal{L} > 400\hat{T}_p$. The horizontal line segments represent sensitivity of the results with respect to $0.1 < \lambda < 0.75$.
 893 The small gray diamonds show the sensitivity of the corrected clustered-averaged total dissipation rates with
 894 respect to the variation of the resulting correction factor \mathcal{C} for the choice of $\alpha_0^b = 1/3\%$ and 3% . Vertical line
 895 segments represent the range of the observed W values for each bin.



896 FIG. 11. Variation of the total wave-breaking-induced TKE dissipation rates with the rate of wind energy
 897 input F . Vertical line segments represent the sensitivity of F values with respect to $2 < c_e < 3$. Definitions of
 898 the rest of symbols and lines are the same as in Figure 10.