

Sparse sampling of intermittent turbulence generated by breaking surface

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 dissi-
73 pation 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

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

84 Here, we revisit the topics of Thomson et al. (2016) by sampling a high-fidelity numerical model
85 in the Lagrangian mode of the surface-following observations. We use a polydisperse two-fluid
86 model (Derakhti and Kirby 2014a) with large-eddy simulation (LES) resolution and volume-of-
87 fluid surface reconstruction (VOF) to simulate the generation and evolution of turbulence and bub-
88 bles beneath 3-D short-crested wave breaking events in deep water (§2). We first scale the model
89 domain to match the observed whitecap coverage values, and we scale the model wave heights to
90 match the wind-wave (*i.e.*, equilibrium) portion of the observed spectrum (*i.e.*, neglecting swell).
91 We then determine the effects of sparse sampling and intermittent breaking, as well as the effects
92 of data occlusion by bubbles and limitations in the vertical extent of the observed profiles (§3). In
93 §4, we comment on the apparent discrepancy between the observed wind-input energy fluxes and
94 total turbulence dissipation rates reported by Thomson et al. (2016). Examination of potential La-
95 grangian sampling bias related to a partially trapped drifter in convergence zones in the turbulence
96 observations is left for future work.

97 2. Methods

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

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

a. Mathematical formulations

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

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

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

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

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

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

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

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

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 velocity in the j direction of the k th bubble size class; ρ^b is the bubble density; and R_k^b includes the source due to air entrainment in the interfacial cells (Derakhti and Kirby 2014a, §2.3), intergroup mass transfer, and SGS diffusion terms. Finally, \mathbf{M}_k^{lg} represents the total momentum transfer 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 the inertia and shear stress terms in the gas phase following Carrica et al. (1999) and Derakhti and Kirby (2014a).

b. Model set-up

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

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

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

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

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

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

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

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

Each numerical case is defined by setting the geometry of the virtual tank and the input wave 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 $t_0 > 12T_g > 200s$ 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

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

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

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

The results demonstrate that both the simulated and observed E^* have a self-similar shape in the range $f_s < f < 2f_s$, which is usually called an equilibrium range of a wave spectrum (Phillips 1985). Second, the shape of the simulated wave spectrum for $f > f_s$ is similar to the observations. The f^{-4} dependence in the simulated spectra $E(f)$ is achieved by using the constant steepness 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 relation for deep water waves. However, in the field spectra there is much more energy at frequencies $f < f_s$ due to the presence of swell. Many field observations of the speed of visible whitecaps have shown that the dominant speed of breaking waves is about half the phase speed of waves at the peak of the energy spectrum (Melville and Matusov 2002; Gemmrich et al. 2008; Thomson and Jessup 2009; Kleiss and Melville 2010; Schwendeman et al. 2014; Sutherland and Melville 2013). Assuming linear dispersion, this means that the frequency associated with breaking waves f_b is 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

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

d. Conversion of the model results to virtual drifters

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

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

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

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

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

3. Results

A glossary of all variables used hereafter is given in Table A1 in the appendix. In our model results, the rate of transfer of energy from the resolved motions to the SGS motions is $\epsilon_{sgs} = \nu_{sgs} |\mathcal{S}|^2$ where ν_{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 resolved rate of strain. ϵ_{sgs} includes both shear- and bubble-induced dissipation (Derakhti and Kirby 2014a). At high Reynolds number, with the filter width much larger than the Kolmogorov length scale, the viscous dissipation rate is typically much smaller than ϵ_{sgs} , and thus ϵ_{sgs} approximates the turbulent kinetic energy (TKE) dissipation rate ϵ , which is commonly considered as a characteristic indicator of intensity of turbulence.

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

278 breaking simulations (see, for example, Derakhti and Kirby 2014a, Figures 11 and 12), our numer-
 279 ical results indicate that wave-breaking-induced ϵ_{sgs} has a strong temporal and spatial variation,
 280 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
 281 with large values of ϵ_{sgs} concentrated near the wave crest and in regions of high void fraction
 282 (bubble void fractions are not shown here). The latter is consistent with the recent laboratory mea-
 283 surements of turbulence dissipation rates ϵ within wave breaking crests by Deane et al. (2016).
 284 They reported large values of $\epsilon > 1 \text{ W/kg}$ during the acoustically active phase of wave breaking
 285 in which air is actively entrained and fragmented into bubbles.

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

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

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

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

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

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

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

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

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

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

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

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

Furthermore, in the fixed reference frame and within the surface breaking layer, $\bar{\bar{\epsilon}}$ is described 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 and 4 for the simulated cases. Our model results demonstrate that the Eulerian maximum of $\bar{\bar{\epsilon}}$ is above $z = 0$, where $\bar{\bar{\epsilon}}$ has persistently large values in $0 < z/H_{eq} < 1$ compared with those in other

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

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

Assuming a power-law relationship

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

where ϵ_0 is the background turbulence dissipation rate due to mechanisms other than visible wave breaking, $\epsilon_0 \approx 0$ in our simulations. Using least squared curve fitting, we obtain $a = 0.077(-0.026, +0.039)$ and $b = 0.94(\pm 0.07)$ (with $R^2 = 0.98$), where coefficients in parentheses represent 95% confidence intervals. This is an interesting and important result, and we will further comment on it in §4.

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

Finally, the results shown in Figure 4a and Figure 5 demonstrate that the Lagrangian statistics of intermittent wave breaking turbulence, obtained from the sampled data by freely drifting platforms, are representative of the corresponding Eulerian statistics when the length of the Lagrangian data is very large compared with the local wave breaking period. In the next section, we examine how 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 $\overline{\epsilon}_n^L$ compare to the associated statistics using all available virtual
 389 drifters $\overline{\epsilon}^L$ (Eq. 10). We note that the normalized length of the considered data \mathcal{L}/T_s used to
 390 obtain $\overline{\epsilon}_n^L$ and $\overline{\epsilon}^L$ are $\approx 9.5n$ and $\approx 9.5 \times 231N_E$ respectively.

391 Figure 6 shows $\overline{\epsilon}_n^L$ (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 $\overline{\epsilon}^L$ (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, $\overline{\epsilon}_n^L$ (solid lines) converges to $\overline{\epsilon}^L$ (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 $\int \overline{\epsilon}_n^L dz_{sf}$ (with mean value $\int \overline{\epsilon}^L dz_{sf}$)
 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 $\alpha^b > 1\%$.

Void fractions above 1% significantly decrease the quality of the data collected by acoustic Doppler methods by decreasing the correlation of coherent pulses (Mori et al. 2007). As a result, a large portion of high dissipation rate values (Figure 8b) in the observed data are occluded by bubbles. Figure 9a shows the comparison between the vertical profiles of averaged dissipation rates obtained by (solid lines) a regular ensemble-time-averaging defined in (10) and (dashed lines) a conditional averaging over data points with $\alpha^b < 1\%$ for the two cases with different W_A values. Although the effect of the occlusion due to bubbles is limited to the breaking surface layer $|z_{sf}| < 0.6H_{eq}$, such data occlusion results in a considerable under-prediction of the total wave breaking dissipation rates in field observations using acoustic Doppler methods. Further, a limited vertical extend of sampled data by drifters causes the underestimation of the total dissipation rates as well.

If we assume that a drifter can only sample the TKE dissipation rates in regions with $\alpha^b < \alpha_0^b$ and up to a depth $|z_0|$, then the incomplete observed wave-breaking-induced TKE dissipation rates, $\int_{z_0}^0 [\bar{\bar{\epsilon}} - \bar{\bar{\epsilon}}_0]_{\alpha^b < \alpha_0^b} dz_{sf}$, will be always less than their corresponding true values $\int [\bar{\bar{\epsilon}} - \bar{\bar{\epsilon}}_0] dz_{sf}$ (Eq. 12).

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

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

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

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

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, $\int [\bar{\mathcal{E}} - \bar{\mathcal{E}}_0] dz_{sf}$. 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

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

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 $\int_{z_0}^0 [\bar{\mathcal{E}} - \bar{\mathcal{E}}_0]_{\text{Obs}} dz_{sf}$ with

471 $|z_0| = 0.42\text{m}$, wave spectra parameters, wind speed U_{10} and air-side friction velocity u_* . Here
 472 $\int_{z_0}^0 [\bar{\epsilon} - \bar{\epsilon}_0]_{\text{Obs}} dz_{sf}$ 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

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

481 in which the estimated λ values decreases from ≈ 0.5 for $U_{10} = 5$ m/s to ≈ 0.15 for $U_{10} = 23$ 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 $\int_{z_0}^0 [\bar{\epsilon} - \bar{\epsilon}_0]_{\text{Obs}} dz_{sf}$ (denoted by small circles) and $\mathcal{C} \int_{z_0}^0 [\bar{\epsilon} -$
 485 $\bar{\epsilon}_0]_{\text{Obs}} dz_{sf}$ (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 $\int [\bar{\epsilon}_0]_{\text{Obs}} dz_{sf}$) is a sea state dependent quantity, and may vary between $O(0.01)$ W/m² (Hwang
 490 and Sletten 2008) and $O(0.1)$ W/m² (Gemmrich 2010; Sutherland and Melville 2015). Here, we
 491 take $\int_{z_0}^0 [\bar{\epsilon}_0]_{\text{Obs}} dz_{sf} = 3.2 \times 10^{-4} [\text{m}^3 \text{s}^{-3}]$ ($\approx 0.3 \text{W/m}^2$) which is approximately 0.9 of the mini-
 492 mum observed $\int_{z_0}^0 [\bar{\epsilon}]_{\text{Obs}} dz_{sf}$ value reported by Schwendeman and Thomson (2015). Applying our

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

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

Performing the clustered averaging described above on the data set of Schwendeman and Thomson (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, all characterized as a developed sea state. The corrected and raw clustered averaged values are denoted by large symbols in Figure 10, where the size and color of the corrected data points represent their relative \mathcal{L}/\hat{T}_p and the wave age values respectively. Further, the range of the observed W values for each bin is denoted by the vertical line segment in Figure 10a. The horizontal line segments represent sensitivity of the results with respect to $0.1 < \lambda < 0.75$. Finally, the small gray diamonds show the sensitivity of the corrected clustered-averaged total dissipation rates with respect to the variation of the resulting correction factor \mathcal{C} for the choice of $\alpha_0^b = 1/3\%$ and 3% .

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

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

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

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

b. Dissipation scaling and the distribution of breaking crests

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

536 where c is the crest speed, as

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

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 (m^3s^{-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.

5. Summary

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

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

TABLE 1. Input parameters for the simulated short-crested (3-D) focused wave packets. In all three cases $N = 10$, $\Delta f/f_c = 0.75$, $T_g/T_s \approx 9.5$, $x_f/L_s \approx 1.6 \sim 1.9$, $y_f = 0$, and $t_f = 15.0$ s. Definitions of all parameters 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]
$\tilde{\varepsilon}_{sgs}$	Horizontal average of ε_{sgs} in a surface-following reference frame	[m ² s ⁻³]
$\bar{\varepsilon}_{sgs}$	Time-averaged ε_{sgs} over a few wave periods obtained from one Lagrangian virtual drifter	[m ² s ⁻³]
$\overline{\varepsilon}^L$	Time-averaged ε_{sgs} over many wave periods obtained from all the Lagrangian virtual drifters	[m ² s ⁻³]
$\bar{\varepsilon}_n^L$	Time-averaged ε_{sgs} obtained from n randomly selected Lagrangian virtual drifters	[m ² s ⁻³]
$\overline{\varepsilon}^E$	Time-averaged $\tilde{\varepsilon}_{sgs}$ over many wave periods	[m ² s ⁻³]
f_s	Spectrally-weighted frequency of a wave filed (Eq. 6)	[s ⁻¹]
f_p	Peak frequency of a wave filed	[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]
$\int \bar{\varepsilon} dz_{sf}$	$= \int \bar{\varepsilon} dz$, total energy dissipation rate	[m ³ s ⁻³]
$\int \bar{\varepsilon}_0 dz_{sf}$	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]

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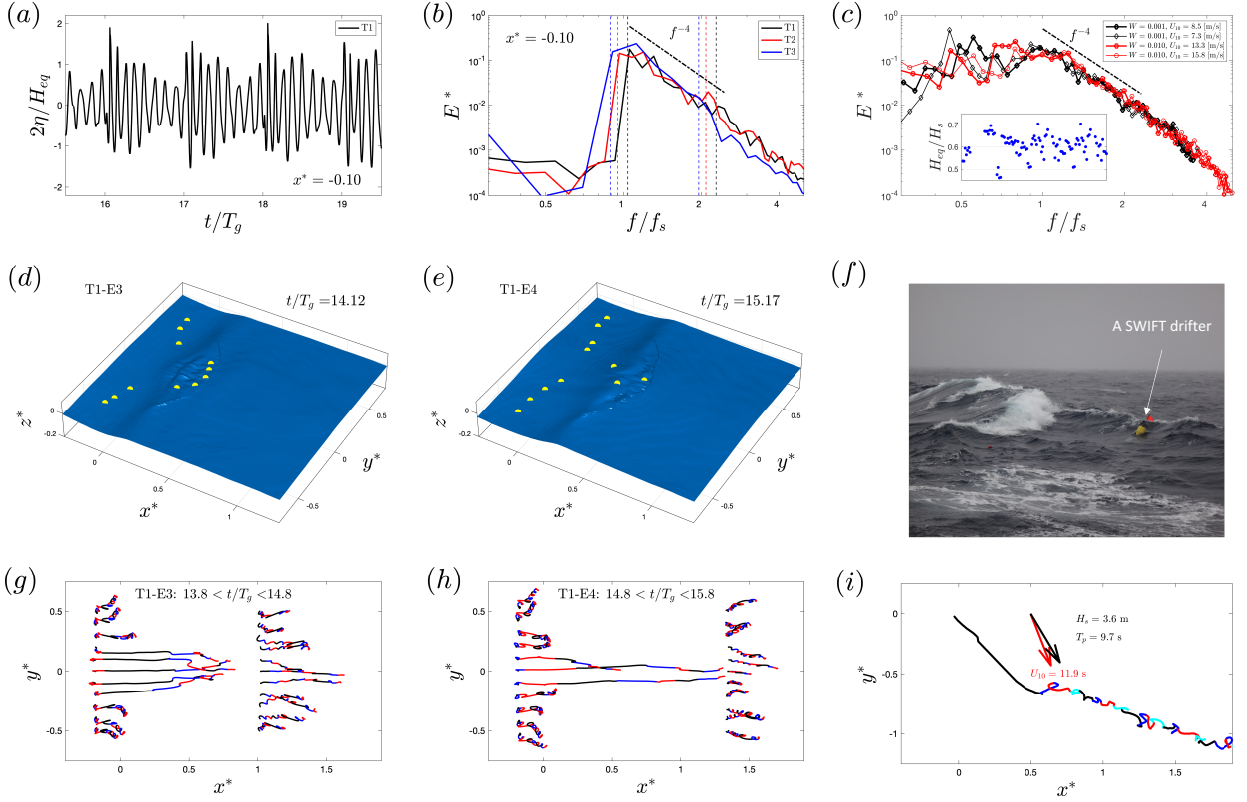


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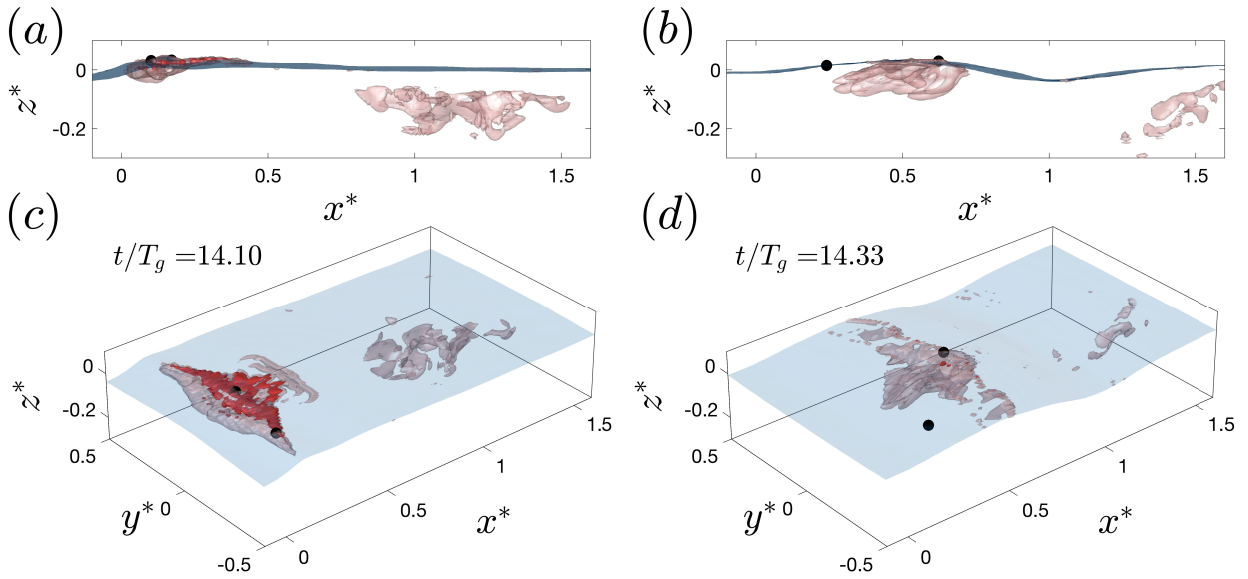


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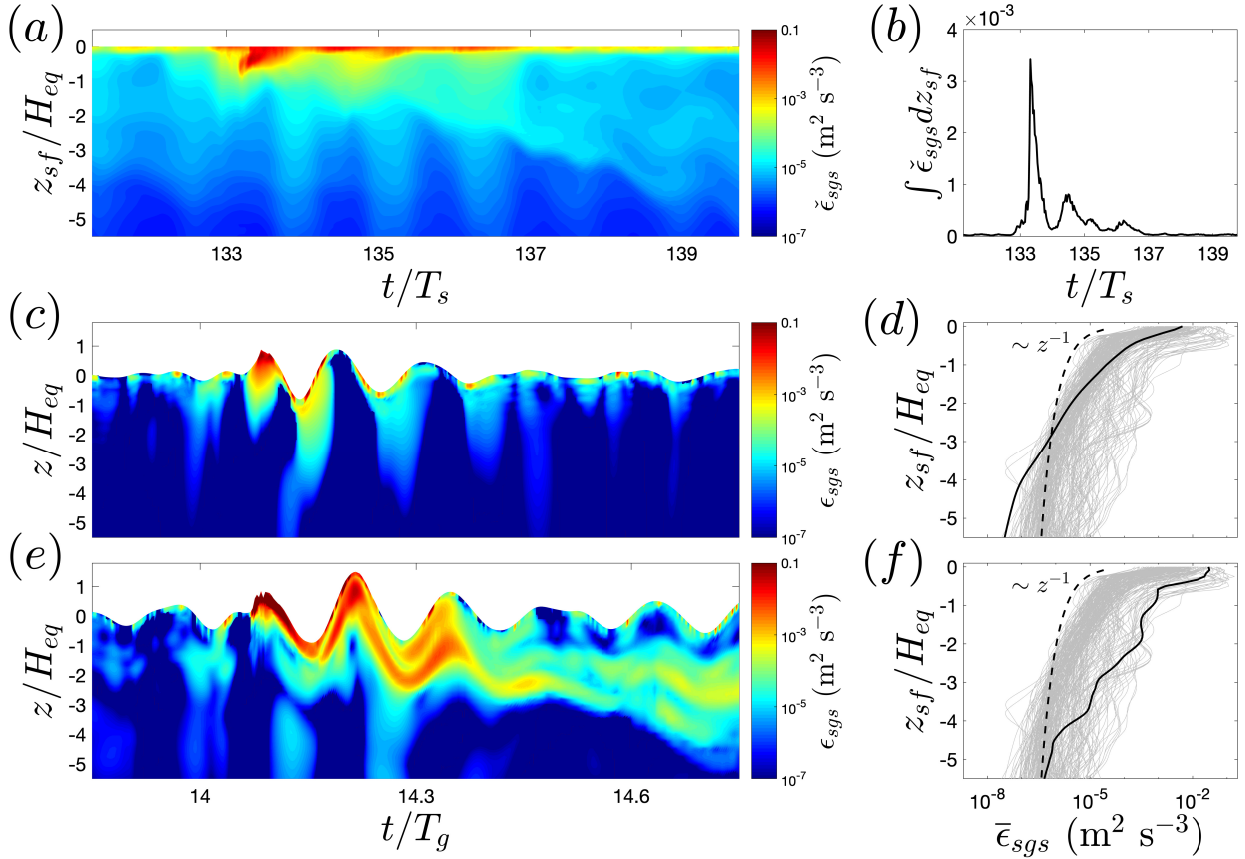


FIG. 3. Various measures of the turbulence dissipation rate ϵ_{sgs} for the breaking event E3 of the case T1. (a) Phase-resolved horizontal-averaged turbulence dissipation rate $\check{\epsilon}_{sgs}$ in the surface following reference frame, (b) temporal variation of the vertically integrated $\check{\epsilon}_{sgs}$, (c, e) phase-resolved ϵ_{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 ($\sim 1/z$).

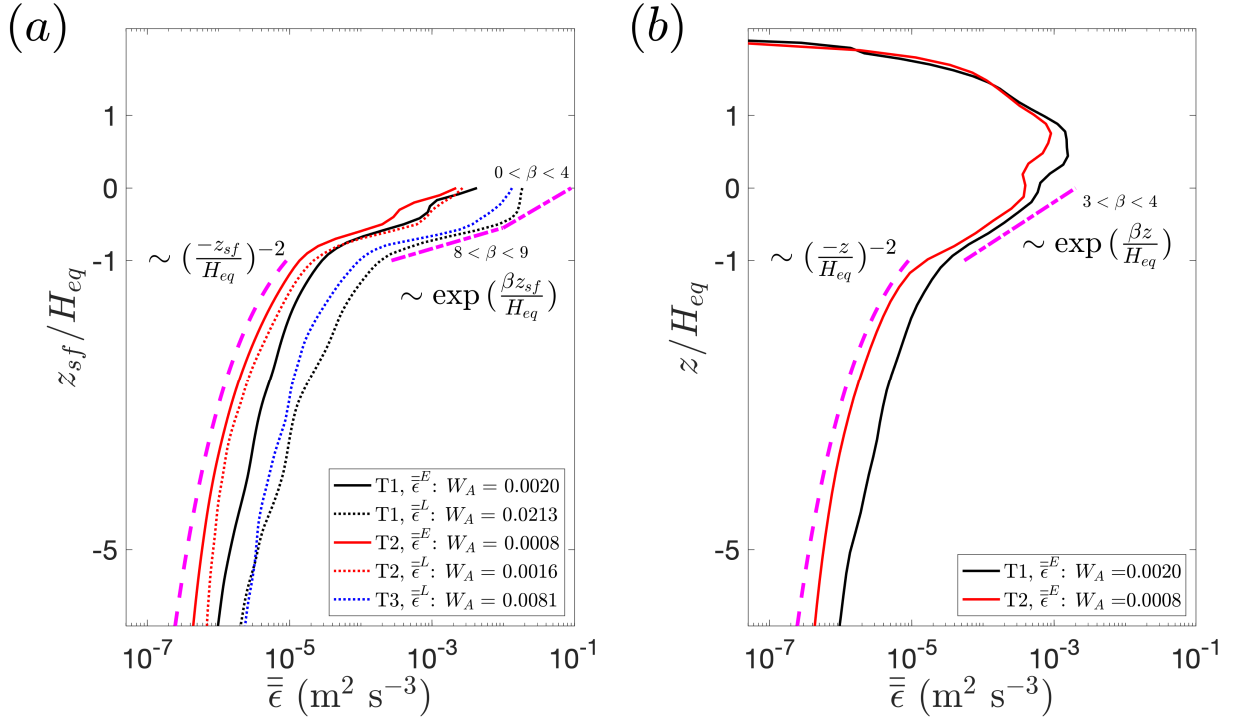


FIG. 4. Model results of the vertical profile of the ensemble-time-averaged turbulence dissipation rates $\bar{\epsilon}$ in (a) the surface-following reference frame, and (b) the fixed reference frame. Here H_{eq} the equilibrium wave height, W_A the active whitecap coverage, $\bar{\epsilon}^L$, and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

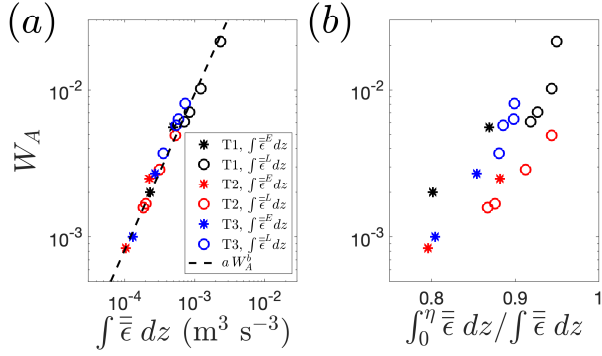


FIG. 5. Model results of the variation of (a) the vertically integrated long-time-averaged dissipation rates, and (b) the fraction of total dissipation above still water level $z = 0$ with the active whitecap coverage W_A (Eq. 9).

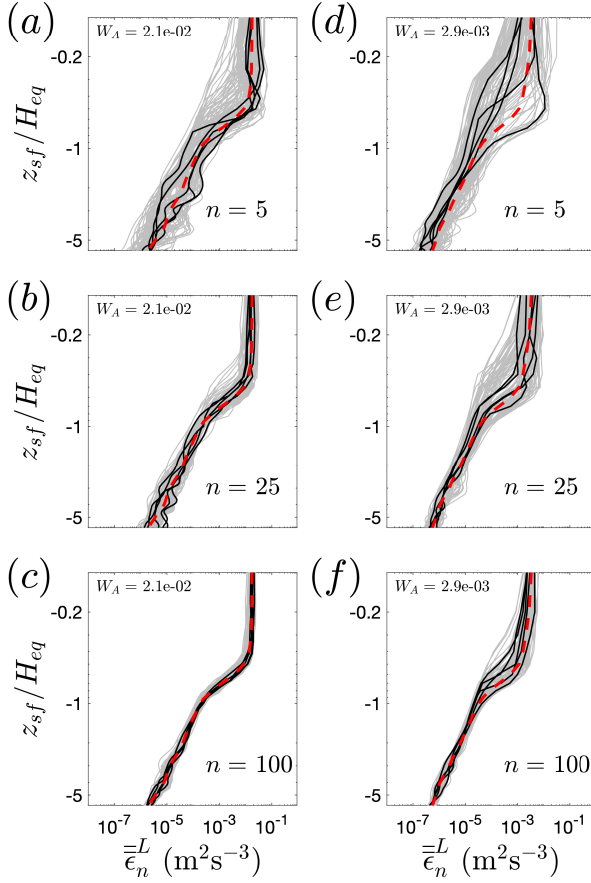


FIG. 6. Variation of the Lagrangian statistics of turbulence dissipation rates obtained from n randomly selected virtual drifters $\overline{\epsilon}_n^L$ (solid lines) with the signal length $\mathcal{L} \approx 9.5nT_s$ for cases with (a – c) $W_A = 2.1 \times 10^{-2}$, and (d – f) $W_A = 2.9 \times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

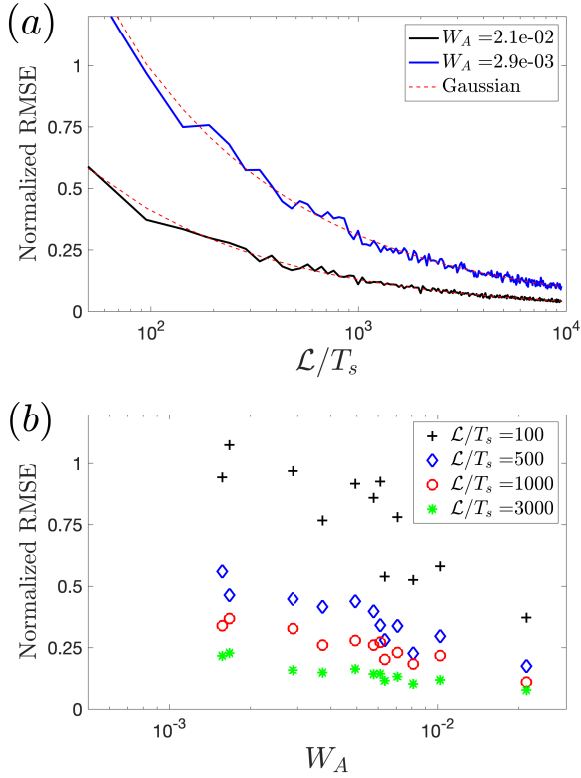


FIG. 7. Variation of the normalized standard error of total turbulence dissipation rate estimates (a) with the signal length $\mathcal{L}/T_s \approx 9.5n$, and (b) with active whitecap coverage W_A . The dashed lines in (a) show the normalized standard error estimation using the Gaussian distribution.

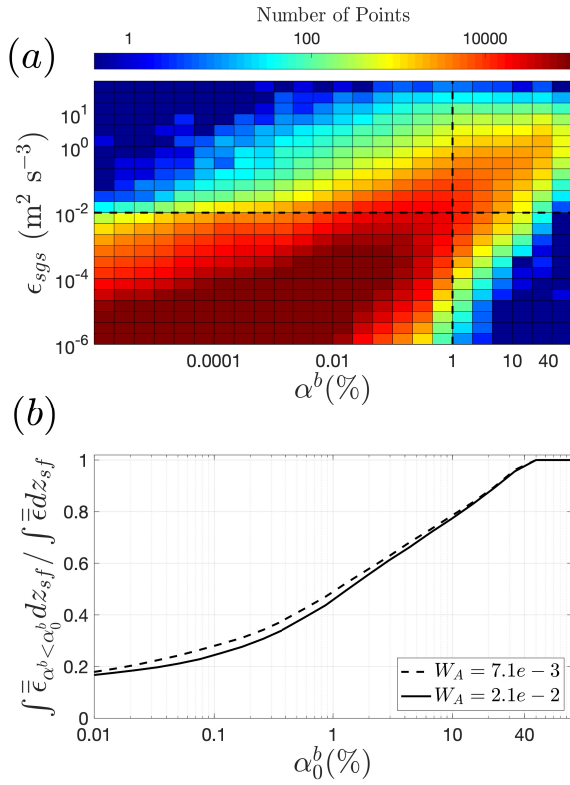


FIG. 8. (a) Example of a 2-D histogram of the model results of the local turbulence dissipation rates ϵ_{sgs} and bubble void fraction, and (b) two examples of the variation of the fraction of the total dissipation within the regions with $\alpha^b < \alpha_0^b$ against α_0^b . In (a), the vertical and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

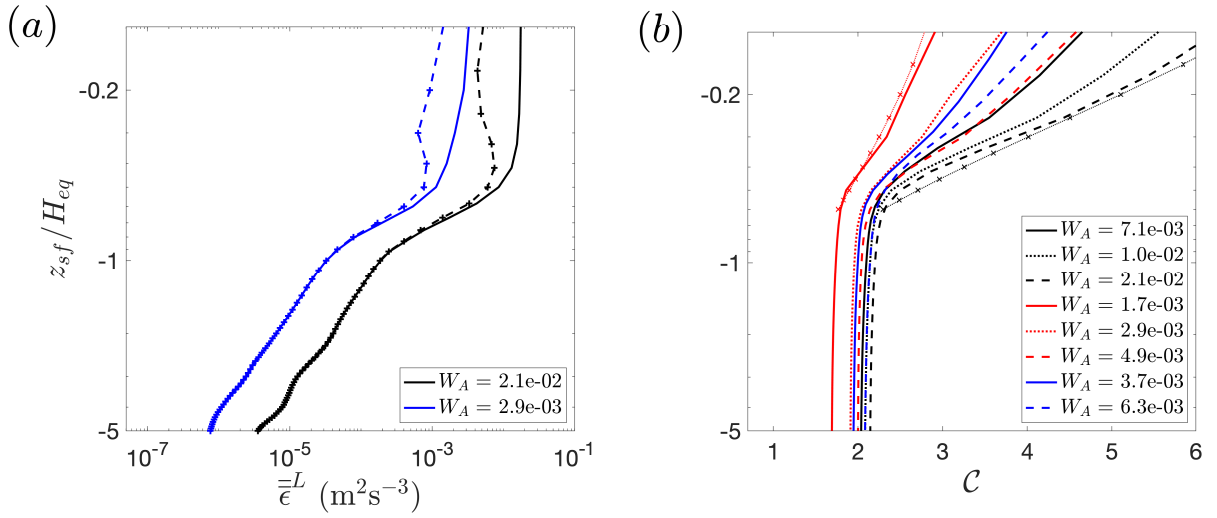


FIG. 9. Incomplete sampling of turbulence dissipation rates. (a) Effect of the bubble occlusion on the Lagrangian averages by considering (solid lines) all model results $\bar{\epsilon}^L$ (Eq. 10), and (dashed lines) only data points in which $\alpha^b < 1\%$. (b) Variation of the correction factor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

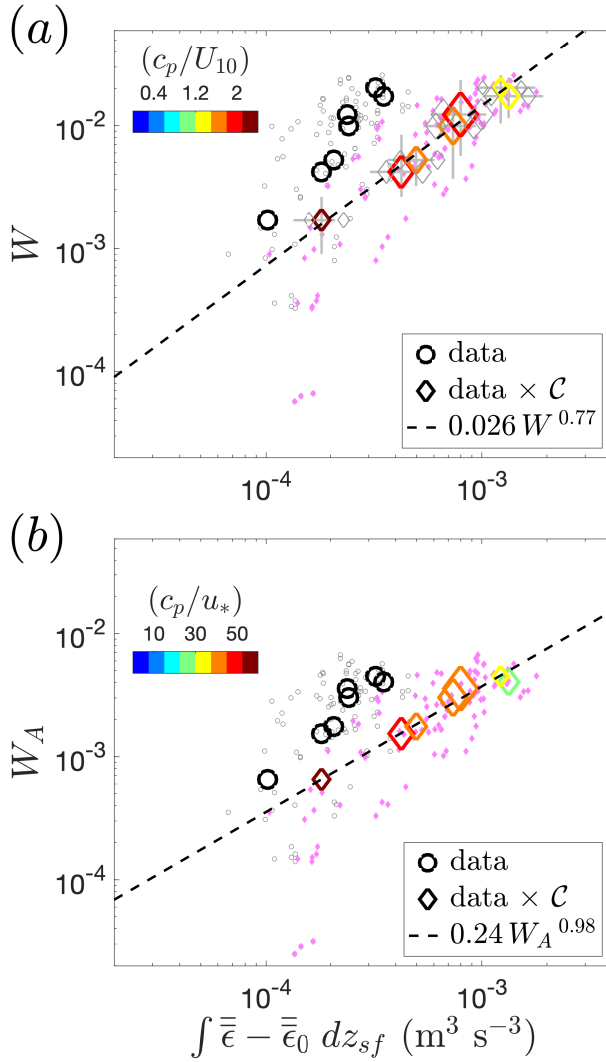


FIG. 10. Variation of the total wave-breaking-induced TKE dissipation rates with (a) total whitecap coverage W , and (b) active whitecap coverage $W_A = \lambda W$ (Eq. 18). Circles show the raw data provided by Schwendeman and Thomson (2015), and diamonds show the corrected results by applying the correction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

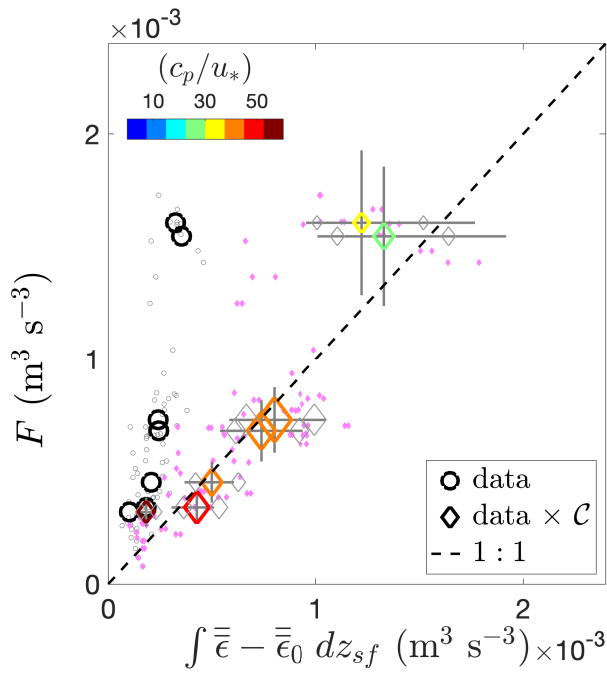


FIG. 11. Variation of the total wave-breaking-induced TKE dissipation rates with the rate of wind energy input F . Vertical line segments represent the sensitivity of F values with respect to $2 < c_e < 3$. Definitions of the rest of symbols and lines are the same as in Figure 10.