The Intensifying Role of High Wind Speeds on Air-Sea Carbon Dioxide Exchange

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Key Points:
• Global CO2 fluxes at high wind speeds were estimated using a new wave-wind dependent formulation
• High wind speed events, which represent some 3% of wind conditions, contribute 18% of the global air-sea CO2 flux
• Approximately 50% of the global flux at high wind speeds is attributed to bubbles associated with wave breaking

Abstract
While it has been known that wave breaking and bubble generation at high wind speeds enhance air-sea carbon dioxide (CO2) exchange rates (F), quantification of their contribution at the global scale remains a formidable challenge. There is urgency to make progress on this issue as a significant uptick in both magnitude and frequency of high wind events (HW) has been documented over the last 3 decades. Using a wind-wave dependent expression for gas transfer velocity (k) that explicitly considers bubbles and a widely used wind-only parameterization, the spatial pattern of k at high winds can be explained by sea surface temperature distribution. The HW, which represent some 3% of wind conditions, contribute disproportionally to the global F (18%) with an increasing trend. Approximately 50% of the global F at high winds is attributed to bubble contribution. The findings are of significance to quantifying CO2 transfer to the ocean interior.

Plain Language Summary
Studies on air-sea carbon dioxide (CO2) exchange seek to determine how rapidly CO2 molecules traverse the air-water interface. This exchange impacts a plethora of processes related to ocean biogeochemistry, oceanic carbon cycling, and CO2 buildup in the atmosphere. At high wind speeds, both conventional aerodynamic transfer processes and bubbles generated by wave breaking are expected to enhance air-sea CO2 gas exchange. Yet, the impact of bubbles in isolation on global and regional CO2 exchange remains a subject of inquiry and speculation. There is some urgency to make progress on this topic because the magnitude and frequency of high wind events (HW) have been steadily increasing over the last 3 decades. The work here demonstrates that bubbles contribute as much as 50% of CO2 gas exchange under high wind speeds conditions, which rarely occur over the ocean (less than 3% of the time). Yet, HW contribute disproportionally to the global air-sea CO2 gas exchange (about 18%).

1. Introduction
Air-sea flux of carbon dioxide (CO2; [F]) is necessary to any global carbon cycle assessment and its concomitant role in regulating the climate system. Generally, F is estimated from a water-side bulk transfer relation given as

\[ F = \left( k \right) \left( K_0 \right) \left( \Delta p_{CO_2} \right) = \left( k \right) \left( K_0 \right) \left( p_{CO_2w} - p_{CO_2a} \right) \]

(1)

where positive fluxes denote oceanic outgassing and negative fluxes denote uptake, \( K_0 \) is the solubility of CO2 (mol L\(^{-1}\) atm\(^{-1}\)) assumed to vary with sea surface temperature (SST) and salinity, p\(_{CO_2w}\) and p\(_{CO_2a}\) are the partial pressure of CO2 in water and air (μatm), respectively, and k is the gas transfer velocity (cm hr\(^{-1}\)) and is the subject of this letter. The parameterization of k in the open ocean has long been a challenge contributing substantial uncertainties to global estimates of F. A number of small-scale physical processes such as waves, turbulence, tidal currents, rain, subskin-to-skin temperature gradients, and surface films are known to impact k and the air-sea CO2 exchange (Broecker et al., 1978; Frew et al., 2004; Ho et al., 1997; Watson et al., 2020; Zappa et al., 2007). However, accommodating all these physical processes and aggregating their effects at scales relevant to climate models remains a challenge. For this reason, common formulations empirically relate k to the dominant mechanism, which is the mean wind speed (U) at some reference height, usually set at 10 m. By mean U, we are referring to time-scales that are sufficiently long to average
over many water-side eddy turnover times making contact with air-water interface but short enough to resolve some aspects of the diurnal wind patterns (usually few hours). These formulations take the form of \( k \propto U^n \) with \( n = 1 \) (Liss & Merlivat, 1986), \( n = 2 \) (Sweeney et al., 2007; Wanninkhof, 1992, 2014), and \( n = 3 \) (McGillis et al., 2001; Pytharoulis et al., 2010; Wanninkhof & McGillis, 1999) or similar polynomial expressions (Nightingale et al., 2000; A. Weiss et al., 2007) all being proposed and receiving partial experimental support. However, recent measurements suggest that such wind-only relations cannot be extrapolated to high \( U \) where bubble formation is likely to occur (Bell et al., 2017; Blomquist et al., 2017). At intermediate to high \( U \), bubbles generated at the ocean surface, primarily associated with air entrainment by breaking waves, disproportionately enhance \( k \) and subsequently \( F \) (Johnson & Liss, 2014; Woolf, 1993, 1997). Significant contribution of bubbles to global \( k \) and \( F \) for CO₂ are now estimated at about 30% and 40%, respectively (Reichl & Deike, 2020; Woolf, 1997). Regionally, hurricanes predominantly occurring in tropical and subtropical oceans have been observed to significantly facilitate CO₂ exchange due to their enhanced \( k \) associated with high \( U \) (Bates, 2007; Bates et al., 1998; Huang & Imberger, 2010).

Global \( U \) above oceans have experienced an uptick in magnitude over recent decades (Young & Ribal, 2019; Young et al., 2011; Zheng et al., 2016), but tracking how this increase in \( U \) influences \( F \) via enhanced bubble formation at such large scales remains difficult (Le Quéré et al., 2010; Wanninkhof & Triñanes, 2017) thereby motivating the present work. A number of theoretical developments and data products are now offering new tools to address this question. More than 30 years of high-resolution satellite \( U \) data is enabling the detection of long-term trends associated with high winds to be unpacked and utilized in the estimation of \( k \). High wind events (HW) predominantly occur in the midlatitudes’ winters over the North Atlantic, the North Pacific, and the Subantarctic oceans. HW are generally caused by synoptic-scale transient eddies along the midlatitude strong storm track or air-sea interaction over sharp SST fronts (Chelton et al., 2004; Minobe et al., 2008; Sampe & Xie, 2007; Small et al., 2008; Spall, 2007; Xie, 2004). Over tropical oceans, HW occur infrequently (<1%) and are predominantly associated with sporadic tropical cyclones (Sampe & Xie, 2007). While the impact of such extreme wind events (\( U > 33 \) m s\(^{-1}\)) on \( k \) has been explored in earlier studies (see references above and Lévy et al., 2012; Liang et al., 2020), the broader implications of HW on \( k \) is unclear.

On the theoretical front, a new semi-empirical equation for \( k \) that accounts for significant wave height (\( H_s \)) and separates \( k \) into contributions of turbulence and bubbles has recently been proposed (Deike & Melville, 2018). This formulation, labeled here as \( k_{D18} \), was shown to be in agreement with a number of gas exchange experiments at low and high \( U \) (Deike & Melville, 2018) and was recently applied to explore bubble-mediated air-sea CO₂ flux (Reichl & Deike, 2020).

Building on prior work (Reichl & Deike, 2020; Wanninkhof & Triñanes, 2017), the focus here is on global variability in temporal trends in \( k \) and \( F \) under high wind speed conditions (to be defined later). With improved knowledge of how winds affect long-term trends in global air-sea CO₂ exchange (Wanninkhof & Triñanes, 2017), the overarching question to be explored is how \( U \) impact \( k \) and \( F \) globally when explicitly accounting for the bubble effect. The new \( k_{D18} \) is employed and contrasted with the widely used wind-only parameterization (Wanninkhof, 2014), labeled as \( k_{W14} \). The manuscript is organized as follows: the datasets and data processing methods are introduced in Section 2. In Section 3, the climatological distribution and long-term trends in \( k \) along with concomitant \( F \) estimates are analyzed focusing on HW. Limitations and concluding remarks are presented in Section 4 and Section 5, respectively.

2. Materials and Methods

2.1. Parameterization of the Air-Sea CO₂ Flux

The coefficient of the wind-only formulation and gas transfer velocities depend on the wind product being used. The \( k_{W14} \), as an update of a prior expression already in use in current climate models (Wanninkhof, 1992), is derived from the high-resolution Cross-Calibrated Multi-Platform (CCMP) wind product. This high-resolution wind product operating under short-term conditions is calibrated to match global ocean bomb-\(^{14}\)C inventories resulting in an “up-grade” to \( k_{W14} \) (Wanninkhof, 2014). The expression for \( k_{W14} \) is given as
where the coefficient “α” is 0.251 in the unit of (cm hr⁻¹)(m s⁻¹)⁻². The Sc is the molecular Schmidt number (>>1) given by the ratio of the kinematic viscosity in seawater and the molecular diffusion coefficient of CO₂, \( U^2 \) is the squared wind speed (or twice the mean flow kinetic energy) measured at 10-m height.

The second expression employed here is a recently developed wind-wave dependent formulation (Deike & Melville, 2018). This formulation separates the bubble-mediated term \( k_b \) from the none-breaking term \( k_{nb} \) and results in a \( k_{D18} \) given as

\[
k_{D18} = k_{nb} + k_b = u_c \left[ A_{NB} \left( \frac{Sc}{660} \right)^{-1/2} + A_b \left( \frac{W_0}{W_0} \right)^{2/3} \right]
\]

where \( A_{NB} = 1.55 \times 10^{-4} \), \( A_b = 1 \pm 0.2 \times 10^{-5} \) m⁻² s², \( W_0 \) is the dimensionless Otswald solubility coefficient, \( g = 9.8 \) m s⁻² is the gravitational acceleration, and \( H_s \) is the significant wave height. \( C_{wh} = \sqrt{gH_s} \) is the ballistic speed, an important parameter for wave breaking, that can be related to the phase speed at the peak of the wave spectrum for fetch-limited conditions (Deike & Melville, 2018). The air-side friction velocity is \( u_* = \left( \frac{\tau}{\rho_{air}} \right)^{1/2} \), where \( \rho_{air} \) is the mean air density and \( \tau \) is a turbulent shear stress assumed to be a function of \( U \) via the quadratic drag-law \( \tau = \rho_{air} C_D U^2 \) with a wind-dependent drag coefficient \( C_D \) at \( z = 10 \) m derived elsewhere (Large, 2006). With this representation, \( u_* \) (m s⁻¹) can be related to an externally supplied \( U \) via

\[
\frac{u_*}{U} = \sqrt{C_D} = \left[ \frac{A_{u1}}{U} + A_{u2} + A_{u3}U \times 10^{-3} \right]^{1/2}
\]

where \( A_{u1} = 2.7 \) (m s⁻¹), \( A_{u2} = 0.142 \) (dimensionless), and \( A_{u3} = 0.076 \) (s m⁻¹). This \( C_D \) representation allows \( k_{D18} \) to be expressed as a function of \( U, c_{wh} \), and \( SST \).

It is to be noted that the \( k_{W14} \) expression is based on matching long-term integrated fluxes (dual tracer technique) but using higher resolution \( U \) (6-hourly), while the data used to develop the \( k_{D18} \) formulation are fitted based on eddy covariance measurements of the flux in the atmosphere (with averaging intervals less than 15 min and sampling frequencies exceeding 10 Hz). Thus, the \( k_{D18} \) formulation has been calibrated and tested at a much higher temporal resolution. Reconciliation of estimates of \( k \) based on eddy covariance, dual tracer and other approaches is still debated in the community (Edson et al., 2011; Wanninkhof et al., 2009).

While the focus here is on HW, trends in \( k \) and \( F \) for all winds were also estimated. When including all wind conditions, the findings derived here parallel conclusions of Wanninkhof and Triñanes (2017), though different \( k \) formulations were used (see Figures S1 and S2).

2.2. Data Products

The \( U, H_s, \) SST, salinity, and pCO₂ data from 1990 to 2018 are used to compute global spatial patterns and trends in \( k \) and \( F \) at high \( U \). All these data are linearly interpolated onto a spatial resolution of 0.5° and averaged to a daily temporal resolution. The 6-hourly CCMP-v2 wind speed data at the 0.25° grid are obtained from the Remote Sensing Systems described elsewhere (Atlas et al., 2011). This wind product is produced using satellite, moored buoy, and model wind data fusion, and it agrees with mooring records and other wind products described elsewhere (Kent et al., 2013; Wanninkhof & Triñanes, 2017). The CCMP product is available since 1987; however, there are appreciable gaps in the records in 1988 and 1989. Therefore, the period from 1990 through 2018 is used here. Both 6-hourly SST and \( H_s \) data at 0.5°C resolution are derived from the European Centre for Medium-Range Weather Forecasts (ECMWF) fifth generation ERA5 reanalysis products described elsewhere (Copernicus Climate Change Service (C3S), 2017). ERA5 is the latest released climate dataset produced using ECMWF’s Integrated Forecast System, which is coupled to an ocean wave
model. A large fraction of \( H_s \) data are missing in polar regions due to ice coverage, especially in winter (Figures S3 and S4). We thus set these missing values in \( H_s \) to zero (meaning no bubble contribution to \( k \)).

The monthly climatological sea surface salinity data at 1° grid are obtained from World Ocean Atlas 2009 (WOA09) described elsewhere (Antonov et al., 2010).

The \( Sc \) for CO\(_2\) is a function of SST and is determined using standard formulation (Wanninkhof & Triñanes, 2017). The solubility is expressed as a function of water temperature (\( T \), in Kelvin) and salinity (\( S \), in ‰) using the empirical relation (R. F. Weiss, 1974):

\[
\ln(K_0) = A_1 + A_2 \frac{100}{T} + A_3 \ln\left(\frac{T}{100}\right) + S \left[ B_1 + B_2 \frac{T}{100} + B_3 \left(\frac{T}{100}\right)^2\right] 
\]

where \( K_0 \) is the solubility (mol L\(^{-1}\) atm\(^{-1}\)). The numerical values for these coefficients are \( A_1 = -58.0931 \), \( A_2 = 90.5069 \), \( A_3 = 22.2940 \), \( B_1 = -0.027766 \), \( B_2 = -0.025888 \), and \( B_3 = 0.0050578 \). The dimensionless Otswald solubility coefficient \( W_0 \) is expressed as:

\[
W_0 = K_0 R T 
\]

where the ideal gas constant \( R \) is taken as 0.08205 L atm mol\(^{-1}\) k\(^{-1}\).

The impact of \( U \) on trends in \( F \) was evaluated using averaged \( \Delta pCO_2 \) estimates. The monthly climatological \( pCO_2 \) centered on year 2005 is from Takahashi et al. (2014). The \( pCO_2 \) for the 2005 reference year is calculated following the approach in Takahashi et al. (2009) and is given as:

\[
pCO_2a = xCO_2 \left( P_{baro} - P_{sw} \right) 
\]

where \( xCO_2 \) is the CO\(_2\) mole fraction or mixing ratio, \( P_{baro} \) is the barometric pressure at the sea surface, \( P_{sw} \) is the water vapor pressure at seawater temperature. The \( xCO_2 \) data are retrieved from the NOAA greenhouse Gas Marine Boundary Layer Reference data product (Conway et al., 1994). The \( P_{baro} \) is from NCEP/NCAR Reanalysis 1 (Kalnay et al., 1996) and the \( P_{sw} \) is from NCAR/UCAR Atlas of Surface Marine Data.

### 2.3. External Factors Influencing \( k \) and \( F \)

A sensitivity analysis was conducted to evaluate the effects of changes in \( \Delta pCO_2 \), \( U \), SST, and \( H_s \) on \( k \) and \( F \) at high \( U \) by varying a given property while setting the other properties to their climatological means (Table S1). The starting values of \( \Delta pCO_2 \) and salinity were set to −5.6 μatm and 34.74 psu, respectively according to climatological global mean salinity and the \( \Delta pCO_2 \) in the reference year 2005. The starting values of \( U \), SST, and \( H_s \) were set to 16.5 m s\(^{-1}\), 8.5°C, and 4.8 m, respectively, according to their climatological mean at high \( U \).

Three scenarios of imposed changes are applied to estimate the sensitivity of \( k \) and \( F \) to factors listed in Table S1. Variations in SST, \( U \), and \( H_s \) for the first two scenarios are based on temporal variability of globally averaged values at high \( U \) during the period 1990–2018. The third scenario is based on (climatological) spatial gradients between tropics and subpolar regions of these factors at high \( U \), as discussed in Section 3.2. Imposed changes in \( \Delta pCO_2 \) are in proportion to trends in differences in oceanic \( pCO_2 \) at Station ALOHA and atmospheric \( pCO_2 \) at Mauna Loa.

The sensitivity of \( k \) and \( F \) to each external factor was assessed from the ratio of the percentage change in \( k \) or \( F \) (labeled \( Y \)) to percentage change in each factor (labeled \( X \)) using

\[
\text{Sensitivity} = \frac{\Delta Y / Y}{\Delta X / X} 
\]

### 3. Results and Discussion

We now define a HW as conditions for which \( U > 15 \) m s\(^{-1}\). This threshold is based on the rapid increase in bubble formation with increasing \( U \). With the significant uptick in both frequency and magnitudes of HW over the last 3 decades (Figures 1a and 1b), we focus our efforts on the drivers of climatological and long-term trends in \( k \) and \( F \) at high \( U \).
3.1. Spatial Pattern of Gas Transfer Velocity at High Winds

The global long-term averages of $k$ at high $U$ ($U_{\text{HW}}$) are approximately 49 cm hr$^{-1}$ for $k_{\text{D18}}$ and 51 cm hr$^{-1}$ for $k_{\text{W14}}$, around 3–4 times higher than the overall mean $k$ at all wind speeds. While HW are rare in tropical and subtropical regions (Figure 1c), they are associated with higher $k$ than in subpolar and polar regions (Figures 1d and 1e). Interestingly, these extremes in $k_{\text{HW}}$ are not associated with extreme winds (Figure 1g) or high waves (Figure 1h), but rather predominantly driven by warmer temperature (Figure 1i). The large geographical variability in SST (−2 to 30°C) and the narrow range of spatial variations in $U$ (about 99% of the range of 15–18 m s$^{-1}$) and $H_s$ (5–7 m) at high $U$ (Figure S5) explains the dominant role of SST in setting the spatial pattern of $k_{\text{HW}}$ for both $k$ parameterizations. As shown by the sensitivity analysis (Table S1), though the $k_{\text{HW}}$ is much more sensitive to changes in SST than $k_{\text{D18}}$ and $k_{\text{W14}}$, a 150% increase in SST (with a reference starting values of 8.5°C) from subpolar regions to tropical regions leads to a 20 cm hr$^{-1}$ increase in $k_{\text{HW}}$. In contrast, changes in $U$ and $H_s$ (with references starting values of 16.5 m s$^{-1}$ and 4.8 m, respectively) between subpolar regions and tropical regions could only induce up to 8 cm hr$^{-1}$ changes in $k_{\text{HW}}$.

As expected, $k_{\text{D18}}$ increases with $H_s$ (Figures 1j–1l). The $k_{\text{D18}} > k_{\text{W14}}$ in regions experiencing frequent HW such as subpolar to subpolar regions (Figure 1f), likely results from $H_s$’s contribution to the bubble parameterization in $k_{\text{D18}}$ at high $U$ (Figures 1h and 1i). However, $k_{\text{D18}}$ is generally smaller than $k_{\text{W14}}$ at high $U$ in polar regions (Figure 1f) where the average $H_s$ is around 2 m (Figures 1h and 1j). Wind-only parameterization may therefore underestimate the bubble-enhanced exchange under high $U$ and $H_s$ conditions and overestimates bubble contributions in polar regions where $H_s$ are lower than expected, potentially due to reduced wind fetch and wave-ice interaction near ice-covered regions, especially in winter (Ardhuin et al., 2020; Herman et al., 2019; Smith & Thomson, 2016; Squire, 2020; Voermans et al., 2019). Uncertainties in bubble-mediated gas exchange velocities in polar and subpolar regions can have a significant impact on the net air-sea CO$_2$ exchange because intense ventilation and deep water formation occur in these regions.

Moreover, a comparison of $k_{\text{SO}}$ calculated using $k_{\text{D18}}$ and $k_{\text{W14}}$ to field measurements of $k_{\text{SO}}$ indeed suggest a better representation of gas exchange when explicitly accounting for the bubbles’ contribution to gas exchange at high $U$ (Figure S6).
3.2. Trends in Gas Transfer Velocity at High Winds

The global average $k_{\text{HW}}$ is increasing at a rate of 0.7 cm hr$^{-1}$ dec$^{-1}$ over the last 3 decades (Figure S7). Spatially, the distribution of long-term trends in $k_{\text{HW}}$ (Figures 2a and 2b) closely follows the trends in $U_{\text{HW}}$ (Figure 2d) with both generally increasing over time (and with no detectable saturation effect), consistent with trends in $k_{\text{HW}}$ being dominantly driven by changes in $U_{\text{HW}}$ over the global ocean. The global average SST at high $U$ has been increasing faster than SST at all wind conditions (Figure S8). The trends in SSTs at high $U$ (Figure 2f), such as in the North Atlantic, North Pacific, and regions along 45°S only marginally alter the trends in $k_{\text{HW}}$. As revealed by a sensitivity analysis conducted in the Northwest Atlantic, a 1.6% increase in $U_{\text{HW}}$ (with a reference starting values of regional mean high $U$ at 17.2 m s$^{-1}$) leads to an approximately 1.3 cm hr$^{-1}$ increase in $k_{\text{HW}}$. In contrast, a 16% increase in SST (with references starting values of regional mean SST at high $U$ at 1.9°C) could only lead to a 0.42 cm hr$^{-1}$ increase in $k_{\text{HW}}$. While the global distribution of temporal trends in $k_{\text{D18}}$ and $k_{\text{W14}}$ agree, they differ in magnitude (Figure 2c), largely due to the impact of trend in $H_s$ over the study period (Figure 2e, as expected from Equation 3).

To further evaluate the processes responsible for the long-term trends in $k_{\text{HW}}$, the bubble- and turbulence-mediated $k_{\text{HW}}$ were analyzed independently. Spatially, the temporal trends in bubble-mediated air-sea gas flux kinetics (Figure 2g) appears to be controlled by $U$ (or $u^*$) and less influenced by $H_s$ (Figures 2d and 2e). A plausible explanation for these results is the sub-unity exponent for $H_s$ and the above-unity exponent for $u^*$, $k_b \propto u_s^{5/3} H_s^{2/3}$. In contrast, the spatial pattern of trends in non-bubble (turbulence) mediated gas flux kinetics at high $U$ are predominantly determined by $U$ (Figures 2h and 2d) because $k_{nb} \propto u_s$. The inconsistent trends in $U$ and $H_s$ observed here (Figures 2d and 2e) could be because the latter term is dominated by remotely generated swell rather than by local wind shear over large parts of the ocean (Semedo et al., 2011; Young, 1999; Young & Ribal, 2019). That $U$ and $H_s$ are not strongly coupled also allows exploring the sensitivity of $k_{\text{HW}}$ to the aforementioned two variables to be treated separately.

For convenience, the ratio of bubble to turbulent contribution to $k$ ($= R_b$) and thus the ratio of bubble component to overall flux as measured by $R_{vb}$ can be expressed as
where an \( R_b \gg 1 \) results in an \( R_{fb} = 1 \) and an \( R_b = 1 \) results in \( R_{fb} = 0.5 \). As \( U \) increases, \( u_\ast \) and \( H_\ast \) increase, and \( c_w b \) generally also increases through \( H_\ast \) (because \( c_w b = \sqrt{8 H_\ast} \)) thereby resulting in increased \( R_b \) with increased \( U \). The estimate indicates that bubble contribution to \( k_{D18} \) is around 45% at high \( U \) conditions, which is expectedly greater than 30% estimated at all winds (Reichl & Deike, 2020; Woolf, 1997).

### 3.3. Trends in Air-Sea CO₂ Flux at High Winds

Interestingly, the most significant increases in \( k_{W14} \) is occurring in regions with high CO₂ exchange with the ocean interior through deep water formation and ventilation (Figure S9, e.g., the Southern Ocean and North Atlantic). As a result of these trends, we observe an enhancement in global mean CO₂ oceanic uptake at high \( U \) at rates of \(-0.040\) mol m\(^{-2}\) y\(^{-1}\) dec\(^{-1}\) and \(-0.045\) mol m\(^{-2}\) y\(^{-1}\) dec\(^{-1}\) when determined using \( k_{D18} \) and \( k_{W14} \), respectively (Figure 3a; enhancement associated only with wind trends, keeping \( \Delta p_{CO₂} \) constant).

The ratio of net \( F \) at high \( U \) to the global net \( F \) measured by \( R_F \) is estimated and expressed as

\[
R_F = \frac{F_{HW}}{F} = \frac{F_{HW}}{\frac{F_{HW}}{N_{HW}} N} = \frac{F_{HW}}{F} = \frac{F_{HW}}{F} \cdot \frac{N_{HW}}{N}
\]

where \( F_{HW} \) and \( F \) are in units of Pg C y\(^{-1}\), \( F_{HW} \) and \( F \) are in units of mol m\(^{-2}\) y\(^{-1}\), \( N \) is the total number of oceanic grids, \( N_{HW} \) is the number of high wind grid days (defined as a grid when daily \( U \) over 15 m s\(^{-1}\)). With a slightly negative trend in global mean \( F \) (Figure S2f), the contribution of HW to air-sea CO₂ exchange (\( R_F \)) increases as a result of increased magnitude (\( F_{HW} \)) and frequency (\( N_{HW} \)) at high \( U \) (with comparable trends for \( k_{D18} \) and \( k_{W14} \) expressions; Figure 3b and Equation 9). As shown by Equation 8, with increasing contribution from bubbles (i.e., \( R_b \)), the bubble-associated flux contribution to total CO₂ flux (\( R_{fb} \)) increases as a saturation function at high \( U \) (Figure 3d). Globally, we estimate \( R_{fb} \) at about 36%, which is commensurate with earlier estimates of some 40% (Reichl & Deike, 2020). However, at high \( U \), \( R_{fb} \) reaches as high as 50% (i.e., \( R_b = 1 \)). These results suggest that both turbulent and bubble contributions have comparable effects on \( F \) at high \( U \) and both components should be considered in future climate studies. Overall, trends in HW are significant because, despite representing a minor fraction of total winds (3%), they disproportionately contribute to air-sea CO₂ exchange (18%).

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**Figure 3.** Temporal evolution of (a) spatially averaged flux at high \( U \) (\( F_{HW} \)), (b) the ratio (\( R_F \)) of net \( F \) at high \( U \) to global net \( F \), (c) flux from the turbulent and bubble contributions at high \( U \) (Partial \( F_{HW} \)), (d) the ratio of bubble (\( R_{fb} \)) and turbulent (\( R_{fb} \)) component under high winds to global net \( F \) quantified using \( k_{D18} \). For (a) and (b), magenta represents \( F \) from \( k_{D18} (F_{D18}) \) and blue represents \( F \) from \( k_{W14} (F_{W14}) \). For (c) and (d), orange represents the turbulent component and purple represents the bubble component. The dashed line indicates the long-term trend. Negative signals mean CO₂ uptake by the ocean.
4. Caveats and Limitations

Our study carries limitations that warrant further elaboration. First, the $k_{H2S}$ and $k_{W14}$ expressions do not explicitly consider bubble-induced supersaturation. Using Large Eddy Simulations coupled with bubble dynamics and a transport model, Liang et al. (2020) showed that bubble supersaturation under hurricane conditions can substantially suppress CO$_2$ outgassing. It is difficult to extrapolate their results here as hurricane winds commence at 33 m s$^{-1}$, while 99% of the HW considered here are in the range of 15–18 m s$^{-1}$ (Figure S5a). Furthermore, the gas transfer velocities estimated using $k_{H2S}$ are in good agreement with eddy covariance field measurements for $U$ in the range of 0–25 m s$^{-1}$. Hence, the $k_{H2S}$ may indirectly include bubble-induced supersaturation effect though it remains unclear whether the equation derived in CO$_2$ sink regions can be applied in CO$_2$ source regions (Liang et al., 2020). Second, time-averaged $U$ products can introduce some biases in all $k$ due to the non-linear relation between $U$ and $k$ (Wanninkhof, 2002), a topic that is better kept for a future inquiry. Third, the $C_D$ expression linking friction velocity to $U$ required by $k_{H2S}$ does not consider thermal stratification in the atmosphere. However, the focus here is on high $U$ where mechanical production of turbulent kinetic energy far exceeds buoyancy forces. Finally, the results may be dependent on the particular $\Delta p_{CO2}$ product employed to estimate $F$. To ascertain that the overall results are independent of the $\Delta p_{CO2}$ product, we re-ran the analyses with the $\Delta p_{CO2}$ data of Landschützer et al. (2014). While the gas transfer velocities are independent of the $\Delta p_{CO2}$ product, the contribution of HW to CO$_2$ fluxes differs in the precise magnitude but not the trends. For the two $\Delta p_{CO2}$ products employed, HW events become increasingly significant over time in air-sea CO$_2$ fluxes (Figure S10).

5. Conclusions

Wind speed observations covering the 1990–2018 period and recent gas exchange parameterizations were used to examine the importance of synoptic high-wind events on CO$_2$ fluxes. These HW are linked to situations when bubbles are expected to play a significant role in gas transfer velocity and air-sea CO$_2$ gas exchange. Given the large spatial variations in SST, it is the primary variable explaining the time-averaged spatial variability in $k$ when conditioned on high wind speeds only (i.e., $k_{HW}$ for both parameterizations). This finding is in contrast to the distribution of $k$ for all winds condition, which is primarily driven by wide variations in $U$ (Figure S11). With increasing occurrence and magnitude of high winds, the intensifying $k_{HW}$ increases air-sea CO$_2$ exchange. It is shown here that approximately 50% of the global CO$_2$ flux is attributed to the bubble-mediated exchange at high wind speeds. Hotspots of natural and anthropogenic CO$_2$ exchange and transfer to the ocean interior occur in high wind regions, such as the North Atlantic and the Southern Ocean. With current and projected increases in both intensity and frequency of extreme wind events (Young & Ribal, 2019; Young et al., 2011; Zheng et al., 2016), it is becoming necessary to quantify such fine-scaled processes that have a disproportionate impact on air-sea CO$_2$ fluxes.

Data Availability Statement

The cross-calibrated multi-platform (CCMP-V2) 6-hourly wind speed are publicly available at www.remss.com/measurements/ccmp. The climatological pCO$_{2w}$ data centered on 2005 was obtained from https://www.ldeo.columbia.edu/res/pi/CO2/carbondioxide/global_ph_data/obsfile.txt. The significant wave height and sea surface temperature data were obtained from https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-single-levels?tab=overview.

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