

Ocean Observations in Support of Studies and Forecasts of Tropical and Extratropical Cyclones

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Author contribution statement

RD, AKY, PCM, and GG elaborated the initial outline, and wrote the first draft of the manuscript. All 29 authors of this community white paper have provided input in the form of language, figures, and recommendations for the future of the ocean observing system in support of studies and forecasts of Tropical and Extratropical Cyclones.

Keywords

Weather extremes, Global Ocean Observing System (GOOS), Coupled Ocean-Atmosphere Weather Forecasts, ocean heat content (OHC), Barrier layer, Western boundary currents, Mesoscale Ocean Features, tropical cyclones, Extratropical Bomb Cyclones, Sea Surface Temperature, ocean mean temperature (OMT), natural hazards, Riverine flows, Cyclone intensity forecast, Coupled Ocean-Atmosphere Forecasts, Biologging

Abstract

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Over the past decade, measurements from the climate-oriented ocean observing system have been key to advancing the understanding of extreme weather events that originate and intensify over the ocean, such as tropical cyclones (TCs) and extratropical bomb cyclones (ECs). In order to foster further advancements to predict and better understand these extreme weather events, a need for a dedicated observing system component specifically to support studies and forecasts of TCs and ECs has been identified, but such a system has not yet been implemented. New technologies, pilot networks, targeted deployments of instruments, and state-of-the art coupled numerical forecast models have enabled advances in research and forecast capabilities and illustrate a potential framework for future development. Here, applications and key results made possible by the different ocean observing efforts in support of studies and forecasts of TCs and ECs as well as recent advances in observing technologies and strategies are reviewed. Then a vision and specific recommendations for the next decade are discussed.

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31 system have been key to advancing the understanding of extreme weather events that originate 32 and intensify over the ocean, such as tropical cyclones (TCs) and extratropical bomb cyclones 33 (ECs). In order to foster further advancements to predict and better understand these extreme 34 weather events, a need for a dedicated observing system component specifically to support 35 studies and forecasts of TCs and ECs has been identified, but such a system has not yet been 36 implemented. New technologies, pilot networks, targeted deployments of instruments, and stateof-the art coupled numerical forecast models have enabled advances in research and forecast 37 38 capabilities and illustrate a potential framework for future development. Here, applications and 39 key results made possible by the different ocean observing efforts in support of studies and 40 forecasts of TCs and ECs as well as recent advances in observing technologies and strategies are 41 reviewed. Then a vision and specific recommendations for the next decade are discussed. 42 Keywords: Tropical Cyclones, Extratropical Bomb Cyclones, Sea Surface Temperature, Ocean 43 44 Mean Temperature, Ocean Heat Content, Weather Extremes, Global Ocean Observing system, Natural Hazards, Barrier Layer, Riverine Flows, Cyclone Intensity Forecast, Coupled Ocean-45 Atmosphere Forecasts, Biologging 46

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49 **1. Introduction**

50 Extreme weather events are natural hazards that affect marine and terrestrial areas around 51 the world and are associated with different temporal and spatial scales (Elsner et al. 2008; 52 Menkes et al. 2016; Zhou et al. 2018). Tropical cyclones (TCs) and their extratropical 53 counterparts, often referred to as extratropical "bomb" cyclones (ECs), are deep low pressure 54 systems that produce and sustain high intensity winds. TCs develop exclusively over the ocean 55 (e.g. Knapp, et al., 2010), while ECs predominantly form in the land-ocean margin, in the vicinity of western boundary currents (e.g. Sanders and Gyaium, 1980), and can also 56 57 occasionally develop over land (e.g. Hakim et al. 1995). TCs and ECs are among the deadliest and most destructive types of extreme weather, often causing widespread damage due to strong 58 59 winds, storm surge, and heavy precipitation. Understanding their dynamical mechanisms and having the ability to accurately forecast them is a critical societal need but remains a challenge. 60

61 The understanding of upper-ocean processes leading to extreme weather events has largely 62 benefited from the climate-oriented sustained ocean observing system (e.g. Legler et al., 2015) 63 and observational process studies. Clearly, ocean observations are vital in ECs and TCs research 64 and forecasts, since they enable examining the details of air-sea interaction processes that can 65 lead to the formation and intensification of these systems (e.g., Leipper and Volgenau, 1972; Sanders and Gyakum 1980; Mainelli et al., 2008; Kuwano-Yoshida and Minobe, 2017). In 66 addition, ocean observations have been increasingly acknowledged by the forecast community as 67 68 a critical piece to improve extreme weather forecasts (e.g Dong et al., 2016). Their important 69 role will likely continue to increase in light of improvements in coupled model capabilities (see 70 below), and future extreme weather projections, which expects intense weather systems to 71 become more frequent (e.g Colle et al. 2015; Bacmeister et al., 2018). In fact, more and more 72 occurrences of record-breaking "Cat-6" types of TCs were observed over the last decade, with 73 three Cat-6 storms (>= 165 kts) forming during 2008-2018 (Haiyan, November, 2013; Patricia, 74 October, 2015; and Meranti, September, 2016), and no Cat-6 TCs recorded globally during 75 1999-2008. These recent changes and projections in occurrence of intense TCs and ECs emphasize the need for a sustained observing system in support of extreme weather studies and 76 77 forecasts.

78 Recent advances in research and ocean observing efforts during the past decade have 79 largely addressed specific recommendations identified by the scientific community during 80 OceanObs' 09 (e.g., Goni et al., 2010) about key areas to focus in support of TC studies and 81 forecasts. Since then, new technologies, pilot networks, and targeted deployments have further 82 expanded the reach of the global observing system and have been used in weather analyses, 83 outlooks, and improved ocean-atmosphere numerical forecasts models. For instance, (i) 84 extensive synergies between the scientific and operational communities continue to facilitate 85 transitioning of research results into operations (e.g., Shay et al., 2014); (ii) networks specifically designed in support of TC studies and forecasts have been implemented (e.g., Domingues et al., 86

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87 2015; Miles et al., 2015); (iii) targeted airborne profiling observations ahead of forecasted TCs

have been extensively implemented (e.g., D'Asaro et al., 2014; Jaimes et al., 2016; Meyers et al.,

89 2015; Zhang et al., 2018); (iv) substantial progress has been made toward understanding the role

90 of upper ocean salinity (e.g., Balaguru et al. 2012a, b; Domingues et al. 2015) and temperature

stratification (Price 2009; Lin et al. 2013b, a; Balaguru et al. 2015, 2018; Glenn et al. 2016) in

92 controlling TC intensity and development; (v) real-time assimilation of ocean observations into

coupled weather forecast systems continues to provide critical information for improved ocean
 representation (e.g., Chen et al., 2017; Dong et al., 2017); and (vi) new, state-of-the-art, coupled

95 numerical weather models have evolved and are now being used in experimental and operational

96 forecasting modes (e.g., Kim et al., 2014).

97 On a broader context, recent advances in ocean observations in support of extreme 98 weather studies and forecasts were accompanied by an overall improvement of global coverage 99 based on in situ and satellite ocean observations, and in their analysis. For example: improved 100 sensor technology and satellite coverage enabled substantial advances in satellite altimetry 101 (Verrier et al., 2018); satellite-derived salinity measurements (Meissner, et al., 2018), now 102 available with the launch of the Aquarius and the Soil-Moisture Active Passion (SMAP) 103 missions, have produced unprecedented spatial coverage of sea surface salinity observations 104 (Meissner et al., 2018); advances in in situ observing systems and networks (e.g Foltz et al., 105 2019; Goni et al., 2019; Todd et al., 2019; Roemmich et al., 2019) have also enabled many 106 groundbreaking research of the global oceans and climate system. In addition, community-wide 107 efforts aimed at advancing data availability, and data assimilation within high-resolution 108 numerical models through the US GODAE (Chassignet et al., 2009), the GODAE/Mercator-109 Ocean forecast system (Drévillon et al., 2008), and other similar efforts, have also provided 110 significant advances in oceanic forecasting capabilities. Currently, there are also ongoing from 111 the Environmental Modeling Center (EMC) within the National Oceanic and Atmospheric 112 Administration (NOAA) to implement a global data assimilation scheme into the Real-Time 113 Ocean Forecast System (RTOFS), and a similar scheme more specifically in support of hurricane 114 modeling systems. Some examples of such systems include the regional scale coupled HYbrid 115 Coordinate Ocean Model - Hurricane Weather Research and Forecast (HYCOM-HWRF), and 116 the HYCOM - Hurricanes in a Multi-scale Ocean-coupled Non-hydrostatic model (HYCOM-117 HMON). Additional details of such hurricane models are explored further in section 4 of this 118 manuscript.

In this community white paper, we describe some of the most important advances in the observational efforts in support of studies and forecasts of TCs and ECs since OceanObs' 09, and present and discuss some potential enhancements to ocean observing strategies for the upcoming future, as well as provide specific recommendations. This manuscript is organized as follows: sections 1.1 and 1.2 provide detailed descriptions of TCs and ECs, respectively, and discuss the role of the ocean in their genesis and intensification; in section 2, key components of the ocean observing system, products, and pilot networks in support of TC and/or ECs studies and forecasts

- 126 are described; section 3 presents selected applications of ocean observations in support research
- 127 into TCs and ECs; in section 4, the impact of ocean observations in the forecast of TC is
- 128 specifically addressed; section 5 describes recent advances and needs concerning data
- 129 management; and in section 6, the recommendations from the scientific community and the
- 130 vision for the upcoming decade are provided.
- 131

132 **1.1 Tropical Cyclones**

- 133 A TC is a fast-rotating storm system that is characterized by a low pressure center that forms in
- the tropics. TCs are observed in seven ocean basins worldwide (Figure 1), namely the: North
- 135 Atlantic, Northeast Pacific, Northwest Pacific, Southwest Indian, North Indian, Southeast Indian,
- and South Pacific. In the North Atlantic Ocean basin, for example, 11-12 named TCs, 6
- 137 hurricanes, and 2-3 major hurricanes develop between June and November in a typical year (e.g.
- 138 Landsea, 1993). TC development is associated with unstable atmospheric conditions, which are
- 139 primarily linked with boundary layer temperature (i.e., SST under the eyewall), and to a lesser
- 140 extent to the upper-level temperatures (see discussion in Balagura et al. 2015). In addition,
- 141 favorable atmospheric conditions for TC genesis and intensification, in order of importance,
- 142 include: (i) relatively little variation in the vertical profile of the environmental winds (low
- 143 vertical wind shear); (ii) an atmosphere characterized by lower-level convergence and upper-
- 144 level divergence; (iii) sufficient environmental mid-and upper-level moisture; (iv) and relatively
- slow (less than 8 m s⁻¹) deep layer steering. Favorable environmental conditions for TC
- 146 intensification have been thoroughly documented in many studies (e.g., DeMaria and Kaplan,
- 147 1994; DeMaria et al., 2010; Kaplan et al., 2010; Knaff et al., 2005; Knaff et al., 2018).
- Under favorable atmospheric conditions, sea surface temperatures (SSTs) above 26°C are needed for development and maintenance of TCs (e.g., Leipper and Volgenau 1972; Dare and McBride 2012). The vertical structure of the upper ocean plays a key role in the intensification and weakening of TCs with both thermal structure (e.g., Leipper and Volgenau, 1972; Shay et al., 2000; Mainelli et al., 2008; Goni et al., 2009; Shay and Brewster, 2010) and stratification (e.g., Lin et al., 2008; Price, 2009; Balaguru et al., 2012a, 2012b, 2015, 2018; Domingues et al., 2015; Emanuel, 2015; Huang et al., 2015; Seroka et al., 2016; Rudzin et al., 2017, 2018) being
- 155 important. The upper ocean heat content is an indicator of how much energy is potentially
- available for TC intensification, while the stratification can act as a barrier to TC-induced
- 157 mixing, suppressing upper ocean cooling, and helping maintain enthalpy fluxes from the ocean
- 158 into the TC. Intense upper ocean mixing events caused by the strong hurricane winds can quickly
- erode the thermal signature of subsurface warm or cold features (Pickard and Emery, 1990),
- 160 leading to SST misrepresenting the ocean thermal energy, and potential TC intensity. Several
- 161 studies (e.g., Mao et al., 2000; Shay et al., 2000; Ali et al., 2007, 2013; Mainelli et al., 2008; Lin
- 162 et al., 2013b) demonstrated the importance of ocean thermal energy, represented by warm ocean
- 163 features.

164 TC intensification involves the interaction of very complex mechanisms at a range of 165 scales, such as internal TC dynamics, upper ocean interaction, and atmospheric circulation. 166 Rapid intensification is often associated with TCs moving over warm ocean features (i.e., upper-167 ocean heat content values larger than 60 kJ cm⁻², Figure 2), which maintain warmer sea surface temperatures (due to suppression of TC-induced sea surface cooling) near the convective center 168 169 of the TC (Shay et al., 2000; Lin et al., 2005, 2009). The thermal energy across the sea surface is 170 central to the enthalpy fluxes that transport heat and moisture from the ocean to the atmosphere, 171 fuelling the TC. For this reason, ocean heat content (OHC) estimates from a variety of datasets 172 and methodologies are routinely used to provide operational guidance and carry out studies for intensity change and rapid intensification (e.g., Goni and Trinanes, 2003; DeMaria et al., 2005; 173 174 Kaplan et al., 2010; Shay and Brewster, 2010; Meyers et al., 2014; Rogers et al., 2017; Knaff et 175 al., 2018; Yamaguchi et al., 2018).

176 Despite these advances in understanding the role that the upper ocean plays in TC 177 intensification, officially-issued short-term intensity forecast errors have not been significantly 178 reduced over the past two decades in any basin (e.g., Figure. 3; DeMaria et al., 2007, 2014). One 179 of the factors contributing to the lag in improvement of TC intensity forecasts relative to TC 180 track forecasts may be the lack of a dedicated ocean observing system with sustained and 181 targeted ocean observations to correctly represent the ocean component in ocean-atmosphere coupled intensity forecast models. Nonetheless, operational intensity forecast tools and models 182 183 have been improving and are starting to reduce official intensity errors (DeMaria et al. 2014), 184 which may, in part, be due to improved atmospheric and ocean data assimilation (DA).

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186 **1.2. Extratropical Bomb Cyclones**

187 An EC is an extratropical cyclone that undergoes rapid deepening of its low pressure by 188 24 hPa or more in a period of 24 hours (i.e., 'explosive cyclogenesis'; Sanders and Gyakum 1980). This process is predominantly maritime, with seldom occurrences over continental land 189 190 masses. Though they are typically winter events, and their genesis involves processes distinct 191 from those associated with TC development, ECs produce winds as strong as hurricanes and are 192 often associated with large rainfall and storm surges. Explosive cyclogenesis is mainly observed 193 within the four basins, namely the Northwest Pacific, the North Atlantic, the Southwest Pacific, 194 and the South Atlantic (Black and Pezza, 2013). Though ECs in the North Atlantic basin occur 195 preferentially along the east coast of North America, in the vicinity of the Gulf Stream current 196 (Sanders and Gyajum, 1980), several cases of ECs have occurred offshore of western Europe (e. 197 g., Burt and Mansfield 1988; Young et al. 1987). Additionally, ECs do occasionally occur over 198 continents. Examples include the "Cleveland Superbomb" in January 1978 (Hakim et al. 1995) 199 and the more recent Colorado EC of March 2019.

200 Atmospheric baroclinic instabilities, upper-level vorticity coupling, and diabatic processes 201 have been acknowledged as the main mechanisms causing ECs genesis (Shapiro et al. 1999: 202 Yoshida and Asuma 2004; Kuwano-Yoshida and Asuma 2008). Over the North Atlantic ocean, the number of explosive cyclogenesis events is modulated by the North Atlantic Oscillation 203 204 (NAO), with positive NAO favoring a larger number of ECs due to stronger atmospheric 205 baroclinicity (Pinto et al., 2009). Large-scale heat convergence linked with the Atlantic 206 Meridional Overturning Circulation is also thought to influence atmospheric baroclinicity and 207 modulate EC activity (Gómara, et al., 2016). While these atmospheric and ocean conditions are 208 mostly linked with the number of ECs developing, the strength, intensification and trajectory of 209 the ECs are known for being directly influenced by upper ocean conditions (Kuwano-Yoshida 210 and Minobe, 2017), with their genesis often associated with oceanic frontal systems. For 211 example, the maximum frequency of explosive cyclogenesis and deepening is found to occur 212 nearthe Kuroshio or the Gulf Stream; these oceanic western boundary currents are associated 213 with large heat fluxes from the ocean to the atmosphere (Sanders and Gyakum, 1980; Ulbrich et 214 al. 2009; Hirata et al. 2015, 2016, 2018; Kuwano-Yoshida and Minobe 2017). Future projections 215 suggest that strong ECs will tend to increase, while the total number of extratropical cyclones is 216 expected to decrease (Colle et al. 2015; Chang 2017). Some of these climate model projections, 217 however, cannot fully resolve sharp gradients linked with the Kuroshio and the Gulf Stream 218 currents (Haarsma et al., 2016). Further studies based on observations, or on high-resolution 219 numerical simulations will likely be needed to confirm these trends.

220 In addition, the intensities of ECs that develop over warmer ocean regions are usually 221 underestimated when compared to those that develop over cooler ocean regions (Kuwano-222 Yoshida and Enomoto, 2013). This is because cloud condensation associated with latent heat 223 release over warm ocean areas is more important for rapid development than the upper-level 224 vorticity forcing (Catto et al., 2010). For example, recent analyses using satellite-based highresolution SST measurements suggest that ECs are affected by SST fluctuations associated with 225 226 fronts and mesoscale eddies around western boundary currents (Booth et al., 2012; Kuwano-227 Yoshida and Minobe, 2017; Hirata et al., 2015; 2016; 2018). Considering that the western 228 boundary currents such as the Gulf Stream and the Kuroshio are warming more rapidly than the 229 global average rate (Wu et al., 2012), the ocean's effect on ECs generation is likely to increase 230 in the future. With respect to ocean response to ECs winds, recent ocean simulations suggest that 231 ECs can induce surface horizontal divergence and upwelling reaching depths of 6,000 m, which 232 can impact the deep ocean circulation and ecosystems through mixing and bio-geochemical 233 transport (Kuwano-Yoshida et al., 2017).

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236 2. Ocean Observations and Parameters in Support of Extreme Weather Studies and 237 Forecasts

The Global Ocean Observing System (GOOS) provides measurements from a diverse suite of observing platforms that enable studies of the complex dynamics of extreme weather systems and help improve the overall skill of extreme weather forecast models. Accurate TC and EC forecasts based on coupled models, for example, require a correct representation of the upper-ocean heat content, vertical density structure, and the mesoscale eddy field. This in turn requires upper ocean observations with high spatial and temporal resolution.

Analyses of ocean observations in the vicinity of TCs and ECs have led to improved understanding of their development and intensification, which occur over distinct geographic domains and during different seasons. The GOOS includes several multi-national ocean observing efforts that support studies and forecasts of both TCs and ECs. While the observational requirements and needs for TCs and ECs are different, some of observing platforms are used in support of studies on both types of storms, and some are specifically used

250 to provide observations for TCs or ECs.

Here we discuss the importance of the integrated ocean observing system and of targeted ocean observations, focusing on their application to TCs and ECs. We also provide an overview of these various components based on several successful examples, which illustrate applications that are helping understand the dynamics of these extreme weather systems, and are also helping to improve the overall skill of their forecast. To be most effective for operational forecasting, observing platforms should transmit their data in real-time via the Global Telecommunications System (GTS).

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259 2.1. Ocean Observations

260 <u>Satellites:</u> Satellite-derived fields of SST¹ and sea surface height (SSH)² are used to estimate the 261 upper ocean heat content (OHC; Leiper and Volgenau, 1972), which is sometimes also referred 262 to as Tropical Cyclone Heat Potential³ (TCHP; Goni and Trinanes, 2003). The TCHP/OHC is 263 defined as the excess heat content in the water column from the sea surface to the depth of the

defined as the excess heat content in the water column from the sea surface to the depth

¹NOAA High-resolution Blended Analysis of Daily SST available at : https://www.esrl.noaa.gov/psd/data/gridded/data.noaa.oisst.v2.highres.html

²Satellite altimetry products made available by the Copernicus Marine Environment Monitoring Service: http://marine.copernicus.eu/

³TCHP fields made available by the NOAA Atlantic Oceanographic and Meteorological Laboratory (NOAA/AOML) at: http://www.aoml.noaa.gov/phod/cyclone/data/

264 26°C isotherm. TCHP/OHC fields can provide key qualitative and quantitative spatial 265 information about areas where TCs may intensify, mainly by identifying the location and thermal characteristics of the oceanic upper layer, including warm eddies and current frontal regions 266 267 (Meyers et al., 2014). For example, high values of TCHP (larger than approximately 50 kJ cm⁻²) 268 have been shown to be linked to intensification of Atlantic hurricanes (e.g., Mainelli et al., 269 2008). It should be acknowledged, however, that OHC from satellite-derived SST and SSHA can 270 sometimes result in biased estimates in regions that are strongly influenced by freshwater 271 sources. In the Bay of Bengal, for example, where river water can persist in the near surface 272 layer, salinity can have a dominant role in determining the subsurface density structure and the SSH (e.g., Yu and McPhaden, 2011); this can have a detrimental effect on the TCHP derived 273 274 from satellite observations. Nevertheless, comparison of TCHP values derived from satellites 275 and in situ observations in the Bay of Bengal has shown that satellite-derived estimates are generally unbiased, and estimates with a precision better than 20 kJ cm⁻² are often obtained 276

277 (Nagamani et al., 2012).

278 Fields of SST and TCHP/OHC are routinely used by the NOAA National Hurricane Center

279 (NHC) and the Joint Typhoon Warning Center for their subjective TC intensity forecasts and

280 quantitatively in the Statistical Hurricane Intensity Prediction Scheme (SHIPS; DeMaria and

281 Kaplan, 1994; DeMaria et al., 2005) and rapid intensification aids (Kaplan et al., 2010; Knaff et

al., 2018). Notable examples of use of satellite fields to assess links between the ocean and

- hurricane intensification include Hurricane Opal (Shay et al., 2000), super-typhoon Maemi (Lin
- et al., 2005), Hurricane Katrina (Figure 2; Goni et al., 2009), and "killer cyclone" Nargis (Lin et
 al., 2009).

286 Underwater gliders: Autonomous underwater gliders (Rudnick, 2016) provide measurements of 287 temperature and salinity in the upper several hundred meters and are becoming key components 288 of the ocean observing system (Liblik et al., 2016; Testor et al., 2019). In addition to the standard 289 measurements of temperature, salinity, and depth-average currents, gliders can be equipped to 290 measure current profiles (e.g., Todd et al., 2017), bio-optical properties, dissolved oxygen, and 291 turbulent microstructure. Because of their adaptability and versatility, gliders fill important 292 observational gaps in the ocean observing system (Liblik et al., 2016), particularly with respect 293 to TC intensity forecasts. Glider observations are, for example, frequently assimilated into 294 ocean-atmosphere coupled models and used for hurricane intensity forecasts. Sustained glider 295 deployments monitor upper ocean conditions in areas frequently impacted by TCs (e.g., 296 Domingues et al., 2015; Miles et al., 2015; Todd et al., 2018) and are part of the NOAA 297 Hurricane Field Program⁴. Gliders also have particular utility measuring ocean processes on 298 continental shelves before and during landfall of TCs, where alternative ocean observations are

- scarce (e.g., Glenn et al., 2016; Miles et al., 2017). In addition to measuring physical variables,
- 300 gliders can carry specialized sensor payloads; sensors for key biogeochemical variables offer the

⁴ NOAA Hurricane Field Program: https://www.aoml.noaa.gov/hrd/programs_sub/HFP.html

301 promise of advancing our understanding of the role of TCs and ECs on ecosystems. For instance,

- 302 dissolved oxygen measurements could help to characterize the storm-driven ventilation of
- 303 subsurface waters in areas with oxygen minimum zones (e.g., the Arabian Sea; Morrison et al.,
- 304 1999).

305 Since gliders move slowly (about 25 km/day) compared to most atmospheric cyclones *O*(300

- 306 km/day), actively piloting them into the paths of storms is generally not feasible due to the short
- lead times of forecasts. Thus, sustained deployment of gliders at locations prone to tropical or
 extratropical cyclones (e.g., Domingues et al., 2015, Glenn et al., 2016; Perry et al., 2017) or
- along oceanic boundaries as part of boundary current observing systems (Todd et al., 2018;
- 310 Testor et al., 2019; Todd et al., 2019) is preferable. Compared to rapid response deployments
- 311 (e.g., Miles et al., 2015; Goni et al., 2017), sustained glider surveillance has the distinct
- 312 advantage of providing critical high resolution observations in the open ocean and continental
- 313 shelf break prior to storm arrival; these observations have been shown to improve the
- 314 representation of the ocean in operational coupled forecasts models of hurricane intensity (e.g.,
- 315 Dong et al., 2017). Most underwater glider data collected in support of Atlantic Hurricane
- 316 studies and forecasts are transmitted in real-time into the GTS and U.S. Integrated Ocean
- 317 Observing System (IOOS) underwater glider data assembly center⁵, and made available for
- 318 immediate use by operational forecast centers.

Surface Drifters: Different types of drifters provide observations of ocean current velocities, 319 320 SST, and sea level pressure (SLP) that are also used in support of weather forecasts, including 321 TCs and ECs. Sustained global observations from drifters are used for constraining satellite SST 322 errors and biases (e.g., Zhang et al., 2009) and have had a positive impact on global weather 323 forecast throughout the troposphere thanks to assimilation of in-situ SLP observations 324 (Centurioni et al. 2017; Horánvi et al. 2017; Ingleby and Isaksen, 2018). Horánvi et al. (2017) 325 showed that, in the case of intense cyclogenesis, SLP observations from drifters made possible 326 large reduction in forecast errors, sometimes the largest among all the other assimilated 327 observations. Furthermore, targeted deployments of drifters are sometimes carried out in front of 328 TCs. During the Atlantic hurricane season, for example, Surface Velocity Program (SVP), 329 Surface Velocity Program Barometer (SVPB) and Autonomous Drifting Ocean Station drifters 330 (Centurioni, 2010; see also Centurioni et al., 2018, for a complete description of the drifter 331 technology) are often air-deployed in front of TCs that may impact the US mainland. New 332 drifters capable of measuring the directional wave spectra of surface gravity waves, termed 333 Directional Wave Spectra Drifters (Centurioni et. al, 2016; Centurioni et al., 2019) have also 334 been deployed ahead of TCs. Successful deployments of various drifters have been carried out in 335 the Atlantic Ocean during the following TCs: Fabian (2003), Frances (2004), Rita (2005), Dean 336 (2007), Gustav (2008), Ike (2008), Isaac (2012) and Michael (2018), and in the western Pacific

337 Ocean during Hagupit (2008), Jangmi (2008), Fanapi (2010), and Malakas (2010). In addition,

⁵U.S. Glider Data Assembly Center: https://gliders.ioos.us/data/

drifters that are part of the Global Drifter Program⁶ array, often come close to tropical cyclones

- and provide valuable SST and SLP observations. In the 2013-2018 period a total of 116 SVP and
- 340 SVPB drifters were located within 55 km of the tracks of systems that eventually developed into
- 341 hurricanes (Centurioni et al., 2019). However, the effect of SLP drifter observation in improving
- 342 TC track forecast has yet to be studied.

343 Air deployed profiling instruments: Airborne profiling instruments are often deployed in targeted 344 sampling mode in front of TCs in the Atlantic and Pacific basins. In the Atlantic, deployments of 345 Airborne eXpendables Bathythermographs (AXBTs), Airborne eXpendable Conductivity Temperature and Depth (AXCTDs), and Airborne eXpendable Current Profilers (AXCPs) are 346 347 generally conducted to sample ocean conditions ahead and under TCs as part of the NOAA 348 Hurricane Field Program (Meyers et al., 2015; Jaimes et al., 2016; Zhang et al., 2018). Paired 349 deployments of AXBTs and dropsondes, for example, provide collocated measurements of SST, 350 air temperature, humidity and TC wind speed that allow for the estimation of bulk air-sea fluxes. 351 Exchange coefficients used in such computations are based on direct flux data (Zhang et al.

- 352 2008). These data provide valuable information that is used to evaluate and improve TC model
- 353 physics such as boundary-layer parameterizations (e.g., Zhang et al., 2015). In recent years, the
- following Atlantic TCs were sampled: Edouard (2014), Harvey (2017), Irma (2017), Maria
- 355 (2017), Nate (2017), and Michael (2018). In the Pacific, paired deployments of AXBTs and
- dropsondes were also carried out during the 2010 Impact of Typhoon on Pacific (ITOP)
- international field experiment (D'Asaro et al., 2014, see Section 3.6).
- 358 More recently, the Air-Launched Autonomous Micro-Observer (ALAMO) (Jayne and Bogue,
- 2017) profiling float was developed to be deployed from an aircraft Sonobuoy-A size tube,
- 360 similarly to AXBTs (Sanabia et al., 2013). A key distinction between ALAMO floats and other
- 361 airborne expendable profiling instruments is that these floats are capable of sampling hundreds
- 362 of profiles continuously before, during, and after the passage of TCs for up to 6 months,
- depending on instrument configurations. During the 2014-2018 period, a total of 60 ALAMO
- 364 floats were deployed in support of the NOAA Hurricane Field Program in both the Atlantic and
- 365 eastern North Pacific basins.
- 366 *Profiling Floats:* Profiling floats (e.g., Riser et al., 2016; Roemmich et al., 2019) offer the
- 367 advantage of providing a sustained long-term and large-scale record of most of global oceans.
- 368 Temperature and salinity observations from Argo floats⁷ are routinely assimilated into
- 369 operational ocean models (e.g., Chassignet et al., 2009) that are used to initialize the ocean
- 370 component of hurricane forecast models. In addition, operational Argo floats have been found to
- 371 be very important in regions where routine and/or opportunistic airborne AXBTs observations

⁶The Global Drifter Program: http://www.aoml.noaa.gov/phod/gdp/index.php and http//gdp.ucsd.edu

⁷Argo Global Data Assembly Center at: http://www.usgodae.org/argo/argo.html

are lacking, e.g., in the case of "Cat-6" supertyphoon Haiyan which devastated the Philippines in
2013 (Lin et al., 2014).

374 Profiling floats also offer valuable information of upper ocean processes contributing to ECs 375 formation and intensification. The typical observation interval for individual Argo floats, 376 however, is 10 days, which is generally too long to capture the rapid air-sea interactions 377 associated with enthalpy fluxes and exchanges. High-frequency and adaptive fine scale profiling 378 float sampling are generally needed to fully capture mesoscale ocean features usually associated 379 with storm intensification, and also to characterize the storm-induced upper ocean response in 380 detail. Argo floats under developing ECs were used to obtain 73 high-frequency profiles of the 381 upper 700 m at 6-hour intervals during the 2015/16 and 2016/17 winters. These data were critical 382 to understanding ocean changes under ECs in the Northwestern Pacific (Kuwano-Yoshida et al., 383 2018) and emphasize the advantage of adaptive profiling float sampling using two-way 384 communication systems. Observation interval and depth of modern Argo floats can be controlled 385 using two-way satellite communication. Interactive operation of the floats with satellite and 386 assimilated data enables high-frequency and high-resolution observation at fronts, i.e. the floats 387 observe short interval if satellite and assimilated data suggest that the floats are located near the 388 SST fronts. These in-situ observations will complement satellite observations, increase temporal 389 sampling, and enable resolving the fine structure associated with SST fronts that may help

- 390 improve EC and TC forecasts.
- 391 In addition, expanding coverage of Biogeochemical Argo float observations also offer
- 392 opportunities for evaluating phytoplankton response to mixing forced by TCs and ECs (e.g.,
- 393 Chacko, 2017) and studying the role that these extreme weather events play in ventilating
- 394 subsurface waters in oxygen minimum zones. These new applications of profiling floats will
- 395 enable detailed investigations of the upper ocean processes involved in EC intensification and
- 396 the role that these extreme weather events play in the ocean biogeochemistry within their main
- 397 formation basins, such as the Northwest Pacific and North Atlantic ocean.

398 EM-APEX (Electromagnetic Autonomous Profiling Explorer) floats were developed to measure 399 profiles of upper ocean temperature, salinity and currents. Velocity estimates are based on 400 measuring the voltage induced by seawater moving through the earth's magnetic field as first 401 pioneered with expendable current profilers (Sanford et al., 1987; Shay et al., 1992) and later 402 added to standard profiling floats (Sanford et al., 2007; 2011). EM-APEX floats were 403 successfully air-deployed in front of Hurricane Frances (2004), during the ITOP missions in the 404 western Pacific (2010), and more recently during Hurricane Michael (2018). The EM-APEX 405 floats can profile to depths of 2000 m or over a specific depth region such as the mixed laver 406 through the seasonal thermocline. Profiling configurations (i.e. sampling rate, and depth) can be 407 changed via two-way Iridium communications, allowing for significant flexibility in an adaptive 408 sampling before, during, and after storm passage, capturing air-sea interactions, and the oceanic 409 response for several weeks following passage. For example, during Hurricane Michael (2018), a 410 EM-APEX float sampled from 30 to 300 m during the storm at one hour intervals to assess the

411 role of current shear in vertical mixing processes to evaluate model parameterizations. With the

- 412 high resolution measurements, the evolution of the Richardson numbers could be determined at
- 413 vertical resolution of 2-4 m in an active entrainment zone. In addition, the momentum flux from
- the surface wind stress into the surface mixed layer provides a method to back out the surface
- 415 drag coefficient that are needed in examining the complex air-sea interactions that occur during
- tropical cyclone passage (e.g., Shay and Brewster, 2010).

417 *Biologging* has also been used to collect in-situ meteorological and physical oceanographic

- 418 observations. Marine mammals (Campagna et al., 2000; Boehlert et al., 2001; Boyd et al., 2001),
- 419 seabirds (Koudil et al., 2000; Watanuki et al., 2001; Charrassin et al., 2002; Wilson et al., 2002),
- 420 sea turtles (Narazaki et al. 2015; Fukuoka et al., 2015) and fish (Block et al., 2001) have been
- 421 adopted as autonomous samplers of oceanographic parameters such as temperature, conductivity
- 422 and depth (Charrassin et al., 2002; Koudil et al., 2000; Watanuki et al., 2001; Wilson et al.,
- 423 2002; Block et al., 2001; Fedak, 2013). These animals often live around western boundary
- 424 currents and frontal systems where TCs and ECs are often observed (Figure 4). In addition, Bio-
- 425 logging by sea turtles observed daily temperature profiles in surface layers in the northwestern

426 Pacific (Narazaki et al. 2015; Fukuoka et al. 2015). The profiles were collected during 77 ECs

427 events in 6 consecutive winters. Both Argo floats and Biologging captured rapid temperature

428 changes under ECs. These ocean observations are crucial to identify near-surface baroclinic 429 zones and ocean-atmosphere fluxes of heat and moisture. Such processes are crucial to the

- 429 zones and ocean-atmosphere fluxes of heat and moisture. Such processes430 successful predictions and simulations of ECs.
- 431

432 **2.2 Ocean Metrics for TC/EC Intensification Studies**

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434 Various parameters derived from in situ and remotely sensed observations that describe 435 energy available in the upper ocean have been used to estimate the potential for TC and EC 436 development and intensification. The TCHP/OHC is one example, but it has dimensions that 437 differ from SST and so cannot be used in place of SST in numerical models. Moreover, Price 438 (2009) demonstrated that depth-averaged temperature is a more robust metric of hurricane-ocean 439 interaction than is OHC. Ali et al. (2015) then used satellite-derived OHC and the depth of the 26 440 ^oC isotherm (D26) to estimate ocean mean temperature (OMT) with a few assumptions. This 441 OMT was a better predictor for Indian monsoon rainfall than SST (Ali et al. 2015; Venugopal et 442 al., 2018). Use OMT in place of SST in numerical models offers potential for improvement in

443 cyclone forecasting.

Thermodynamically, the subsurface ocean affects TCs through its control of TC-induced cold SST wakes. When the cold wake is weak (less than about 0.5°C), TCHP is a very good predictor of TC intensification, exceeding the skill of other predictors such as SST and vertically-averaged temperature (Figure 5; Balaguru et al., 2018). However, when the cold wake is strong (>0.5°C) (e.g., when SST is very warm and temperature stratification is shallow) ocean dynamic temperature (Tdy) performs significantly better (Figure 5; Balaguru et al., 2018). Tdy is

- defined as the ocean temperature averaged from the surface to the post-storm mixed layer depth,
- 451 which depends on the upper-ocean stratification as well as the TC intensity and translation speed
- 452 (Balaguru et al., 2015). An additional limitation of temperature-based metrics such as TCHP and
- 453 OMT is that they do not account for salinity, which has been shown to be important in all TC
- 454 basins (Wang et al., 2011; Balaguru et al., 2012b; Grodsky et al., 2012; Neetu et al., 2012;
- 455 Domingues et al., 2016; Foltz and Balaguru, 2016; Balaguru et al., 2016). Satellite-based sea
- 456 surface salinity measurements, when combined with subsurface in situ observations, can provide
- 457 further information about the salinity stratification. This additional metric, when incorporated
- 458 into Tdy may result in further improvements to statistical TC prediction schemes and enable
- 459 more meaningful validation of operational ocean analyses and forecasts.
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- 461

3. Highlighted Applications and Results

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In this section, key results, sampling strategies, and applications of ocean observations in support of studies and forecasts of TCs and ECs are described. These case studies provide additional information on some of the successful examples of employing data derived from the GOOS, new pilot networks, and targeted deployments to enhance our understanding of the ocean-atmosphere interaction processes that can lead to TC and EC intensification.

468

3.1. The 2011 and 2012 Atlantic Hurricane Seasons: hurricanes Irene (2011) and Sandy (2012).

471 Over the broad continental shelf of the Middle Atlantic Bight along the US East Coast, 472 research carried out with gliders observations has shown that cool subsurface waters (i.e., the 473 "Cold Pool"; Houghton et al., 1982) can be mixed with the surface waters under intense wind 474 conditions, thereby impacting storm intensity. Since the Cold Pool is obscured from the view of 475 satellites, in-situ observations, such as those obtained by gliders, are needed to capture its 476 properties and impact on cyclone intensity (Glenn et al., 2016; Seroka et al., 2016). For example, 477 a glider deployed ahead of Hurricane Irene (2011) observed larger than usual ahead-of-eye-478 center cooling of over 6°C (Seroka et al., 2016) caused by intense mixing of surface waters with 479 cold subsurface waters forced by the hurricane winds. Subsequent ocean and atmosphere model 480 sensitivity studies identified this process as the missing component necessary to capture Irene's 481 rapid weakening just prior to landfall. In contrast, glider observations collected during Hurricane 482 Sandy (2012) showed that the storm winds were downwelling favorable and led to offshore 483 advection of the subsurface Cold Pool waters, which prevented upper ocean cooling and favored 484 the sustained intensity of Sandy (Miles et al., 2015, 2017).

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486

487 **3.2. The 2014 Atlantic Hurricane Season: Hurricanes Gonzalo (2014) and Fay (2014)**

488 Studies carried out using all ocean observations, including those from underwater gliders, 489 in the western tropical Atlantic and Caribbean Sea, were used to assess the pre- and post-storm 490 ocean conditions associated with Hurricanes Fay and Gonzalo (2014). When Hurricane Gonzalo 491 passed north of Puerto Rico, the general background ocean conditions were provided by Argo 492 floats and satellite-derived SST and SSH fields. In addition, there was one glider surveying upper 493 ocean temperature and salinity structure in the vicinity of the projected path of Gonzalo (Figure 494 6a). This glider was the only observing platform to capture the presence of a 20-m-thick barrier 495 layer (Domingues et al., 2015), a salinity stratified layer (e.g., Sprintall and Tomczak, 1992) 496 within the deeper isothermal layer. This layer inhibited vertical mixing and limited surface 497 cooling forced by Gonzalo's winds to only 0.4°C, allowing the storm to intensify to Cat-4 498 (Domingues et al., 2015). When Gonzalo subsequently crossed the path of Fay near Bermuda 499 (Figure 6a), it weakened from Cat-3 to Cat-2 due to the upper ocean cooling of approximately 500 4°C observed in the wake of Fay (Goni et al., 2017).

501

3.3. The 2017 Atlantic Hurricane Season: Hurricanes Harvey, Irma, Jose, Maria, and Nate 503

The 2017 Atlantic hurricane season was one of the most active in recent history with 17 named storms, and six major hurricanes. Underwater gliders, profiling floats, XBTs, airborne observations, and other observing platforms collected crucial ocean data to assess upper ocean conditions and changes before, during, and after the passage of multiple hurricanes. Here we describe ocean observations and key results from Hurricanes Harvey, Irma, Maria, Jose, and Nate. Data from the ocean observing system were used in support of operational hurricane intensity forecasts.

511 In August, Hurricane Harvey developed in the tropical Atlantic and passed through the 512 Caribbean Sea south of Puerto Rico. In this area, observations from one underwater glider 513 showed that a relatively shallow mixed layer favored cooling of the upper ocean. Together with 514 the moderate wind shear, this contributed to Harvey's lack of intensification within the 515 Caribbean. Once Harvey reached the Gulf of Mexico, where OHC derived from Argo floats was 516 at a record level and SST exceeded 30 °C (Trenberth et al., 2018), it intensified from a tropical depression (16 m s⁻¹ / 56 km/h sustained winds) into a Cat-4 hurricane (59 m s⁻¹ / 212 km/h 517 518 sustained winds) in less than 48 hours before making landfall along the Texas coast with 519 devastating effects.

520 In September, SST values of ~30°C were observed across the western Atlantic and 521 Caribbean (Figure 7), which, along with low wind shear, helped sustain the development and 522 intensification of Hurricanes Irma, Maria and Jose (Camp et al., 2018). Hurricane Irma, the 523 strongest TC globally in 2017, reached its maximum intensity (Cat-5) on September 6, while 524 traveling over waters north of Puerto Rico and Hispaniola. Observations from underwater gliders 525 showed that a fresh water barrier layer ~15 m thick (Figure 8a) inhibited mixing between the upper ocean and colder underlying waters, similar to Hurricane Gonzalo (2014; Domingues et 526 al., 2015, Dong et al., 2017). These observations also revealed that the upper 50 m of the ocean 527 528 cooled by approximately 1°C as a result of storm-induced mixing. A few days after Irma, 529 Hurricane Jose passed 2-3 degrees of latitude to the north of Irma's track. Jose's trajectory 530 coincided with the cold wake left by Hurricane Irma, so it interacted with a relatively cooler and 531 well mixed upper-ocean as observed by underwater glider data. These cooler ocean conditions 532 may have contributed to its weakening from Cat-4 to Cat-3. Later in the month, Hurricane Maria 533 passed through the Caribbean Sea and then the same area as Irma in the tropical North Atlantic. 534 Following a landfall in Dominica, Maria reached peak intensity on September 20 with maximum 535 sustained winds of 78 m s⁻¹(280 km/h). Underwater glider observations revealed the existence of 536 a very stable barrier layer that was approximately 30 m deep along the path of Maria (Figure 8a), 537 providing favorable ocean conditions for intensification. On September 20, Maria made landfall 538 in Puerto Rico as an intense Cat-4 hurricane. By the end of September, positive SST anomalies 539 recorded before the passage of these hurricanes had dissipated due to the intense mixing caused 540 by these major storms, and SSTs closer to neutral conditions were observed (Figure 7a). Farther north, Todd et al. (2018) used glider observations and volume transport measurements in the 541 542 Florida Straits to show that the Gulf Stream exhibited a large freshwater anomaly that was 543 attributable to rains from Irma and also a transient reduction in volume transport that was 544 attributable to wind forcing associated with the passing storms; further studies with numerical 545 simulations are needed to better understand the dynamics of the storm impacts on the western 546 boundary current.

547 In October, Hurricane Nate developed and steadily gained strength over warm waters of 548 the northwestern Caribbean Sea. Once Nate reached the Gulf of Mexico, EM-APEX floats 549 located near the projected track (Figure 9) were reprogrammed to profile every 2 to 4 hours. 550 returning vertical profiles of temperature, salinity, currents, dissolved oxygen, chlorophyll 551 fluorescence, backscatter as a proxy of particle concentration, and chromophoric dissolved 552 organic carbon. In addition, one hundred and forty AXCPs and AXCTDs were deployed from 553 the NOAA WP-3D aircraft prior to, during and after Nate (Figure 9). These observations showed an upper ocean velocity response with magnitude of 0.5-0.75 m s⁻¹ and rotation of the current 554 555 vectors with increasing depth that led to strong current shear at depths of 40-60 m. The 556 development of strong shear favored the deepening of the oceanic mixed layer under Nate by 10 557 to 15 m and mixed layer cooling of 1.5 - 2 °C. The observed response was predominantly nearinertial in character, and likely impacted the air-sea fluxes and the intensity and structure of the 558 559 storm (e.g., Jaimes et al., 2016).

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562 **3.4. Impact of Riverine Outflows on Tropical Cyclones**

564 Areas in the Caribbean Sea and Tropical North Atlantic, where hurricanes commonly 565 intensify, are sensitive to different freshwater sources, including major rivers such as the 566 Amazon and Orinoco (e.g., Kelly et al., 2000; Balaguru et al., 2012a; Johns et al., 2014), and the 567 Mississippi River (Goni and Domingues, 2019), which can contribute to the formation of barrier 568 layers. In regions under the influence of strong fresh water sources, such as in the Bay of Bengal, 569 low salinity conditions at the surface may sustain thermal inversions in the upper layer, which 570 may further help suppress the TC-induced SST cooling (e.g., Sengupta et al., 2008). Barrier 571 layers can be tens of meters thick, and have been indicated as a potential contributor to the rapid 572 intensification of several TCs worldwide (e.g., Balaguru et al., 2012b).

573 Several major Atlantic hurricanes in 2017 encountered pre-existing barrier layer 574 conditions along their trajectories (Section 3.3, Figure 8a). Analysis of satellite-derived chlorophyll data⁸ for August 2017 (Figure 8b) indicates that freshwater plumes from the Amazon 575 576 and Orinoco rivers were advected into the Caribbean basin, contributing to barrier layer 577 formation. Comparison with historical chlorophyll data for the tropical North Atlantic Ocean and 578 Caribbean Sea (Figure 8c,d) suggests that entrainment of freshwater plumes from these rivers 579 into the basin-scale circulation in 2017 may have caused unusually strong freshwater transport 580 into these areas. While investigation is still ongoing to assess the potential impact of these 581 freshwater conditions on the 2017 hurricanes, these results emphasize that the correct 582 representation of salinity conditions within coupled TC forecast models can be key to produce 583 accurate hurricane predictions. This may be especially true for areas that are particularly 584 sensitive to large freshwater sources, such as the Caribbean Sea, Gulf of Mexico, and tropical 585 North Atlantic Ocean.

586

587 **3.5. Development of BioLogging as an ocean observation platform for ECs**

588

589 Flight and drift paths of sea birds soaring and floating over the ocean surface enable 590 measurement of fine-scale winds and currents. Yonehara et al. (2016) and Goto et al. (2017) 591 found out that fine-scale flight trajectories by recording one position per second and minute 592 provide 5-min to 1-hour interval surface wind direction and speed along the trajectories. The 593 bird-estimated wind directions showed good agreement with those from satellite, although wind 594 speeds were slower than satellite winds because sea birds flew at lower altitudes than 10 m at 595 which satellite winds were calibrated. Only the wind estimates from three birds had meaningful 596 impact on data assimilation when severe rainfall occurred in Japan associated with two typhoons

⁸ NASA Ocean Color website: https://oceancolor.gsfc.nasa.gov/

using regional numerical forecast system (Wada et al., 2017). Yoda et al. (2014) developed a

- new method for obtaining in-situ ocean current measurements by using sea birds with
- 599 GPS/GNSS loggers floating at the surface as Lagrangian current sensors akin to drifting buoys.
- 600 The sea birds forage boundary areas between two oceanic mesoscale eddies where primary
- 601 productivity and prey density are thought to be high. The current data from sea birds improved
- 602 reproducibility of eddies through data assimilation into an operational ocean nowcast/forecast

603 system (Miyazawa et al., 2015).

604 Biologging of temperature and salinity measurements derived from turtles also has the 605 potential for improving numerical simulations in support of EC forecasts. Loggerhead turtles, for 606 example, favor waters warmer than 15°C, which corresponds to the northern edge of the 607 Kuroshio and its extension near the surface in winter. A feasibility study for data assimilation of 608 temperature measurements by the turtles suggests that the turtle measurements captured the 609 warm core rings separating from the Kuroshio Extension better than the Oyashio intrusion branches (Miyazawa et al., 2019). The improved ocean representation of such features may allow 610 for better EC forecast through a more accurate simulation of air-sea interaction fluxes associated 611 612 with these warm ocean rings and meanders.

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615 **3.6. The 2010 ITOP field campaign**616

617 The ITOP international field campaign in the western North Pacific Ocean is an important 618 example for future field observation strategy and planning (Figure 10; D'Asaro et al., 2014). The 619 western North Pacific was chosen because this basin is where the largest number and the most 620 intense TCs are usually recorded (Figure 1). In the summer of 2010, the ITOP field campaign used targeted aircraft AXBT observations to collect the pre-storm temperature profiles ahead of 621 622 three TCs of distinct intensity: Megi, Fanapi, and Malakas. Supertyphoon Megi (with peak 623 intensity 82 m s⁻¹/296 km/h, Cat-5) was the most intense TC recorded globally until 2010, while 624 Fanapi was a Cat-3 moderate TC and Malakas was a Cat-2 TC. The pre-TC ocean conditions 625 were different for these three TCs (Figure 11a). Among the three, Megi intensified over warm 626 ocean temperatures, characterized by TCHP values larger than 140 kJ cm⁻², and D26 of 120 m. 627 In contrast, both Fanapi and Malakas travelled over waters with shallower D26 and TCHP values 628 lower than 100 kJ cm⁻². Analysis of the available ocean observations revealed that these large 629 differences in upper ocean heat content played a key role in the intensification of these TCs (Lin 630 et al., 2013a; D'Asaro et al., 2014).

In addition to assessing the pre-TC ocean conditions using AXBT profiles, the paired ocean-atmosphere observations during TC intensification were also collected. These observations were used to evaluate air-sea sensible and latent heat fluxes. The correct representation of these fluxes is needed to obtain accurate TC intensification forecasts since they represent the ocean-to-atmosphere energy transfer that leads to TC intensification. Direct observations of air-sea fluxes were obtained deploying co-incident/co-located atmospheric

- 637 dropsoundes and ocean AXBTs during TC-penetration flights (see Figure 11c). With these
- 638 unique observations obtained during ITOP, accurate air-sea sensible and latent fluxes were
- obtained (Lin et al., 2013a; D'Asaro et al., 2014), revealing that enthalpy fluxes were
- 640 substantially larger during Supertyphoon Megi as it reached Cat-2 (Figure 11c) and then
- 641 continued to intensify into a Cat-5 Supertyphoon. Results from ITOP emphasize the value of
- 642 paired, co-located, ocean-atmosphere observations to improve model prediction performance and
- 643 for improving our understanding on the role that different types of ocean conditions can play in
- 644 the TC intensification processes.
- 645

646 **4. Impact of Ocean Data in Tropical Cyclone Intensity Forecasts**

647 A variety of observations collected near TCs in recent years have impacted the fidelity of 648 TCs forecasts, typically by reducing errors and biases in analyses used to initialize the ocean

649 component of coupled prediction models. For example, Halliwell et al. (2011) analyzed the

650 impact of multiple factors toward reducing errors in HYCOM ocean analyses in the Gulf of

651 Mexico prior to Hurricane Isaac (2005). They determined that assimilation of ocean observations

652 is a leading-order factor in reducing initialization errors in comparison to ocean model attributes

653 such as vertical mixing and surface flux parameterizations, along with model resolution.

654 More recently, Dong et al. (2017) conducted observing system experiments (OSE) focused 655 on the influence of conventional ocean observing systems plus underwater glider data on 656 prediction of Gonzalo's intensity. A twin experiment was performed comparing an analysis that 657 assimilated underwater glider data from July 15 to October 13 along with other in-situ and 658 satellite observations to an analysis produced by an unconstrained model simulation. These two 659 analyses were then used to initialize the high-resolution HWRF-HYCOM coupled forecast 660 system (Dong et al., 2017). Assimilation of subsurface observations from gliders improved the 661 representation of pre-storm vertical structure of both temperature (Figure 6b) and salinity. 662 capturing the barrier layer previously observed in the region (Domingues et al., 2016). 663 Consequently, forecast intensity errors (e.g., Figure 6c) were reduced by approximately 50% as a 664 result of assimilating all available observations, enabling a substantially improved forecast for

665 Hurricane Gonzalo.

666 OSEs are now being conducted for the 2017 North Atlantic hurricane season. The fields of 667 mean TCHP and D26 presented in Figure 12 demonstrate the impact of assimilating all ocean 668 profilers (Argo and Alamo floats plus underwater gliders). Comparing fields produced by an 669 unconstrained model simulation (Figures 12c and 12d) to observation-based estimates provided 670 by the NOAA/AOML TCHP analysis product (Figures 12a and 12b), the unconstrained model 671 produces TCHP that is too small and an upper-ocean warm layer that is too thin across the entire 672 North Atlantic hurricane development region. Assimilation of all available ocean profiles 673 (Figures 12e and 12f) substantially corrects these large-scale biases. The planned next step in this 674 analysis will be to assess the impact on intensity prediction by using these fields to initialize the 675 HYCOM-HWRF prediction model.

676 Observing System Simulation Experiments (OSSEs) have also been performed over the 677 North Atlantic hurricane region. Given that the Nature Run, a validated, unconstrained, and 678 realistic ocean simulation by a state-of-the-art ocean model, is known, it is possible to evaluate 679 new observing systems and alternate deployment strategies for existing systems. Previous OSSEs 680 have quantitatively assessed the positive impacts of existing observing systems and different 681 deployment strategies for systems, such as underwater gliders and picket-fence deployments of 682 thermistor chains (Halliwell et al., 2017a), and also for pre-storm airborne ocean profiler surveys 683 (Halliwell et al., 2017b). More recently, OSSEs were performed to demonstrate the advantages 684 of collecting ocean profiles from moving platforms such as gliders compared to collecting 685 profiles from stationary platforms. These results are summarized by Fujii et al., (2019). Moving 686 forward, OSSEs will continue to be an important tool for the design and implementation of 687 optimized ocean sampling strategies in support of both TCs and ECs forecasts, while OSEs will 688 also continue providing further quantitative information on the impacts of different components 689 of the existing ocean observing system.

690

691 **5. Data Management**

692

Efficient data management, including data transmission is critical for ensuring
observations are available in real-time or near-real-time for assimilation into forecast models.
Latency in data availability can have unwanted downstream effects on the use of observations for
operational purposes.

697 For weather forecasting, it is critical that Data Assembly Centers (DACs) and operators 698 transmit data in real time to systems such as the GTS to ensure data availability for forecasters 699 and to validate models. In order to make the data available for assimilation into forecast 700 numerical models, most of the data obtained by the different observational platforms considered 701 here need to be transmitted in real-time or near-real time through different satellite networks. 702 After reception on land, the data typically undergo platform-dependent quality assurance (QA) 703 and quality control (QC) procedures that are designed to identify possible inaccuracies in the 704 observations. For most platforms these QA/QC procedures include tests designed to identify data 705 gaps or missing values, spikes or unrealistic gradients in the data, and invalid dates or locations, 706 among other error sources. The data are normally not modified during QA/QC, but individual 707 records are flagged according to the results of the tests applied, or the data from a malfunctioning 708 platform may be blacklisted and removed altogether from GTS distribution. The data are then 709 encoded into different traditional alphanumeric formats (e.g., FM 63-XI Ext. BATHY for XBTs 710 and AXBTs, FM 64-XI Ext. TESAC for Argo floats and underwater gliders; World 711 Meteorological Organization, 2015a - Part A), or into binary universal form for the 712 representation of meteorological data (FM 94-XIV BUFR, World Meteorological Organization, 713 2015a - Part B). For example, the data format TM315009 is used by the Lagrangian Drifter 714 Laboratory at the Scripps Institution of Oceanography, by MeteoFrance, and by the UK Met

715 Office for their contribution to the Global Surface Drifter Array, and for submission into the

- 716 GTS (World Meteorological Organization, 2015b) for near real-time distribution and numerical
- 717 model assimilation. Other data centers includes the Global Temperature and Salinity Profile
- 718 Programme (GTSPP; XBT, Argo floats, underwater glider), the U.S. IOOS Glider Data
- 719 Assembly Center (GDAC), and NOAA/NCEI (XBT, Argo, underwater glider) as part of long
- term archival and for distribution for other delayed-mode scientific applications. At this step the
- data may be submitted to delayed-mode QC that may result in flags for individual records or in
- modifications to the data set to ensure the highest possible data quality for all applications.

For research and retrospective analysis, data management is important to ensuring collected observations from various platforms, operated by diverse organizations, is easily available, QA/QCed, and compatible with relevant standards. DACs can be leveraged to provide a diverse observation platform community a single place to store, share, archive, and quality control their data. In addition to providing standardized, easy to access, QA/QCed ocean observations critical for extreme weather events.

729

730 6. The vision for the next 10 years

731 6.1 Ocean observations in support of tropical cyclones studies and forecasts

732 An integrated multiplatform ocean observing system for studies and forecasts of TCs is 733 not currently in place. Analysis of ocean observations from the largely climate-focused ocean 734 observing system often provides valuable information on the mechanisms and processes 735 associated with these extreme weather conditions. Ocean data in support of extreme weather 736 events need to focus on resolving upper ocean features such as barrier layers, spatial variability 737 of warm currents, mesoscale ocean heat content changes, and surface waves (Centurioni et al., 738 2019) prior to and during the season in each basin where TC occur, with distribution of data in 739 real-time. However, the scientific and operational requirements of observing platforms, such as 740 profiling floats (Roemmich et al., 2019), moorings (Foltz et al., 2019; Masumoto et al. 2019; 741 Smith et al., 2019), and expendable probes (Goni et al, 2019), do not explicitly target these 742 needs. Sustained and targeted high-resolution ocean observations provide a means to better 743 understand the processes responsible for the rapid evolution of the ocean and its feedback on the 744 atmosphere during these extreme weather conditions. These concerns have been presented and 745 discussed in workshops on TCs from a global perspective, as for example where WMO 746 Recommendations focused on structure and intensity of TCs (Shay et al., 2014).

Pilot networks of sustained multi-platform observations and targeted observations in the tropical Atlantic during hurricane season have proven to provide key upper ocean observations to initialize numerical ocean-atmosphere coupled forecast models in areas where TC intensification and weakening may occur. The assimilation of ocean observations allows for a better representation of ocean conditions within coupled TC forecast models, which in turn provides a more realistic simulation of air-sea interactions and flux exchanges, generally resulting in an 753 improved TC intensity forecast (e.g. Chen et al., 2017; Dong et al., 2017). OSEs (e.g., Dong et 754 al. 2017) need to be extended to more storms in order to provide a more robust estimate of the 755 benefit of various types of observations. These experiments should ideally be performed using 756 operational models so as to quantify the benefit of ocean observations in operational conditions. 757 Furthermore, the OSSE approach dedicated to hurricanes should continue to be followed in order 758 to optimize the deployment of dedicated TC ocean observations, typically gliders and air-759 deployed profilers. Carrying out OSSEs and OSEs to design, implement, and assess the impact 760 of new sustained components within the ocean observing system (e.g., underwater gliders, 761 profiling floats, drifters, etc.) will be key to continued improvement of TC intensity forecasts, since significant errors still remain in data-assimilative ocean analyses due to existing 762 763 observations being scarce in space and time. Improvements in spatial and temporal coverage of 764 ocean observations should improve the ocean representation within coupled TC forecast models, 765 which in turn will allow for better forecasts. Targeted ocean sampling, when appropriate, also 766 has the potential to help improve TC predictions (e.g., Chen et al., 2017). In addition, 767 improvements in data availability for the forecast community are also essential for ensuring that 768 ocean observations reach operational forecast centers in real-time. In one effort to help with this 769 requirement, the EMC and the National Data Buoy Center (NDBC) within NOAA are working 770 to increase the frequency of data transfer to the GTS.

771 For the next decade, coupled model systems will extend to multi-way dynamic coupling. 772 In recent years, NOAA/EMC has demonstrated 3-way dynamic coupling with HYCOM-773 WaveWatchIII-HWRF model. This would allow for revisiting air-sea interaction dynamics in 774 greater detail, and also exploring observational measurements to support research and 775 simulations. The importance of air-sea flux exchanges to TC development is widely known, yet 776 simulations are still based on bulk parameterizations. To support evolving modeling efforts, 777 observational efforts should accordingly extend to collecting data on waves, sea spray, 778 roughness, turbulence, and relative humidity over the ocean. For example, measurements derived 779 from turbulent microstructure sensors, such as those based on underwater gliders (e.g., St. 780 Laurent and Merrifield, 2017) and moorings (e.g., Warner et al., 2016) will help obtain direct 781 measurements of diapycnal heat flux and temperature diffusivity that can be used to develop, 782 assess, and validate turbulent mixing schemes employed in coupled forecast models.

783 Expansion of sustained and targeted upper ocean observations in locations where TCs 784 often intensify is one of the best strategies to support hurricane studies and forecasts. Underwater 785 gliders and other autonomous vehicles offer one option for carrying out sustained surveillance in 786 support of TCs studies and forecasts, given that these vehicles can be remotely operated along 787 predetermined routes, they can provide observations in real-time continuously for several 788 months, they withstand hurricane-force winds, and they can be refurbished and serviced for multi-year applications. Targeted and rapid response observations also provide critical 789 790 information that instruments surveying in sustained mode cannot. For instance, air-deployed 791 instruments are particularly useful since they are deployed from aircraft already tasked with

- storm surveillance, they are logistically easier to position along the forecast track ahead of a TC.
- 793 Flexible deployments of in-situ marine and airborne platforms allows for co-located
- 794 measurements with other air/ocean observing systems that are key for advancing our
- understanding air-sea fluxes across the oceanic surface. These can also help provide precious
- data points for future air-sea coupled data assimilation methods under active consideration for
- balanced initialization of next generation coupled hurricane models.

798 Further advances in satellite remote sensing are expected to improve the representation of 799 features that impact storm development. For instance, the advent of wide-swath, high-resolution 800 altimetry (e.g., Fu and Ubelmann, 2014) will enable the evaluation of air-sea interaction 801 processes during high-wind events in detail, such as, for example, the generation of internal 802 waves in the wake of TCs. Satellite measurements of surface salinity have potential for 803 improving our understanding of the oceanic factors and processes that lead to TC intensification, 804 especially in the western Atlantic and the Bay of Bengal, where there is persistent shallow salinity stratification. It is important that these measurements continue, along with satellite SST, 805 806 sea level, and winds.

807 Considering the positive impacts of upper-ocean observations from pilot networks, and 808 targeted deployments, the following key recommendations have been identified to continue and 809 enhance ocean observations in support of TCs:

- Maintain the elements of the observing system that have proven valuable for Tropical
 Cyclone ocean research and operational intensity forecast.
- Utilize numerical Observing System Experiments to quantify the impact of the current
 ocean observing platforms in Tropical Cyclone forecasts.
- Evaluate optimal observational strategies in support of Tropical Cyclone studies and
 forecasts using numerical Observing System Simulation Experiments.
- Implement sustained and targeted pilot observations (gliders, profiling floats, drifters, etc.)
 dedicated to improving Tropical Cyclone intensity forecasts; and foster co-incident, co located air-deployed profile observations (AXBTs, AXCTDs, floats, thermistor chains, etc.)
 of ocean temperature, salinity, and currents.
- Foster additional sustained measurements of sea level pressure (e.g., from drifters and moorings), and of waves, sea spray, and mixed-layer turbulence (e.g., from gliders) to help
- 822 develop, evaluate, and validate boundary layer parameterizations.
- Use upper ocean metrics (e.g., Tropical Cyclone Heat Potential, ocean mean temperature,
 barrier layer thickness, etc.) derived from profile and satellite ocean observations in the
 operational evaluation and validation of numerical forecast models.

- Continue with efforts focused on improving coupled ocean-atmospheric numerical weather
- models, especially those relating to enhancing ocean data assimilation techniques and mixedlayer parametrizations.
- Create an ocean database easily accessible to the scientific community to facilitate research
 in support of assessments of the role of the ocean in Tropical Cyclones studies.
- Enhance data management efforts to transmit and QA/QC data in real-time for assimilation
 in operational forecast models.
- 833

834 6.2 Ocean observations in support of extratropical bomb cyclones studies

- Recommendations to improve the understanding of ocean-atmosphere interactions during ECevents are:
- Increase efforts to implement and improve coverage of high-frequency and high-resolution
 observations using profiling instruments and Biologging to detect oceanic fronts associated
 with western boundary currents in winter.
- Enhance efforts dedicated to observing surface wind, and waves, using surface drifters, and
 floating seabirds equipped with weather, GNSS, and motion sensors, respectively, to
 estimate air-sea flux exchanges under Extratropical Bomb Cyclones.
- Foster additional efforts aimed at observing ocean turbulent mixing induced by
 Extratropical Bomb Cyclones using, profiling floats and other platforms (e.g. gliders,
 moorings, etc).
- Incorporate real-time meteo-ocean observations, including ocean bottom pressure, in
 moorings from the Tsunami monitoring network in support of Extratropical Bomb Cyclones
 studies and forecasts.
- 849

850 Air-sea interactions under ECs are poorly understood because of the sparseness of in-situ 851 observations and lack of satellite observations caused by thick clouds and heavy rain. Seabirds 852 are often observed to fly and float under ECs to forage (Yoda et al., 2014; Yonehara et al., 2016; 853 Goto et al., 2017), providing an additional potential source of environmental data. Estimation of 854 surface winds and waves using Biologging GNSS and motion sensor can provide useful 855 information about air-sea interaction processes under ECs as well as their temperature and 856 pressure measurements. In addition, the development of profiling floats equipped with motion 857 sensors can also help to provide metrics to evaluate ocean mixing near the sea surface.

- To monitor Tsunami, several real-time observation networks of ocean bottom pressure
- have been established (Kaneda, 2010; Bernard and Meinig, 2011; Lawson et al., 2011;
- 860 Mochizuki et al., 2017). Most of the sites in the networks are located under the area where ECs

861 frequently develop. Recently, seismic stations on land can catch microseisms induced by ECs

862 (Nishida and Takagi, 2016). The real-time monitoring networks will provide oceanic responses

to ECs and informations of winds and waves which may contribute to forecast improvement ofECs.

865

866 **7. Summary**

In this community white paper, we provide a summary of current ocean observing efforts, and recent research findings in support of studies and forecasts of TCs and ECs. Substantial progress has been made over the past decade in ocean observations, improving our understanding of the role that the ocean plays in the evolution of TCs and ECs, and transitioning state-of-the art coupled forecast models to operational mode. These advances have largely addressed recommendations made by the scientific community for OceanObs'09 (e.g., Goni et al., 2010) and emphasize the critical value of sustained and targeted ocean observations, real-time data

transmission, and multi-platform efforts.

With recent advances in ocean modelling and coupled atmosphere-ocean modelling,
operational forecasts increasingly rely on assimilating real-time ocean measurements to produce
accurate ocean, weather and extreme weather forecasts. For example, assimilation of ocean
observations can dramatically improve hurricane intensity forecasts (e.g., Dong et al., 2017).
OSEs can assist in quantifying the impact of upper ocean observations on TC and ECs forecasts.
Similarly, OSSEs can also be applied to various regions to design optimal and cost-effective
deployment strategies for both targeted and sustained observations in support of ECs and TCs.

882 Given the large benefits provided by the ocean observing system in support of extreme 883 weather studies and forecasts, it is critical that current components are maintained and possibly expanded over the next decade. In addition, new technologies, pilot networks, and targeted 884 deployments are greatly expanding the observation capabilities; incorporating these components 885 into the sustained observing system will likely greatly benefit studies and forecasts of TCs and 886 887 ECs. Finally, considering the large number of countries whose coastal areas are often impacted by TCs and ECs, results and advances presented here emphasize the critical value of carrying out 888 889 a coordinated international effort in the design, implementation, maintenance, and data 890 management of key aspects of ocean observations that will ensure the feasibility of logistical, 891 operational, and research activities.

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893 Author Contributions Statement

894

RD, AKY, PCM, and GG elaborated the initial outline, and wrote the first draft of the

896 manuscript. All 29 authors of this community white paper have provided input in the form of

- 897 language, figures, and recommendations for the future of the ocean observing system in support
- 898 of studies and forecasts of Tropical and Extratropical Cyclones.
- 899

900 Conflict of Interest Statement

901

Author Hyun-Sook Kim was employed by company I.M. Systems Inc. All other authors declare
 no competing interests

- 905 **References**
- 906

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Inteview

	Figure Contions
	Figure Captions
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⁹ source: https://www.nhc.noaa.gov/verification/verify5.shtml

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- 1436 during Nate's passage of Nate over the Gulf. (a) Pre-storm OHC structure on 5 October 2017. (b)
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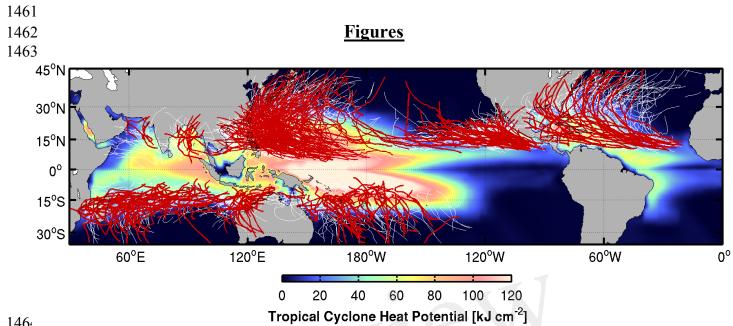
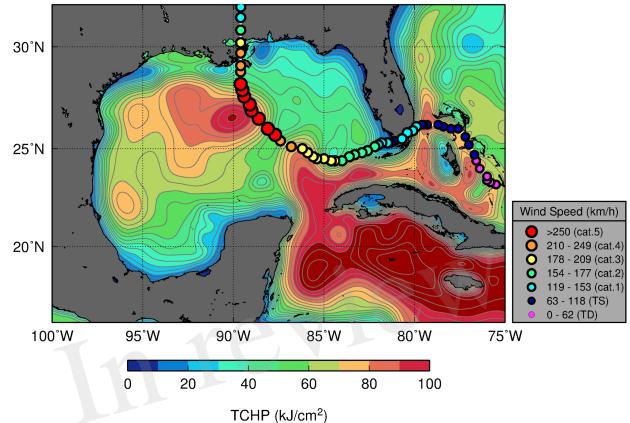


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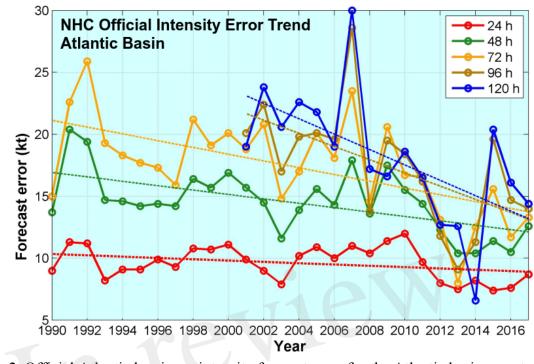


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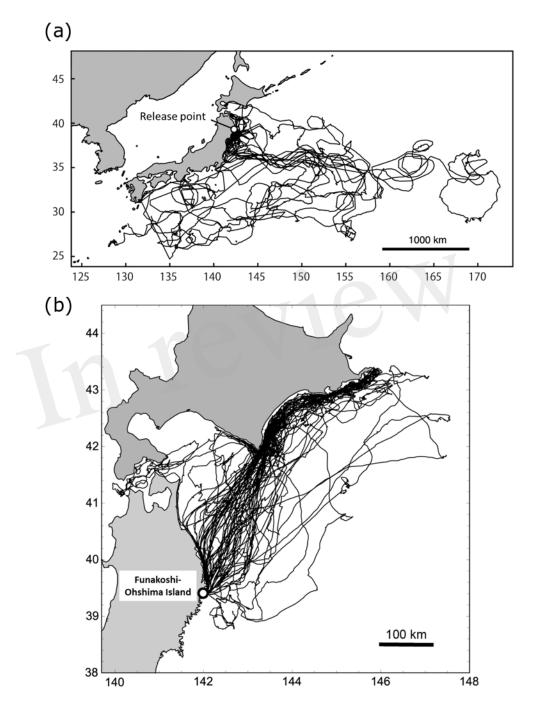
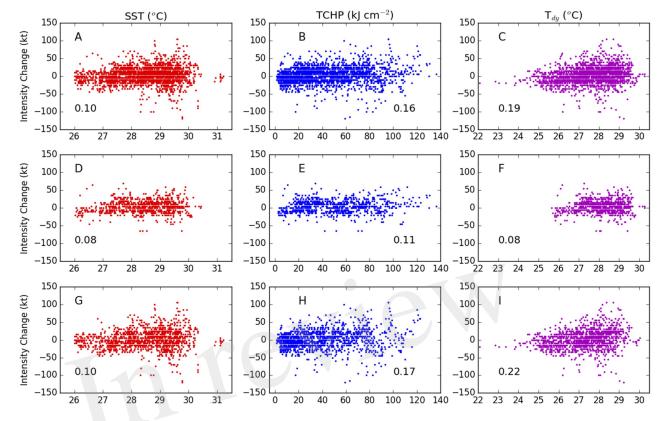


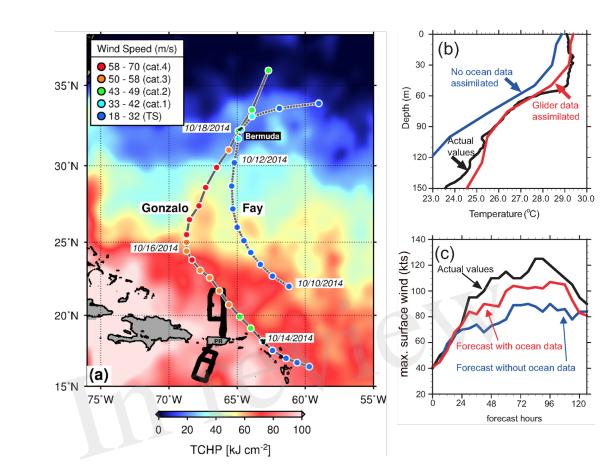
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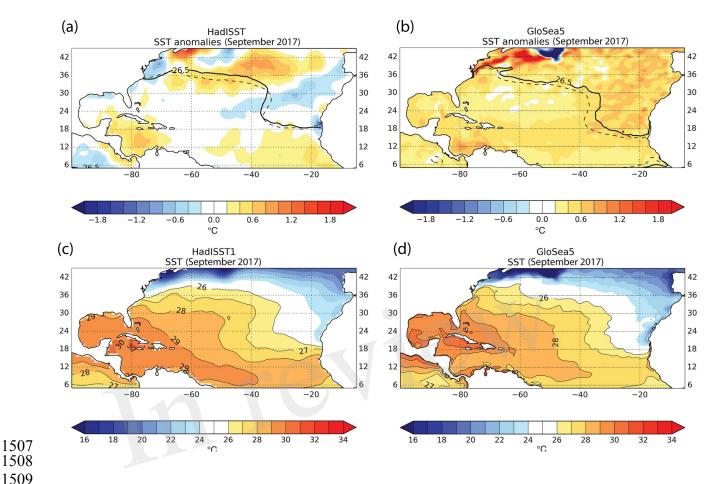
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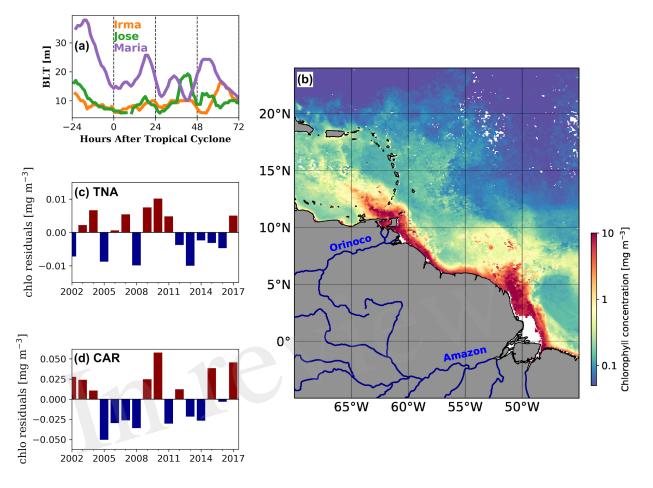
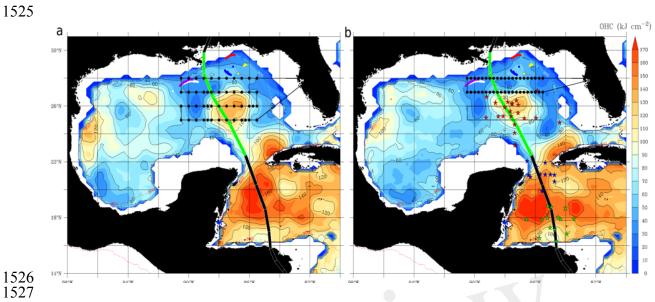


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1536 40 T100 - Average Temperature in top 100m easnfs 9/23/2010 30 27 28 29 ITOP 2010 CONTROL CENTER apan MONTEREY, CA Intensive operations period · August - October 2010 30 Satellites, Models, Forecasts Latitude (^oN) ٠ aiwan T. Malakas T. Fanapi Aircraft Ships DotStar Air-Deployed C130 Floats and Drifters Megi 20 CTD Moorings -Gliders Dropsondes AXBT SFMR Guam Philippine 10 130 Longitude (°E) 140 110 120 150

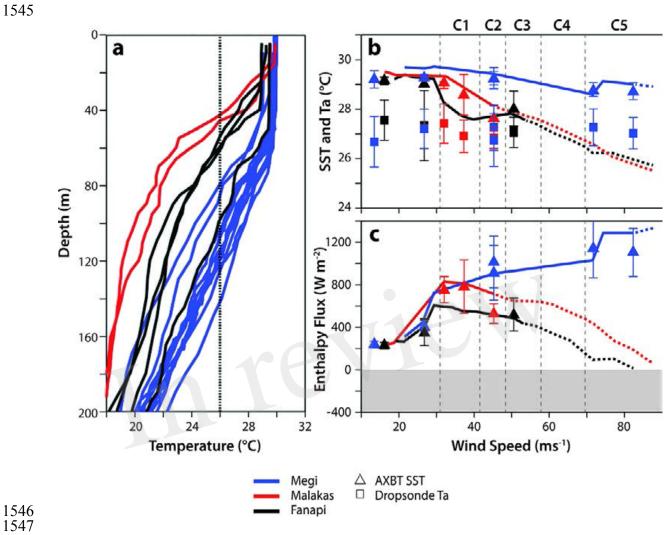


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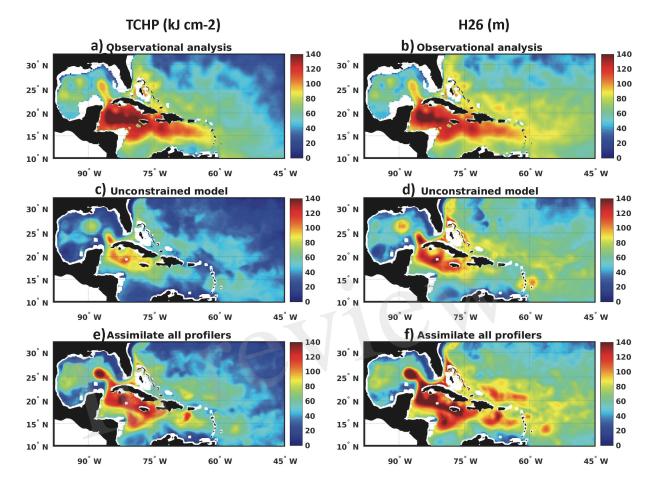




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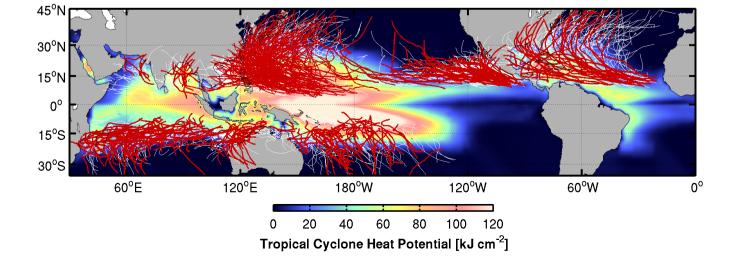
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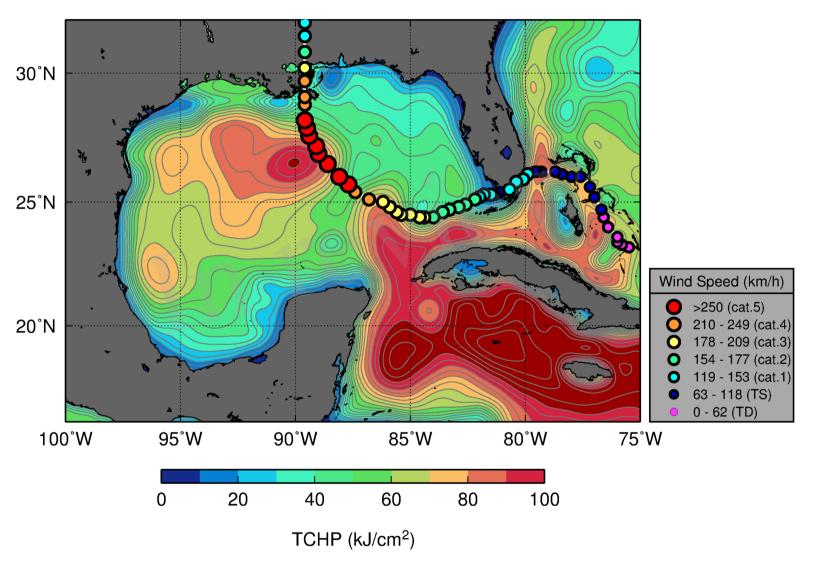
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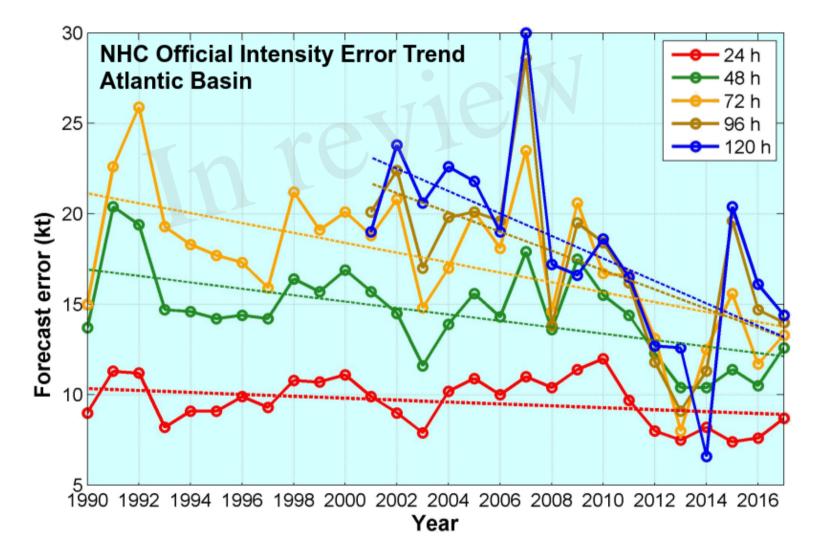


Figure 3.JPEG

