

1 **Supplemental material for “Global impacts of**
2 **subseasonal wind variability on ocean surface stress,**
3 **buoyancy flux, and mixed layer depth”**

D. B. Whitt¹, S. A. Nicholson², and M. M. Carranza¹

1. Contents of this file

4 Figures S1 to S8.

2. Validating the CTL simulation

5 This supplementary text reports on a validation analysis that compares the model
6 output against in situ observations.

7 Several previous studies have validated mesoscale-resolving CESM model solu-
8 tions/configurations that are similar to the CTL experiment by comparing the model

Corresponding author: D. B. Whitt, National Center for Atmospheric Research, PO Box 3000,
Boulder, CO 80307 (dwhitt@ucar.edu)

¹National Center for Atmospheric
Research, Boulder, CO, USA.

²Southern Ocean Carbon-Climate
Observatory (SOCCO), CSIR, Cape Town,
South Africa

9 results to independent observations [*Bryan and Bachman, 2015; Johnson et al., 2016;*
10 *Harrison et al., 2018; DuVivier et al., 2018; Delman et al., 2018*]. Here, we discuss a few
11 new comparisons that demonstrate the model’s ability to simulate the observed climato-
12 logical MLD in the CTL experiment (as well as its deficiencies in this regard). Unlike
13 previous validation efforts, we compare against observations using *in situ* profile data and
14 the same MLD definition for the observations and model data (i.e., as computed in the
15 model, and described in *Large et al. [1997]*, page 2427). In-situ profile data of temperature,
16 and salinity are converted to conservative temperature and absolute salinity, from which
17 density is derived using the international thermodynamic equation of seawater (TEOS-
18 10, [*McDougall and Barker, 2011*]). Quality-controlled instantaneous profile data from
19 the Coriolis Ocean Dataset for Reanalysis (CORA v5.1, [*Cabanes et al., 2013*], available
20 at <http://marine.copernicus.eu/services-portfolio/access-to-products/>) collected between
21 January 2000 and December 2017 are used for the analyses. The MLD is calculated for
22 each profile, after the density is linearly interpolated to a uniform grid with 5 m reso-
23 lution and then smoothed with a 15 point (75 m) moving average. Some smoothing of
24 the observed density profiles is necessary in order to make meaningful global comparisons
25 between the model and observations. In particular, the observed MLD is highly sensitive
26 to the smoothing in crucial regions where the MLD is deep and the stratification is weak
27 (e.g., the Labrador Sea during winter). However, the MLD is not particularly sensitive
28 to either the definition or the smoothing procedure over most of the ocean most of the
29 time. For example, Figure S7 compares the annual mean MLD calculated with this *Large*
30 *et al. [1997]* method and the algorithm proposed by *Holte et al. [2017]* as well as the MLD
31 defined by a 0.03 kg/m^3 density threshold [*de Boyer Montégut et al., 2004*]. The definition

32 used here produces a climatological mean MLD that is mostly (although not everywhere)
33 somewhat deeper than these other two definitions.

34 Monthly climatologies of MLD, sea surface temperature (SST) and sea surface salinity
35 (SSS) for the top 10 m are constructed by bin-averaging profile data in 2° bins for each
36 month, and model output was regridded to the same regular grid for comparisons against
37 observations. Significance for the differences between model and observations was evalu-
38 ated by means of a bootstrap method over 100 realizations, and a confidence level of 95%,
39 assuming all observed profiles are independent.

40 Biases in climatological mean surface temperature and salinity relative to observations
41 are almost everywhere less than 3°C and 1 g kg^{-1} but mostly smaller (Figure S2). There
42 are notable cold surface temperature biases in the model in the Irminger and Norwegian
43 Seas, as well as in the vicinity of the northwest corner of the North Atlantic Current (east
44 of Newfoundland). Notable warm surface temperature biases are observed on the eastern
45 boundaries of North and South Pacific as well the western margin of the subtropical
46 North Atlantic (north of Cape Hatteras) and subtropical North Pacific, including much of
47 seaward extension of the Kuroshio, and the Southern Ocean. The modeled global ocean
48 surface salinity is mostly fresher than the climatology from observations, particularly at
49 mid-to-high latitudes [similar to a previous low-resolution CESM, see *Griffies et al.*, 2009].
50 However, there is a notable fresh bias in the vicinity of the northwest corner of the North
51 Atlantic Current. Notable salty biases are observed in the western tropical North Atlantic
52 and along a thin strip just north of the equator that extends seaward from the eastern
53 boundaries of the Atlantic and Pacific. On a global scale, the magnitudes of these biases
54 are generally comparable to or smaller than the magnitude of the biases observed in 500-

55 year simulations on a grid with nominal 1° resolution with the same CORE-1 forcing
56 protocol [Griffies *et al.*, 2009].

57 Model MLDs are also qualitatively similar to observed MLDs, although there are im-
58 portant differences (Figure S3). For example, simulated mean mixed layers are generally
59 deepest in the observed sites of deep mixing and winter watermass formation, including
60 a narrow deep mixing band in the Sub-Antarctic zone of the Southern Ocean, the subpo-
61 lar North Atlantic including the Labrador Sea, above the Greenland-Iceland-Faroe Ridge
62 complex, and parts of the Norwegian and Barents Seas. Biases are small during late sum-
63 mer, but modeled MLDs are almost universally slightly shallower than observed during
64 that season. Notable biases include annual mean MLDs that are too deep in the modeled
65 central to eastern tropical Pacific compared to observations (Figure S3). Conversely, hot
66 spots of deep winter MLDs are almost universally too shallow in the simulations, including
67 in the subpolar North Atlantic during March. In addition, the simulated MLDs are too
68 shallow in most of the observed deep winter mixing band in the Subantarctic Southern
69 Ocean during September. Somewhat surprisingly, these results contrast with those of
70 *DuVivier et al.* [2018], as there is no clear indication that modeled MLDs are too deep
71 in any significant part of this Subantarctic deep mixing band (compare with their Figure
72 10). This contrast highlights the rather strong and qualitative sensitivity of MLD biases
73 to the chosen definition of the MLD and its precise construction in crucial regions of deep
74 MLD and watermass formation.

75 It may be noted that there is no data to compare to in the Arctic, which is a disadvan-
76 tage of using an Argo-based data set; but, the generally shallow mean of 32 m in CTL
77 and 48 m mean difference from shallowest to deepest depth in the monthly climatology

78 is consistent with other observations [*Peralta-Ferriz and Woodgate, 2015*]. Further, the
79 spatial distribution of the MLD is qualitatively consistent with observations in the Arc-
80 tic. Deeper mixed layers and largest seasonal cycle amplitudes are in the Barents Sea,
81 Marakov Basin, and Eurasian Basin and shallower mixed layers and smaller seasonal cycle
82 amplitudes are in the Canada Basin and the Chuckchi Sea (not shown).

83 Some of the differences between the model and observations are probably attributable
84 to the temporal inconsistency between the atmospheric forcing data sets (1958-2000) and
85 the MLD observations (after 2000), but presumably not all.

3. Supplemental Figures

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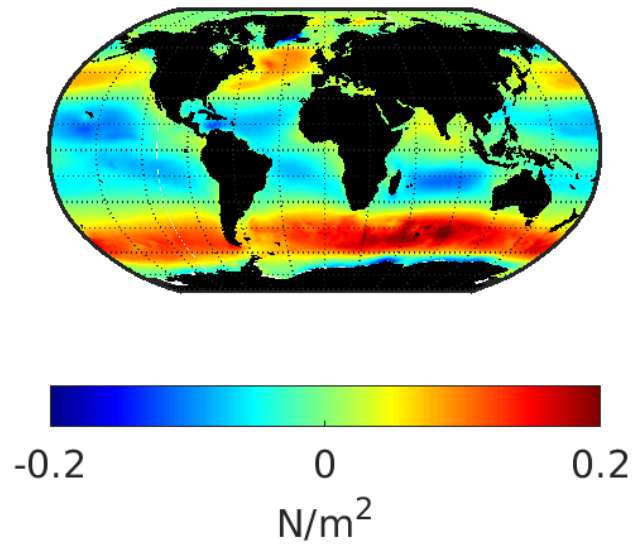
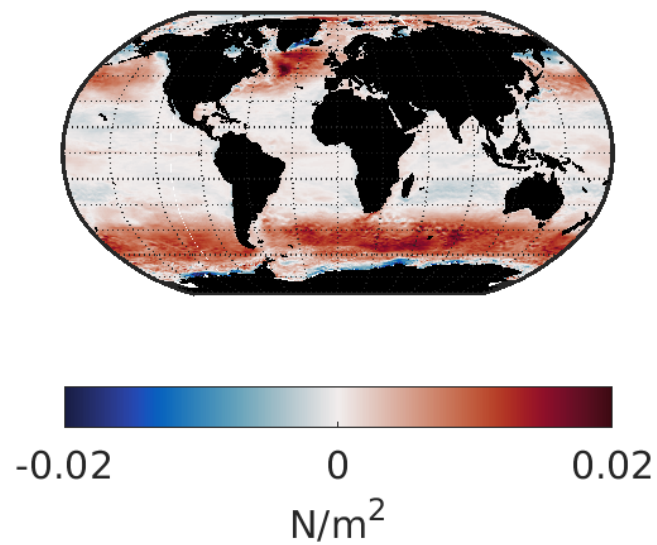
(A)zonal stress CTL**(B)zonal stress CTL-LP**

Figure S1. (a) Mean zonal stress in the CTL simulation, and (b) the difference between CTL and LP simulations.

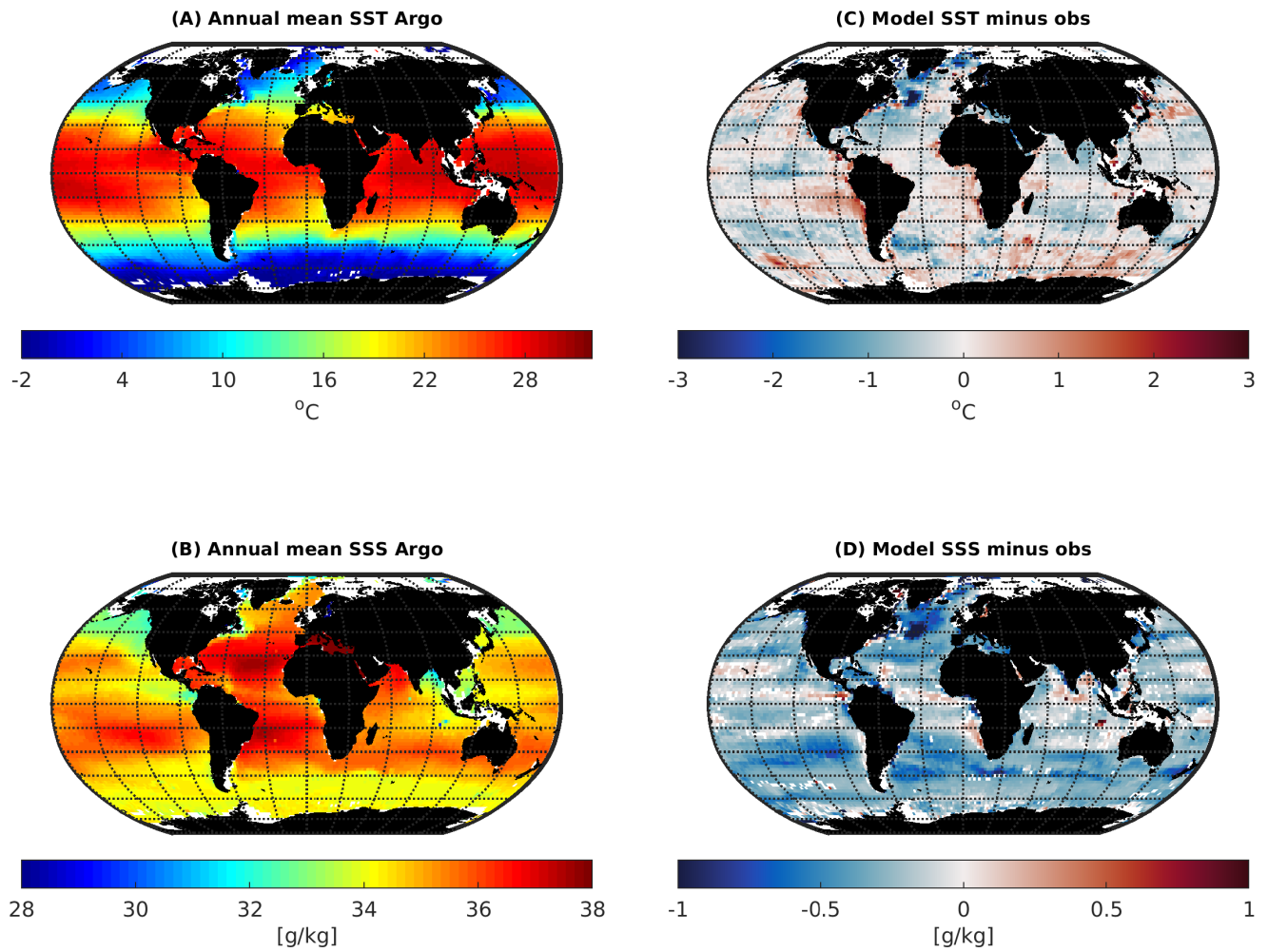


Figure S2. A comparison between the global mean temperature and salinity averaged over the top 10 m from Argo profiles computed between 2000 and 2017 and the simulated five-year time-mean temperature and salinity in the top model grid cell (0-10 m) of the CTL simulation. These results may be compared with Figures 7-8 in *Griffies et al.* [2009], where surface temperature and salinity are compared in several lower-resolution models forced by the CORE-1 normal year.

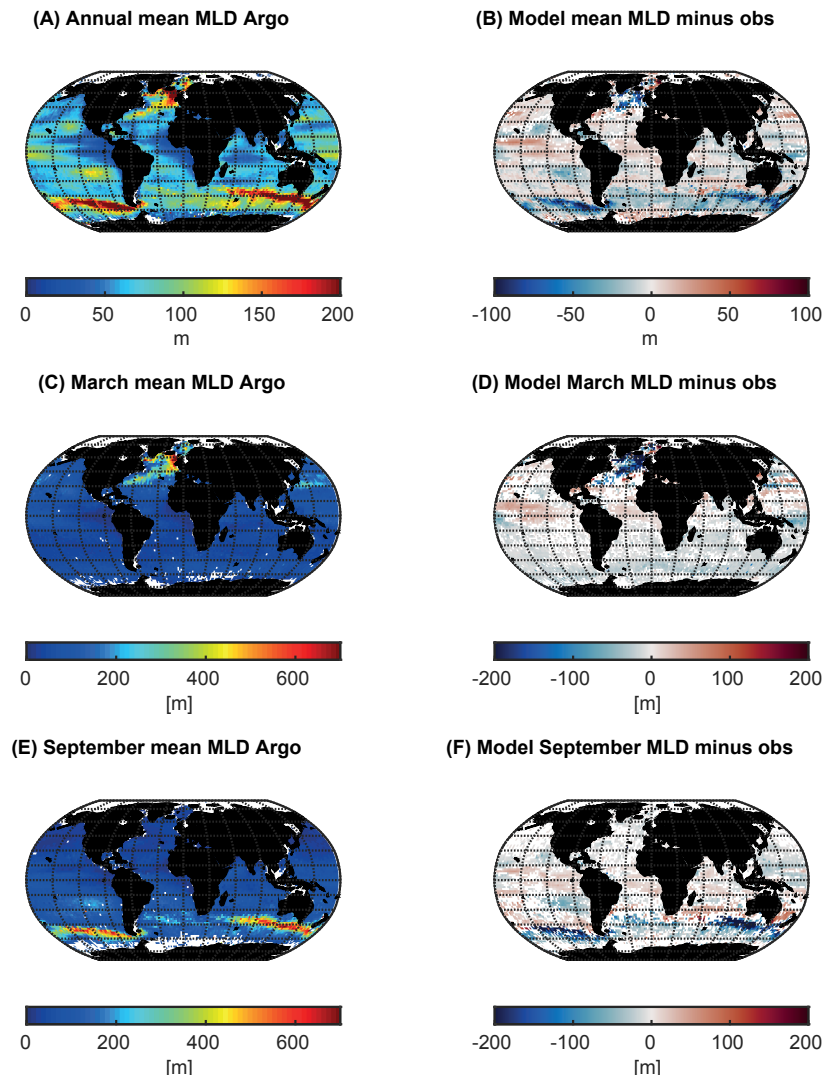


Figure S3. A comparison between the observed (Argo, 2000-2017) and simulated climatological MLD, both of which are calculated using the *Large et al.* [1997] algorithm. (Top) Annual means, (middle) March means, and (bottom) September means.

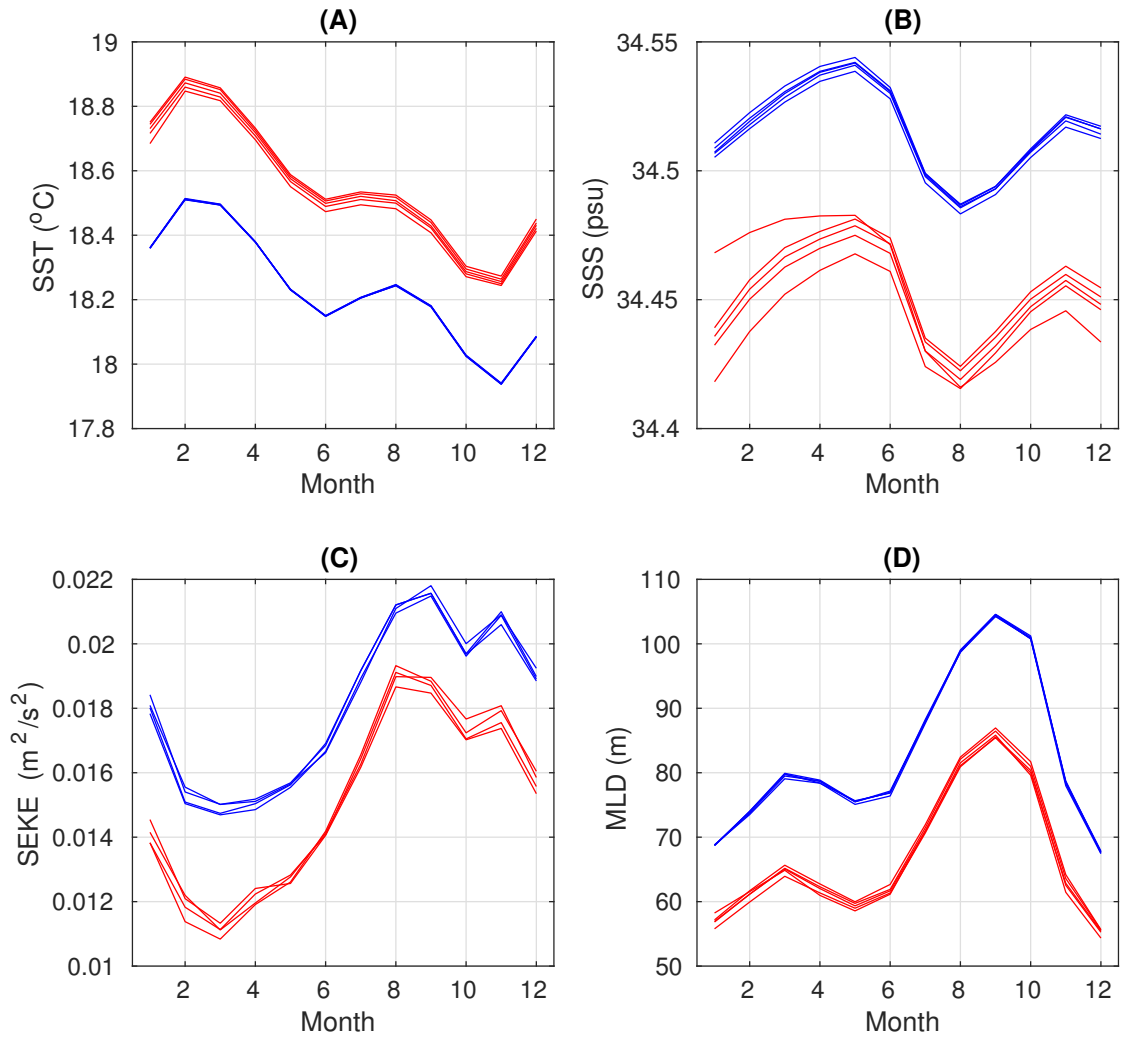


Figure S4. Time series of global mean surface temperature (A), salinity (B), a measure of surface eddy kinetic energy $SEKE = 1/2(u^2 - \langle u \rangle_t^2 + v^2 - \langle v \rangle_t^2)$ (C), and mixed layer depth (D) in the CTL (blue) and LP (red) simulations. Each line represents one of the sequential five years of each simulation. The fact that the lines are tightly clustered in each simulation demonstrates that the drift on inter-annual timescales in LP and CTL is relatively small compared to the amplitude of the seasonal cycle and the magnitude of the differences between LP and CTL. Note that there are only 4 lines for kinetic energy and in that case the time average $\langle \rangle_t$ only includes years 18-20 and 22, because the kinetic energy was not output to the monthly average files during year 21.

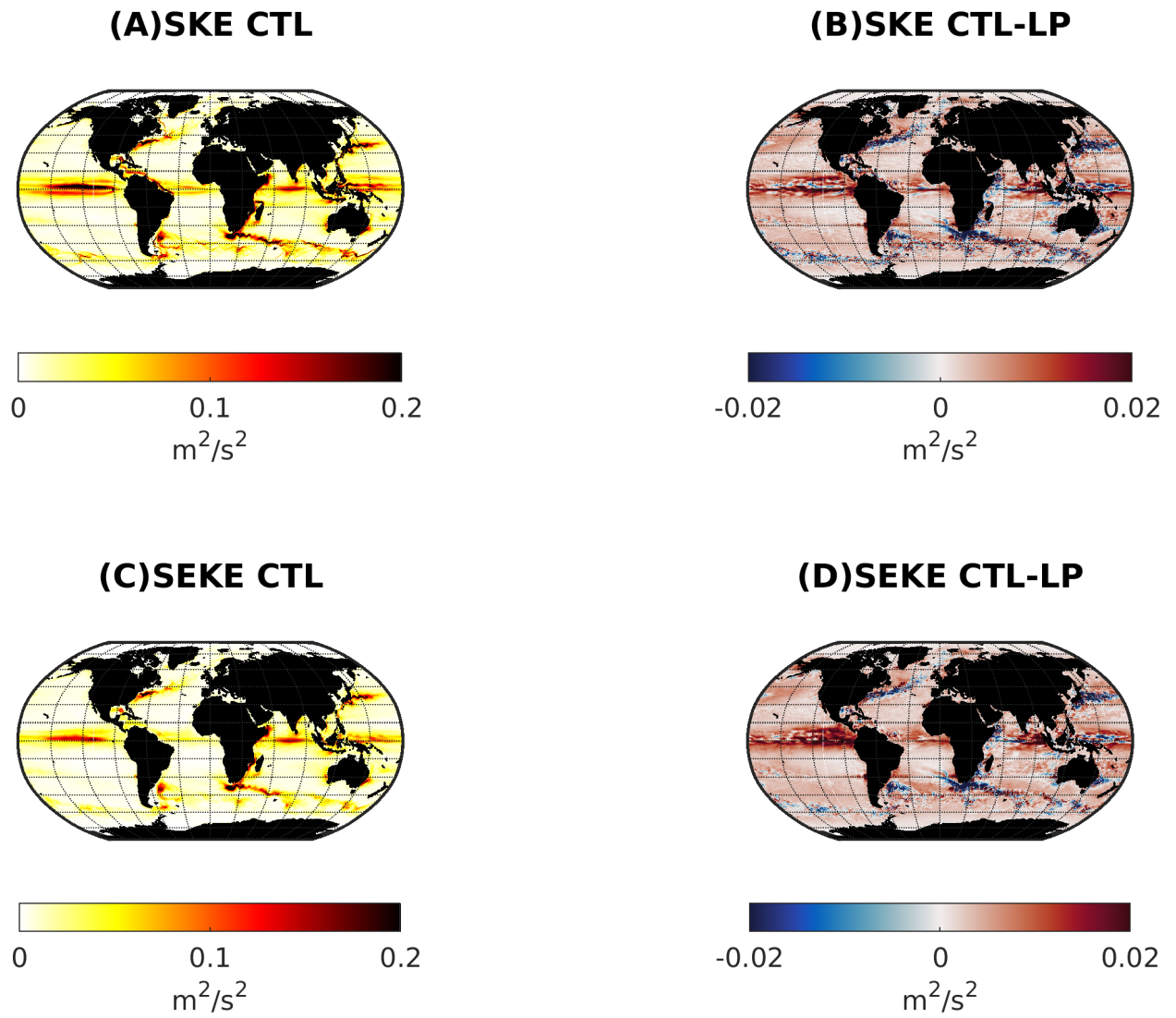


Figure S5. (a) Mean surface kinetic energy $SKE = \langle 1/2(u^2 + v^2) \rangle_t$ and (c) surface eddy kinetic energy $SEKE = SKE - 1/2(\langle u \rangle_t^2 + \langle v \rangle_t^2)$ and (b)-(d) the respective differences between CTL and LP simulations. Note that the time average $\langle \rangle_t$ only includes years 18-20 and 22, because the kinetic energy was not output to the monthly average files during year 21.

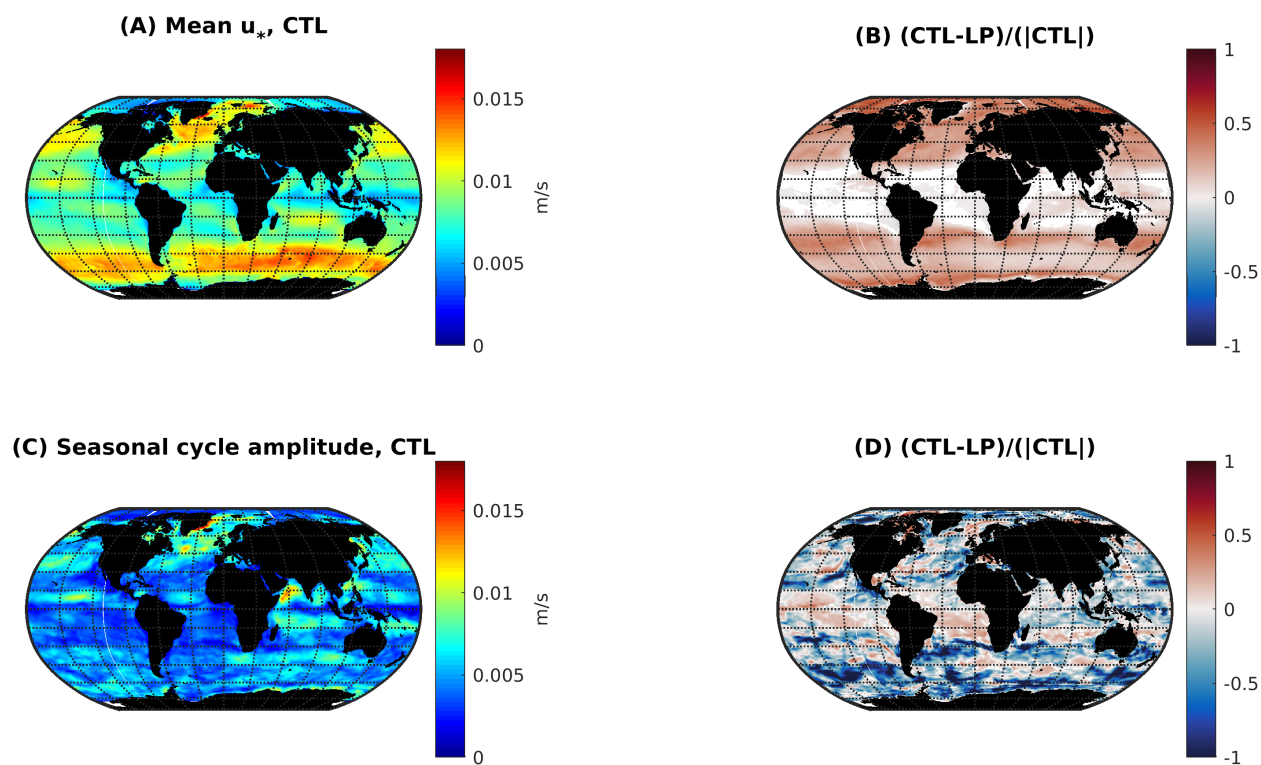


Figure S6. As in Figure 1 of the main manuscript, but friction velocity u_* .

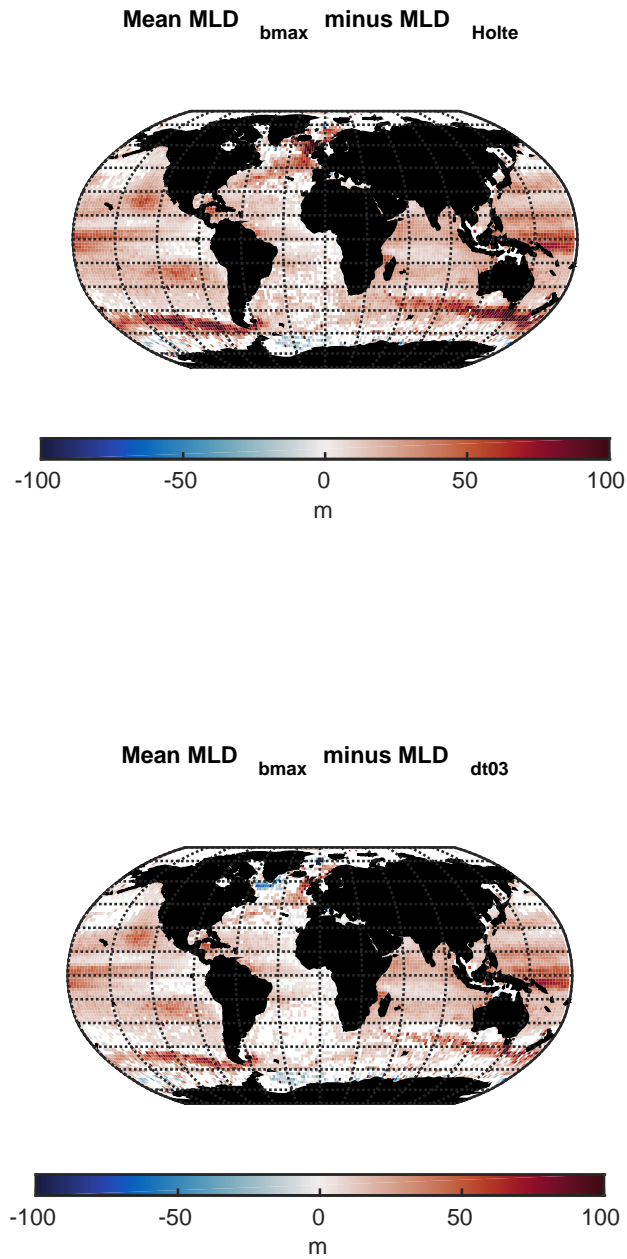


Figure S7. Differences between observed annual mean Argo MLD climatologies for different MLD definitions. Top: MLD “bmax” (as defined here and modified from *Large et al.* [1997]) and MLD “holte” [as defined in *Holte et al.*, 2017]. Bottom: MLD “bmax” minus the MLD defined by a 0.03 kg/m^3 density threshold [*de Boyer Montégut et al.*, 2004] (bottom).

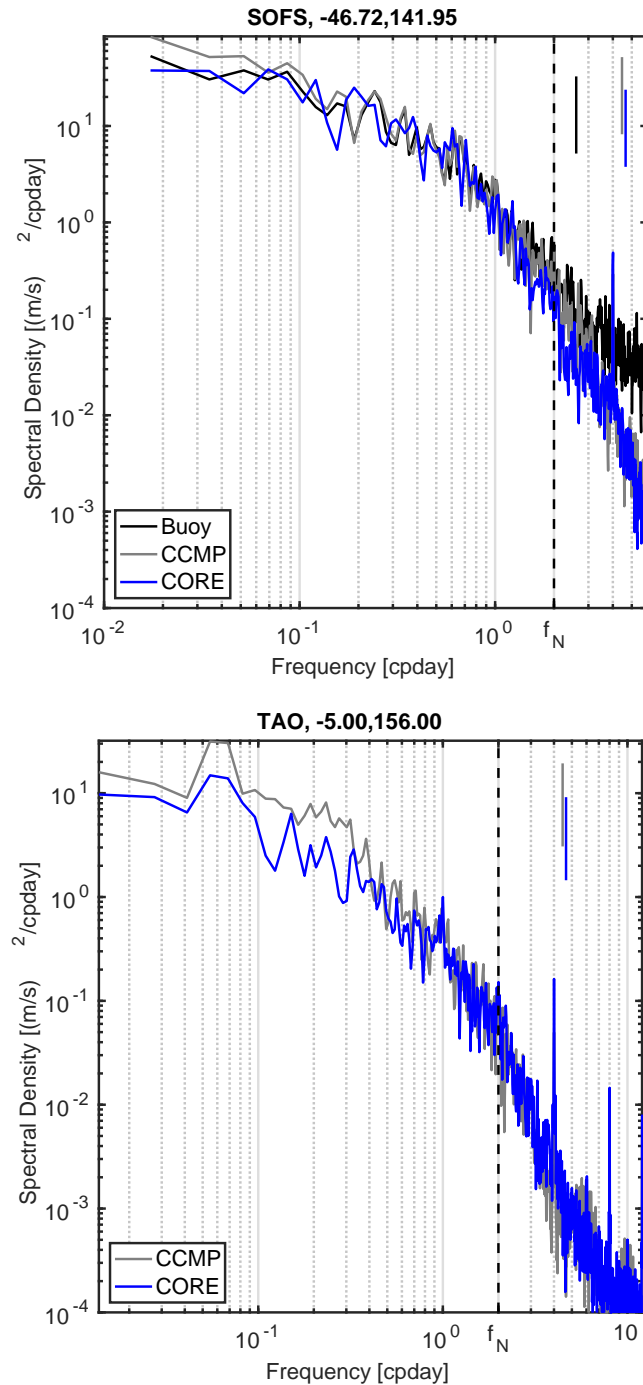


Figure S8. Some example comparisons between speed power spectra derived from the CTL winds, the CCMP satellite winds, and buoy measurements. The power spectra should not be identical because the observations are from particular years, whereas the CTL has climatological power at each frequency. The location (latitude °N, longitude °E) is as indicated in the titles. The top comparison is from the year 2010, the bottom is from the year 1995.