

1 **CheapAML: A simple, atmospheric boundary layer model for use**
2 **in ocean-only model calculations**

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ABSTRACT

4
5 We develop a model of the marine atmospheric boundary layer for ocean-only modeling
6 in order to better represent air-sea exchanges. This model computes the evolution of the
7 atmospheric boundary layer temperature and humidity using a prescribed wind field. These
8 quantities react to the underlying ocean through turbulent and radiative fluxes. With two
9 examples, we illustrate that this formulation is accurate for regional and global modeling
10 purposes and that turbulent fluxes are well reproduced in test cases when compared to
11 reanalysis products. The model builds upon and is an extension of Seager et al. (1995).

1. Introduction

The ocean surface exerts strong control on the atmospheric boundary layer through momentum, heat and moisture exchange across the ocean-atmosphere interface. Realistic ocean modeling places a premium on this influence, but common practices can omit it. For example, specifying fluxes at the ocean surface misses this connection; thereby, for example, locating strong oceanic heat loss away from the subsurface structure they should be associated with. Even high resolution, high accuracy scatterometer wind stress data can be uncorrelated with the surface velocity expression of a free running ocean simulation. Another common approach consists of specifying atmospheric temperature, humidity and wind and then diagnosing from them air-sea fluxes with a bulk equation. However, the ocean naturally develops scales much finer than the resolutions currently available from all atmospheric reanalysis products, and the imprint of these scales on the exchanges, with any subsequent feedbacks on the atmospheric variables, are lost. Given the low heat capacity of the atmosphere, the reaction of the atmosphere can be strong, thus influencing later heat exchanges and precipitation.

To fully capture the ocean-atmosphere connection at the interface would require a fully coupled ocean-atmosphere model. It is an open question if such a thing currently exists, however, even if it did, the computational burden associated with its use would be restrictive. It is thus of practical value to have alternatives that replicate at least some of the important coupling features.

Seager et al. (1995; S95 hereafter) in an insightful paper, recommended a partial solution to this problem. They proposed the use of a thermodynamically active, but dynamically passive, atmospheric boundary layer. The wind was specified, relieving the need to compute atmospheric dynamics, and with an assumption of rapid equilibration of the atmospheric temperature and humidity, the atmospheric state was diagnosed. Applications in models showed clear improvement in the flux structure relative to other products. Such an approach provides a means of overcoming many of the leading order omissions in ocean

39 modeling associated with either flux specification, the use of bulk formulae or relaxation
40 boundary conditions while simultaneously retaining the computational efficiency and flexi-
41 bility inherent in ocean only modeling.

42 Our goal here is to modernize the S95 thermodynamic atmospheric boundary layer model.
43 The four primary distinctions between our approach and that of S95 are in the (1) use of
44 modern flux algorithms, (2) abandonment of the equilibrium assumption, (3) calculation of
45 an accurate fresh water flux and (4) ease of migration to parallel computing. Specifically,
46 the modular model design permits the user to either develop a subroutine using the flux
47 algorithms of their choice, or to choose from the methods of Large and Pond (1982); hereafter
48 LP82 or COARE3 (Fairall et al. 2003). We discard the equilibrium assumption because
49 modern computing involves specifying atmospheric data typically at rates of a few times per
50 day, as opposed to the use of lower frequency climatological data as was the custom at the
51 time of S95's original paper. Coupled with this is the prediction, rather than diagnosis, of
52 the atmospheric state, which removes the need to solve an elliptic equation. The latter, aside
53 from suppressing 'weather' responses, are difficult to efficiently migrate to parallel platforms.
54 The impact of the equilibrium hypothesis has been studied by Hazeleger et al. (2001). They
55 added a parameterization of atmospheric storms to S95 and found significant impacts in
56 several regions of the Pacific. The use of daily wind (as opposed to a monthly climatology)
57 greatly modifies latent and sensible heat fluxes as well. Indeed the sensible heat fluxes scales
58 as $\propto u\Delta T$, with u the magnitude of the surface wind and ΔT the temperature difference
59 between the ocean and the atmosphere. When suppressing the variability of u and ΔT
60 by taking a mean value, it appears that the fluxes will be systematically underestimated,
61 especially when the variability of u is large (not shown).

62 Our model, named CheapAML, has been implemented in the MIT General Circulation
63 Model (MITgcm) and is available with the standard distribution as a package. This fortran
64 code computes the forced atmospheric temperature and relative humidity tendency terms;
65 it can be downloaded and modified for any model.

66 In principle, this atmospheric module can be applied to any model configuration, e.g.
67 coarse grids, eddy resolutions and for regional to global configurations, without tuning.
68 However, we show in the following that in certain regions, this model can introduce a drift
69 due to unmodelled physical processes. In such regions (mainly the tropics), it is preferable to
70 adjust some parameters or to introduce a relaxation towards climatological values. Another
71 limitation might be the availability of a fine resolution wind field (spatial and temporal) for
72 meso- or submeso-scale ocean applications. We discuss the effect of wind-SST interaction
73 (Small et al. (2008)) even though this has not been implemented in CheapAML.

74 The paper is organized as follows. The model is described in Section 2. In Sections 3 and
75 4 we describe the results of two verification experiments. First a regional experiment in the
76 Gulf Stream is presented, followed by a global modeling application. The conclusions are
77 given in Section 5. Appendix A compares several methods currently available to compute
78 latent and sensible heat flux.

79 **2. Model Equations**

80 *a. Main equations*

81 The basic assumptions of CheapAML are that atmospheric reanalysis variables like hu-
82 midity and temperature are accurate on large scales and of these the least sensitive to ocean
83 surface structure is velocity. We thus accept atmospheric velocity as a known and develop
84 equations governing the atmospheric tracer fields of temperature and water. This shortcut
85 avoids the complexities of atmospheric dynamics and instead concentrates on thermody-
86 namics. The shortcomings of this assumption are discussed by Small et al. (2008) who
87 demonstrate that the wind can be modified by ocean mesoscale eddies. This in turn can im-
88 pact air-sea fluxes. Including this in our boundary layer model would essentially turn it into
89 a coupled model and we opt not to do so. The fundamental equation solved by CheapAML
90 is:

$$s_t + ADV(s) = -F_z + \nabla \cdot (K\nabla s) - \lambda(s - s_c), \quad (1)$$

91 where s is either atmospheric potential temperature, T (degrees Celsius), or water vapor
 92 content, q (kg water/kg air), F is the appropriate property flux whose vertical divergence
 93 influences s , K is an atmospheric diffusivity, λ is an inverse of a relaxation time scale, and s_c
 94 is a specified value of temperature or humidity. We later discuss the impact of this relaxation
 95 term. The advection term is written in Boussinesq divergence form as

$$ADV(s) = \nabla \cdot (\mathbf{u}s). \quad (2)$$

96 Most atmospheric reanalysis products provide horizontal wind velocities at a given height
 97 (10 m often). These are used in the solution of (1), and for this reason, we regard (1) as
 98 representing the evolution of the tracer s at the standard height. Some advection schemes,
 99 like many monotone methods, require a three-dimensional velocity field, and later on we will
 100 argue that the calculation of precipitation is improved when including the vertical velocity
 101 as part of the computation. Consequently, vertical velocity, w , is diagnosed according to

$$w_z = -(u_x + v_y). \quad (3)$$

102 We also employ the simplifying assumption that the atmospheric boundary layer is de-
 103 scribed by a known, but possibly variable, thickness h . The model also employs a specifica-
 104 tion of the tracers s over land and allows time dependence in those specifications.

105 The tracers are governed by the forced advection and diffusion equation (1). The physical
 106 forcing is assumed to be governed primarily by turbulent vertical transports whose divergence
 107 enters into tracer evolution. For potential temperature, the vertical flux divergence at the
 108 standard height is estimated using

$$-F_z^T = \frac{F^+ - F^-}{\rho_a C_p h}, \quad (4)$$

109 where h is the boundary layer thickness, ρ_a the atmospheric density, C_p the atmospheric
 110 heat capacity and $F^{+,-}$ represents the energy fluxes at the top and bottom of the layer,
 111 respectively. Employing the convention that positive fluxes are upward, the formulae for the
 112 fluxes are

$$F^+ = -F_s^\downarrow + \frac{F_l^\uparrow}{2} + L, \quad (5)$$

$$F^- = -F_s^\downarrow + \frac{F_l^\downarrow}{2} + F_{ol}^\uparrow + L + S, \quad (6)$$

113 where F_s^\downarrow is the solar short wave flux, $F_l^{\uparrow\downarrow}$ the up and downwelled atmospheric long wave
 114 flux, F_{ol}^\uparrow the upwelled oceanic long wave flux, L latent heat flux and S sensible heat flux.
 115 The boundary layer model is meant as a sub cloud layer model, implying that condensation
 116 happens at the top of the boundary layer. Therefore the latent heat release associated
 117 with the condensation is not realized in the boundary layer, but instead escapes to the
 118 atmosphere above. The latent flux, L , is thus common to both the formulae in Eqs. (5-6).
 119 These turbulent fluxes are computed using a user chosen algorithm (currently, the options
 120 are LP82 or COARE3 algorithms).

121 Note that solar shortwave is common to both fluxes, implying that it transits the atmo-
 122 spheric layer without loss. This is not precisely true, but implies that the contents of solar
 123 forcing should be the net forcing absorbed at the surface, accounting for albedo reflection.
 124 Long wave radiation is computed according to the standard Stefan-Boltzmann law:

$$F_l = \epsilon\sigma T^4, \quad (7)$$

125 where ϵ is an emissivity. These empirical parameterizations have been found to yield accurate
 126 estimates and are consistent with approximations about the optical depths of the atmosphere
 127 reflecting the level of model simplification (Talley et al. 2011).

128 In order to accurately compute the net heat flux at the ocean surface, we must account
 129 for the emission of long wave radiations by clouds and aerosols. The dynamics are not simple

130 and depend upon detailed cloud structure. Clark et al. (1974) proposed a formulation for
 131 the net longwave at the ocean surface (see also the review by Josey et al. 1997):

$$F_l^{net} = \epsilon\sigma SST^4(0.39 - 0.05\sqrt{e})(1 - \lambda C^2) + 4\epsilon\sigma SST^3(SST - T), \quad (8)$$

132 with C an externally provided cloud fraction, e the water pressure in millibar and λ is a
 133 latitude dependent coefficient. We use here $\lambda = 0.5 + |\textit{latitude}|/230$ (see Clark et al. 1974),
 134 and the latitude expressed in degrees from the equator.

135 Water vapor forcing takes the form:

$$-F_z^q = \frac{E - F_q^\uparrow}{\rho_a h}, \quad (9)$$

136 where E and F_q^\uparrow represent evaporation and moisture entrainment at the top of the boundary
 137 layer respectively. Evaporation is computed as the latent heat flux divided by the latent
 138 heat of evaporation.

139 The flux of humidity F_q^\uparrow at the top of the boundary layer parameterizes water vapor
 140 entrainment and transport at the top of the boundary layer. We retain the same parame-
 141 terization as S95:

$$F_q^\uparrow = \mu\rho_a C_{de} |\mathbf{u}| q, \quad (10)$$

142 with C_{de} the exchange coefficient for evaporation, $|\mathbf{u}|$ the magnitude of the wind (see Sec. 2b),
 143 and μ a coefficient set to 0.25 (see also the discussion in S95). If we interpret the coefficient
 144 in Eq. (10) as a entrainment time scale τ_e , we have

$$\tau_e \sim \frac{h}{\mu C_{de} |\mathbf{u}|} \simeq 10 \text{ days}, \quad (11)$$

145 using the approximation $C_{de} = 10^{-3}$ and $|\mathbf{u}| = 5 \text{ m s}^{-1}$. These numbers may vary but do
 146 provide a time scale of the entrainment at the top of the boundary layer.

147 Precipitation is generally one of the most difficult atmospheric variables to predict and a
 148 model of this simplicity will suffer when applied to the wide variety of realistic precipitative
 149 conditions. Precipitation in the boundary layer only enters the water vapor budget as a
 150 small correction and we chose to not retain it. We only compute it as a diagnostic field for
 151 the ocean fresh water budget.

152 We here describe a parameterization that is physically justified and which has performed
 153 reasonably well in tests. The applications in the next sections illustrate its strengths and
 154 weaknesses, and we provide methods by which the weaknesses can be addressed.

155 In our parameterization, precipitation is directly related to vertical wind. We allow
 156 precipitation only if the vertical wind, w , is upward and adjust the precipitation according
 157 to the size of w . We compute the large scale precipitation (LSP) as:

$$\text{LSP} = \max\left(\rho_a h \frac{q - 0.7q_s}{\tau_1} \left(\frac{w}{w_0}\right)^2, 0\right), \quad (12)$$

158 q_s being the saturation specific humidity at the temperature T , τ_1 a precipitation time scale,
 159 w the vertical velocity, w_0 a reference vertical velocity. The multiplicative nondimensional
 160 term modulates the strength of the precipitation using a threshold set with w_0 . The square
 161 factor is set to better separate high and low values of vertical wind. The numerical values
 162 used in the following examples are: $\tau_1 = 40$ h and $w_0 = 7.5 \times 10^{-6}$ m s⁻¹. The value of τ_1
 163 is not really the precipitation time scale since it is modulated by $(w/w_0)^2$. The pattern of
 164 $(w/w_0)^2$ is essentially zero and reaches values of 10 in regions of intense upward wind.

This parameterization systematically underestimates precipitation near the equator. We
 therefor add a correction for the region where $q > 0.2$ kg/kg (see Fig. 7); the convective
 precipitation (CP) is computed as

$$\text{CP} = \max\left(\rho_a h \frac{q - 0.9q_s}{\tau_2}, 0\right), \quad (13)$$

165 with $\tau_2 = 6$ h. All these constants have been determined manually to match predicted
 166 and observed patterns. Parameter estimation via regression has proven unreliable since

167 precipitation is a very localized event. We assume that runoff is part of the ocean-land
168 interaction and is thus not represented here.

169 *b. Air-sea turbulent fluxes*

170 Several algorithms computing the turbulent momentum, heat and water fluxes at the
171 air-sea interface have been developed. An early attempt in wide usage is that in LP82, who
172 provide a formula for the primary drag C_d in terms of the air speed. They then proceed to
173 compute evaporation, sensible heat and stress according to

$$E = C_{de}|\mathbf{u}|(q_s^{SST} - q), \quad (14a)$$

$$S = C_{dh}|\mathbf{u}|(SST - T), \quad (14b)$$

$$\tau = C_{dd}|\mathbf{u}|^2, \quad (14c)$$

174 where the coefficients C_{dx} are computed simultaneously and q_s^{SST} is the saturation specific
175 humidity of the atmosphere evaluated at the local sea surface temperature. This flux calcu-
176 lation has recently been revisited by Fairall et al. (2003), who use Monin-Obukov similarity
177 theory to relate observations of atmospheric variables at standard heights and the stability
178 of the air column to air-sea fluxes. Provision is also made for the ocean wave state, when
179 computing the so-called roughness length. Our implementation of the COARE3 algorithm
180 assumes by default the wave model of Smith (1988) dependent upon the wind, but also
181 permits the specification of wave data. Other flux parameterizations are available, such as
182 Beljaars (1994) although they have not been added to CheapAML as of this writing: only
183 LP82 and COARE3 have been implemented. The various parameterizations yield somewhat
184 distinct estimates for the fluxes, as shown in Appendix A.

185 *c. Boundary values and relaxation*

186 CheapAML also requires the specification of the tracer s on the lateral boundaries (when
187 implemented in an open boundary configuration) and on land, and allows time dependence
188 in those specifications. The current version of CheapAML does not include a land module.
189 Instead, the temperature and humidity are strongly relaxed toward provided, and possibly
190 time varying, fields. This is the primary role of the last term of Eq. (1); thus, default
191 specifications are $1/\lambda = 2$ hours over land and $\lambda = 0$ over the ocean. A secondary use of the
192 λ parameter is to nudge the model towards observations in the manner of data assimilation
193 and thus correct for missing model physics. The quantity λ can be specified as a field variable
194 to facilitate this, with the limit of large λ everywhere converging to the classical case where
195 atmospheric variables are specified and fluxes are computed using bulk formulae.

196 *d. Height of the boundary layer*

197 The vertical fluxes of temperature and humidity given in Eqs. (4) and (9) both depend
198 on the height of the boundary layer h . The default configuration of CheapAML assumes
199 a uniform h value of 1000 m, a value which provides demonstrably useful fluxes. How-
200 ever, our experience is that it is often advantageous to provide CheapAML with readily
201 available boundary layer thickness data. For example, for global configurations, a constant,
202 single value for the boundary layer height fails to return broadly accurate fluxes and marine
203 boundary layer behavior. In the extra tropics, h varies seasonally from lower than 500 m in
204 the summer to more than 1200 m in the winter. In the tropics h remains between 600 m
205 and 1000 m all year long. CheapAML is designed to optionally accept temporally varying h
206 values, such as are available from the ERA reanalysis data set.

207 In all the following experiments, we use the daily varying climatology of the boundary
208 layer height provided by the ERA data set.

3. A Regional experiment

a. Mean and variability

To illustrate and assess CheapAML performance, we apply our model to the separated Gulf Stream (GS), where ocean eddying, weather and flux are particularly strong. Here, the SST can vary by more than 10 degrees C over 100 km, which is precisely the characteristic grid space of common reanalysis data sets (Kalnay et al. (1996) for the NCEP/NCAR reanalysis, Uppala et al. (2005) for the ERA reanalysis).

We examine here how CheapAML compares to ERA atmospheric temperature, humidity and fluxes in a confined regional configuration with prescribed SST. The boundaries of the chosen domain are $75^{\circ}\text{W} - 45^{\circ}\text{W}$ in longitude and $34^{\circ}\text{N} - 45^{\circ}\text{N}$ in latitude. 'Truth' fluxes have been computed by specifying ERA wind velocity, temperature and humidity at six hourly intervals, with linear interpolation to times in between. CheapAML fluxes are computed by specifying boundary atmospheric variables and default λ values. The underlying SST also comes from the ERA data set. Here again, at each time step, the SST is interpolated between the two nearest records. All fields are spatially interpolated to the finer resolution of $1/12$ degrees in latitude and $1/10$ degree in longitude (a roughly isotropic grid). 'Truth' is accepted as the ERA computed fluxes.

Figures 1 and 2 summarize the evolution of the atmospheric variables for January, 2007. We plot the monthly mean CheapAML and ERA atmospheric temperature and humidity. The mean temperature difference between CheapAML and ERA does not exceed 0.7 degrees C (Fig. 1). There is a cold bias of CheapAML relative to the 'Truth' over the GS path and a warm bias elsewhere. There is also good agreement in the variability pattern of temperature (computed on a daily basis) in both cases. It is slightly over estimated by 0.7 degrees C over the cold side of the GS and underestimated by 0.5 degrees C over the warm side.

We draw similar conclusions when looking at the mean and standard deviation of daily

235 values of humidity (Fig. 2). The mean humidity is overestimated outside of the GS path
236 by 0.8 g/kg in accordance with the temperature bias. The pattern of variability has a
237 similar shape in ERA and CheapAML. In the latter, it is underestimated everywhere with
238 a maximum over the warm side of the front.

239 The heat and moisture fluxes (not shown) also show similar results, i.e. differences are
240 in the 10–20% range, which is well within the uncertainty of bulk flux parameterizations
241 (see also Appendix A). Comparable results are found if fluxes are computed using the S95
242 approach. An improvement of CheapAML relative to S95 lies in the variability pattern. The
243 variability of temperature and humidity are underestimated over the entire region by more
244 than 1 degree C and 1 g/kg respectively by S95. This relatively weak variability is consistent
245 with the equilibrium hypothesis, and reflects a lack of 'weather' on air-sea exchange.

246 *b. Representation of the extreme events*

247 To further examine the time variability, we compare several time series of temperature
248 taken in the middle of the domain (298°W, 39°N). In Fig. 3a, we compare a 3 months time
249 series of temperature from the ERA reanalysis (thick black line), CheapAML (blue line) and
250 S95 (red line). Both S95 and CheapAML are in good agreement with ERA, with correlations
251 of 0.98 and 0.99 respectively. However bias appears especially in the representation of the
252 extremes (cold and warm events). We plot on Fig. 3b the Probability Density Function
253 (PDF) of these three time series using the same plot convention. CheapAML represents
254 correctly both warm and cold events, if slightly overestimating the cold. The global shape of
255 the ERA PDF is well reproduced — especially the bimodality that is observed during this
256 period. The PDF of the S95 simulation is much more peaked at the center of the distribution.
257 These results reflect the S95 equilibrium hypothesis.

258 The PDF of humidity at the same location and for the same period is plotted on Fig. 3c.
259 Whereas the S95 simulation has a peaked distribution of humidity, CheapAML recreates
260 more accurately the observed distribution.

261 The sensible and latent heat flux time series at this location are also well captured by
262 both CheapAML and S95 (correlation above 0.98 in each case). We only show on Fig. 4
263 the PDF of these time series. The thick black line corresponds to heat fluxes computed
264 using surface fields from ERA but applying the COARE3 formulation. The dashed line
265 corresponds to raw heat fluxes extracted from the ERA dataset. The comparison illustrates
266 the differences inherent to state-of-the-art flux algorithms, which are considerable (see also
267 Appendix A). Once again CheapAML yields a better PDF when compared to the COARE3
268 implementation than does S95.

269 *c. Impact on the oceanic circulation*

270 We now extend the simulations to a more realistic ocean experiment. Our intent is to
271 illustrate the impact of the two AML formulation (S95 and CheapAML) on the oceanic
272 structure. We still focus on the region of the separated GS and use the MITgcm (Marshall
273 et al. 1997) with open boundary conditions. The model obtains boundary data from the
274 HYCOM ocean reanalysis dataset (Chassignet et al. 2007). The starting date is January, 1st
275 2007 and the model is run for one month. The spatial resolution is 1/12 degrees in latitude
276 and 1/10 degrees in longitude. We use 39 vertical levels with a 'fine' resolution of the upper
277 levels (10 m) and low resolution of the lower level (500 m). This resolution resembles the one
278 employed in many modern global OGCMs. The oceanic mixed layer is computed according
279 to the KPP formulation (Large et al. 1994) and the mixed layer depth is estimated from a
280 Richardson number criteria.

281 We first compute the air-sea exchange using S95 and then using CheapAML. Since the
282 oceanic states quickly differ after several days, it is not useful to compare the mixed layer
283 depth at a specific location. We plot in Fig. 5 the evolution of the mixed layer depth averaged
284 over the entire domain. The S95 integration is plotted with a dashed line, the CheapAML
285 integration is plotted with a thin line and the HYCOM reanalysis is plotted with a thick
286 line. During this month the mixed layer depth increases everywhere in both simulations, as

287 expected. The first 15 days are well reproduced in the CheapAML run whereas the last 15
288 days are better reproduced in the S95 run. After one month, we observe a notable difference
289 of 50 m in the two runs whereas the HYCOM reanalysis lies in between this two estimates.
290 The only possible reason to explain these two evolutions lies in the differences in the heat
291 and fresh water fluxes. This illustrates the discrepancies of the oceanic state that occur due
292 to uncertainties in forcings.

293 The spatial average of net heat flux is given in Fig. 5. The associated mean values for
294 this period are: 303 W m^{-2} for CheapAML, 486 W m^{-2} for S95, 357 W m^{-2} for Hycom, and
295 304 W m^{-2} for OAFflux (Yu and Weller 2007). The favorable comparison between OAFflux
296 and CheapAML partly reflects their common basis in the COAR3 algorithm, but they do use
297 scatterometer winds while we use ERA winds. The small difference between our numbers
298 is consistent with the idea that local feedbacks on fluxes due to wind modifications by the
299 oceanic mesoscale are not a major systematic error.

300 The overall character of the means is consistent with the mixed layers diagnosed in the
301 three cases. The interpretation of the differences between The CheapAML and S95 with
302 Hycom is unclear since the latter is an assimilative product.

303 Nevertheless, we can associate the rapid deepening of the mixed layer with the peaks
304 of the net heat flux. S95, which consistently predicts a higher net heat flux, has a more
305 pronounced mixed layer deepening rate. While all the net heat flux curves are well correlated
306 (above 0.9 for all pairs), the magnitude of the storm peaks are very different. For Jan, 1st
307 (same ocean state for all experiments), we note a difference of 300 W m^2 between OAFflux
308 and Hycom. The maxima in net flux during storms are either reached by Hycom or S95.
309 In contrast, Hycom, OAFflux and CheapAML agree well at the minima (Jan, 7th and Jan,
310 16th) — with a difference of less than 50 W m^2 between them.

311 4. Global experiment

312 In the previous section, we demonstrated the benefits of using CheapAML in a regional
313 model. We here show how CheapAML can be used for global experiments, where we again
314 use prescribed SSTs. In a configuration corresponding to the ERA setup (1.125 degrees
315 resolution), we constrain the SST to follow the ERA SST field (perfect ocean model exper-
316 iment). Over the ocean, we let the air temperature and humidity adjust via CheapAML.
317 The model is initiated in January, 2000 and run for thirteen months; we focus on the last
318 month. We do not have a varying land-sea mask and treat sea-ice points as land points.
319 Atmospheric forcing is drawn from ERA. CheapAML is deployed using standard λ values
320 and ERA boundary layer heights.

321 *a. Atmospheric variables*

322 We compare the mean temperature computed in January 2001 with that in ERA (Fig. 6).
323 The bottom panel of Fig. 6 corresponds to the differences between the middle and the top
324 panel. The largest bias is observed in the Northern hemisphere. More generally, when
325 looking at the June-July-August maps, we conclude that the largest bias occurs in the
326 winter hemisphere. We observe a cold bias near the western boundaries whereas the center
327 of the Atlantic and Pacific are subject to a warm bias. In the tropics, we observe a warm
328 bias in the region of strong convection. These errors reflect processes not modeled here but
329 do not exceed 1.5 degrees C.

330 Figure 7 focuses on the mean humidity field. Again, we note that the patterns in the two
331 upper panels are in accordance. The humidity maximum in the tropics is well reproduced
332 as are the specific patterns in the extra-tropics. The bottom panel is the difference between
333 the simulated and observed humidity. The maximum differences reach ± 1 g/kg (typically
334 O(5-10%)) in the tropics as well as in the extra-tropics. The humidity is mainly overestimated
335 in the northern and southern part of the oceanic basins whereas it is mostly underestimated

336 in the tropics.

337 The regions where the air temperature or humidity fields are not well estimated reveals
338 the zones where the outgoing flux at the top of the boundary layer or the precipitation is not
339 properly described in our model. Other physical phenomena are thus at work in these regions
340 (convection, clouds, vertical motion). Knowing these discrepancies, one could structure λ so
341 as to minimize model bias.

342 The precipitation field for January 2001 is plotted on Fig. 8. Although the parameteri-
343 zation mentioned in Sec. 2 is extremely simple, some skill is observed. The two upper panels
344 of Fig. 8 argue that the global patterns of the convective precipitation are well reproduced
345 in the tropics. However, the large scale precipitation in the northern Pacific and Atlantic
346 are underestimated. These precipitations are associated with the position and strength of
347 the stormtrack and can not be easily reproduced using our single layer model.

348 *b. Net heat flux*

349 Figure 9 compares the net heat flux computed by CheapAML with that given by OAFflux
350 (Yu and Weller 2007) for January 2001. CheapAML captures correctly the large scale pat-
351 tern (enhanced heat flux over the western boundary current in January and North-South
352 asymmetry), but appears to underestimate the heat flux. In the northern hemisphere where
353 the mean heat flux is positive, it is underestimated by $[20 - 70] \text{ Wm}^{-2}$. This pattern is
354 clearly related to the temperature pattern anomaly observed in Fig. 6: an atmosphere that
355 is too warm prevents strong air sea-flux in that region. The difference seen in the southern
356 hemisphere has a pattern that is very similar to the humidity differences: a dry mixed layer
357 leads to a larger evaporation and thus an increase of the net heat flux. In July the situation
358 is reversed (not shown). This is also in accordance with the temperature and humidity dif-
359 ferences observed in July (not shown): the northern hemisphere Q_{net} is large enough. Note
360 also that some differences might also be explained by the longwave parameterization (see
361 Eq. 8), although this heat source remains small compared to the sensible and latent heat

362 flux.

363 5. Conclusions

364 a. Summary

365 We introduce here a simple atmospheric boundary layer model for the computation of air-
366 sea exchange in ocean-only modeling. In the boundary layer, temperature and humidity are
367 advected by a prescribed wind. Temperature and humidity adjust with the underlying SST
368 mainly through sensible and latent heat exchanges. The value of this model is to capture
369 part of the non-local feedback of the ocean surface on air-sea exchanges, while stopping
370 well short of computing a full coupled ocean-atmosphere model. We believe that for an
371 oceanic model, it is preferable to use CheapAML than to prescribe the temperature and
372 humidity (or fluxes) from a reanalysis data set: as soon as the oceanic state deviates from
373 the observed state, the reanalysis temperature and humidity fields and the oceanic state are
374 not related anymore. The computational cost of using CheapAML is minimal, and does not
375 materially increase the execution time of the model run. Furthermore, CheapAML captures
376 the 'weather' impacts of the atmosphere on air-sea exchange with improved fidelity relative
377 to its predecessor, S95.

378 Using a regional and a global configuration we tested the skills of CheapAML. In a
379 small region subject to large spatial variations of SST, we show that this slab atmospheric
380 model is able to accurately reconstruct the mean temperature and humidity fields as well
381 as their variability. Analyzing several time series of atmospheric tracers and fluxes at given
382 locations, we argue that CheapAML reproduces the mean as well as the extreme events
383 correctly. The extreme events (for e.g. cold air outbreak) are of great importance for the
384 oceanic dynamics. We illustrate this impact using the evolution of the mixed layer depth
385 when the ocean is subject to these different fluxes. When deployed globally, zones appear
386 where temperature and humidity are subject to biases. These biases, although small, are

387 inherent to the simplifications performed to construct this model, and can be reduced through
388 nudging.

389 The main differences between this model and its predecessor Seager et al. (1995) are
390 the elimination of the equilibrium assumption and the provision of a water budget. Here,
391 we explicitly integrate in time the equation of evolution of temperature and humidity. We
392 also updated the computation of the air-sea fluxes using a more recent formulation (Fairall
393 et al. 2003). An accurate computation of the air-sea fluxes is in fact the primary goal of this
394 study. The fresh water flux budget exhibits similarity with the observed fresh water budget
395 although a better representation of precipitation might help to increase the accuracy of this
396 forcing. Moreover, we propose here a fully parallel code whereas the computation of S95
397 requires the knowledge of atmospheric variables in the entire domain and is thus harder to
398 parallelize.

399 *b. Remaining issues*

400 Among issues for future development are the development of atmosphere-land and at-
401 mosphere sea-ice modules. Such regions are handled by means of strong relaxation towards
402 specified values; these are clear areas for improvement.

403 Clouds are also not parameterized in this model. It is however possible to adjust the
404 solar input to mimic the presence of clouds, although not dynamically.

405 Several studies (see Small et al. 2008, for a review) indicate that there is a small correla-
406 tion observed between the wind speed and the SST; the wind being accelerated over warm
407 SST. This interaction could also lead to some possible refinement of our model, especially
408 in regions of strong SST fronts or eddies. Pezzi et al. (2004) and Jin et al. (2009) proposed
409 parameterizations of the wind–mesoscale eddies interaction. According to their results, the
410 detailed structure of the oceanic eddies is affected by this interaction. How this may impact
411 the large scale ocean circulation remains to be seen.

412 *c. Practical use of CheapAML*

413 We recommend the use of CheapAML via the MITgcm (Marshall et al. 1997), where
414 it was first developed as a package. Several options are available, e.g. formulation of the
415 fluxes (LP82 or COARE 3) or choice of the advection scheme (flux limited versus centered
416 differences). It can be used either for regional modeling purposes (open boundary condi-
417 tions) or for global modeling (zonally periodic boundary conditions). The current version of
418 CheapAML assumes the model domain is bounded by constant grid lines, eg, for a sphere,
419 the boundary consists of one northern and one southern latitude, and single eastern and
420 western longitudes.

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The turbulent air-sea fluxes

Figure 10 is an example of the differences observed in the strength of the latent and sensible heat fluxes when computed using several methods. Several studies already mention the differences between these products (cf. Kubota et al. (2008) and references therein for an example over the Kuroshio Extension or Rouault et al. (2003) for the aghulas current or Kubota et al. (2003) for a global comparison).

We report four different computations of the turbulent heat fluxes for January 2007: the ERA reanalysis Beljaars (1994), the NCEP/NCAR reanalysis Kalnay et al. (1996), COARE 3 using ERA temperature, humidity and wind and LP82 also using ERA surface variables.

We observe large differences especially for the sensible heat flux estimations. The latent and sensible heat fluxes are maximum when estimated with LP82. They reach respectively 670 and $250^{\circ}\text{W m}^{-2}$. The patterns are the same for three computation that uses ERA reanalysis. It reflects the presence of meanders in the GS. The coarse resolution of the NCEP/NCAR reanalysis does not allow a fine comparison. However we clearly see that there is a good agreement in the magnitude of NCEP and COARE 3.

The comparison of SH-ERA40 and SH-COARE3 is consistent with Fig. 4. It appears that ERA40 produces significantly lower sensible heat fluxes than COARE3. This difference of almost 100 W m^2 over the warm core of the Gulf Stream can have tremendous effects on the oceanic circulation as illustrated in Fig. 5. Since all atmospheric and oceanic variables are the same in that case in this computation, this difference is only due to the estimation of the exchange coefficient C_d (see Eq. 14).

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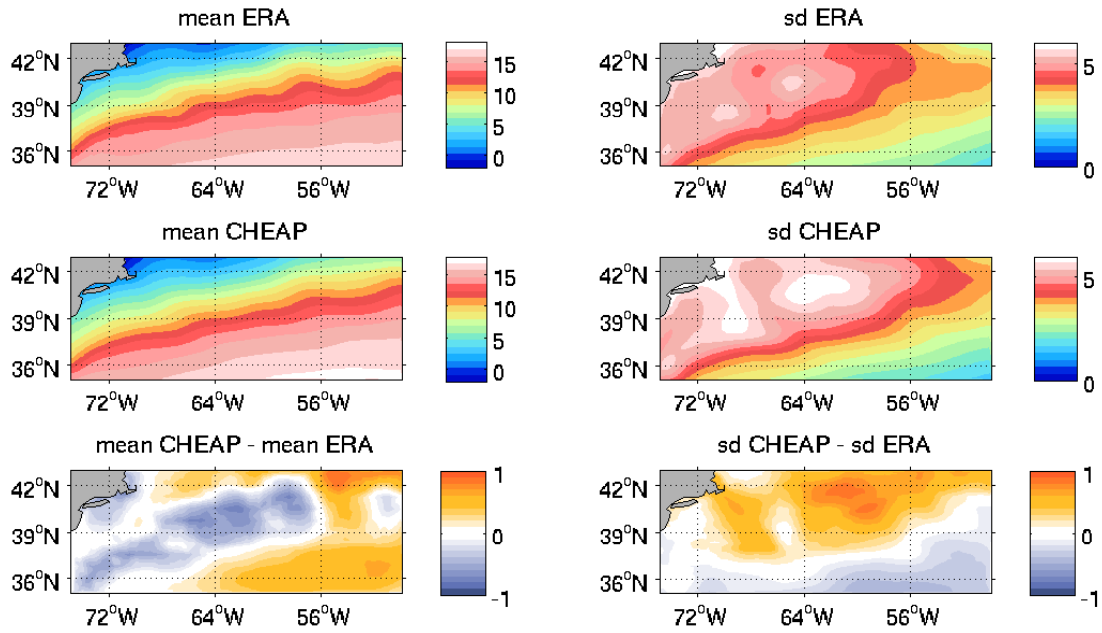


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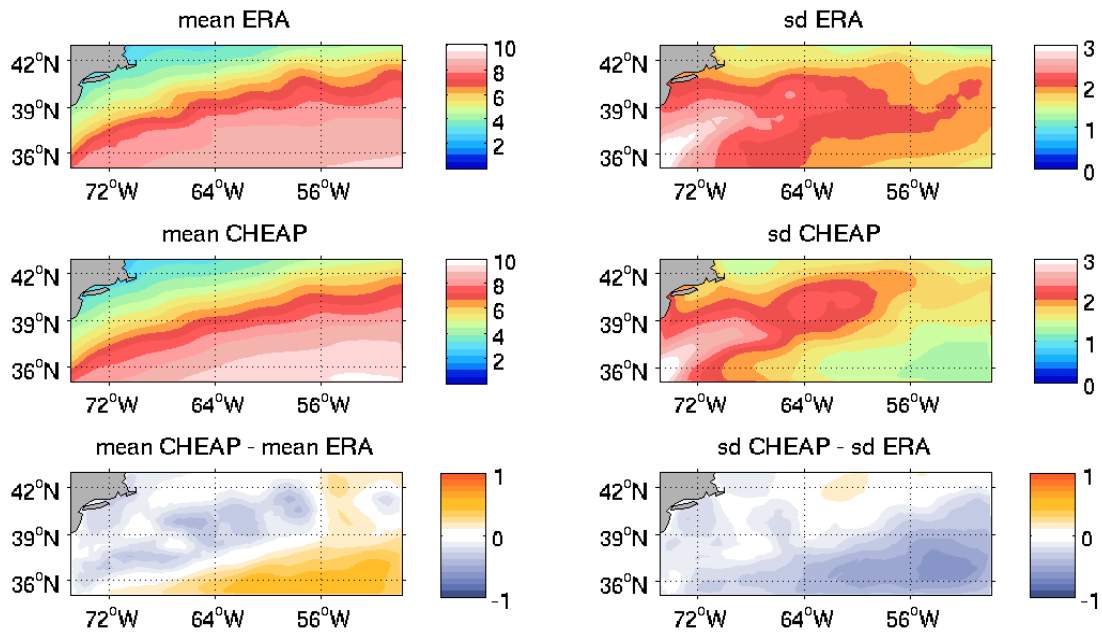
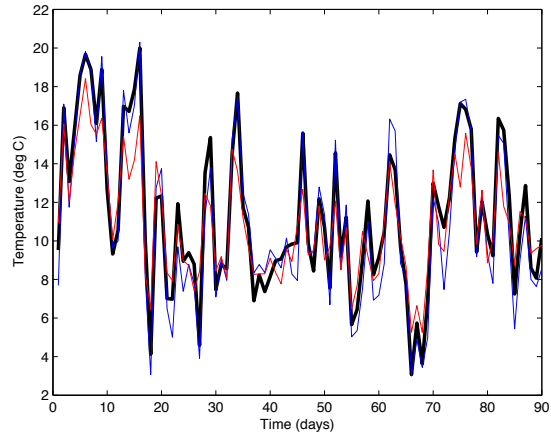
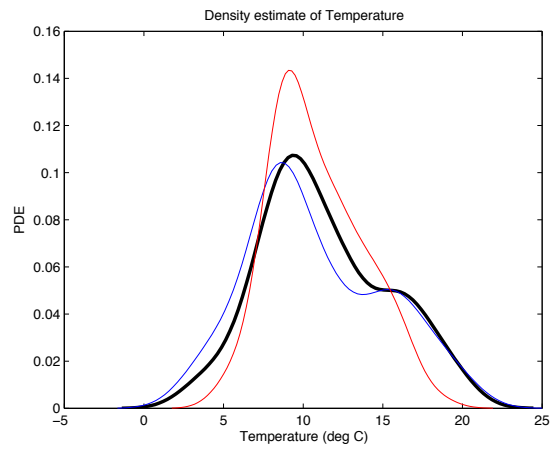


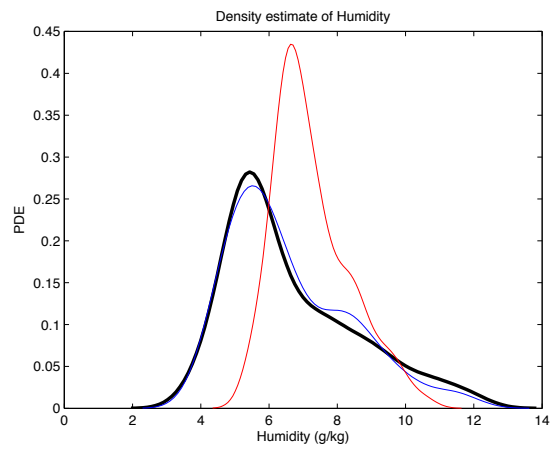
FIG. 2. Same as Fig. 1 but for humidity. (units are g/kg)



(a)



(b)



(c)

FIG. 3. Top: time series of the atmospheric temperature in the middle of the regional model (cf. Figs. 1-2) from January, 1st 2007 to Mar, 31st 2007. The thick black line is the ERA temperature; blue CheapAML and red is S95. Middle: PDF of these three time series. Bottom: PDF of the humidity time series using the same color conventions.

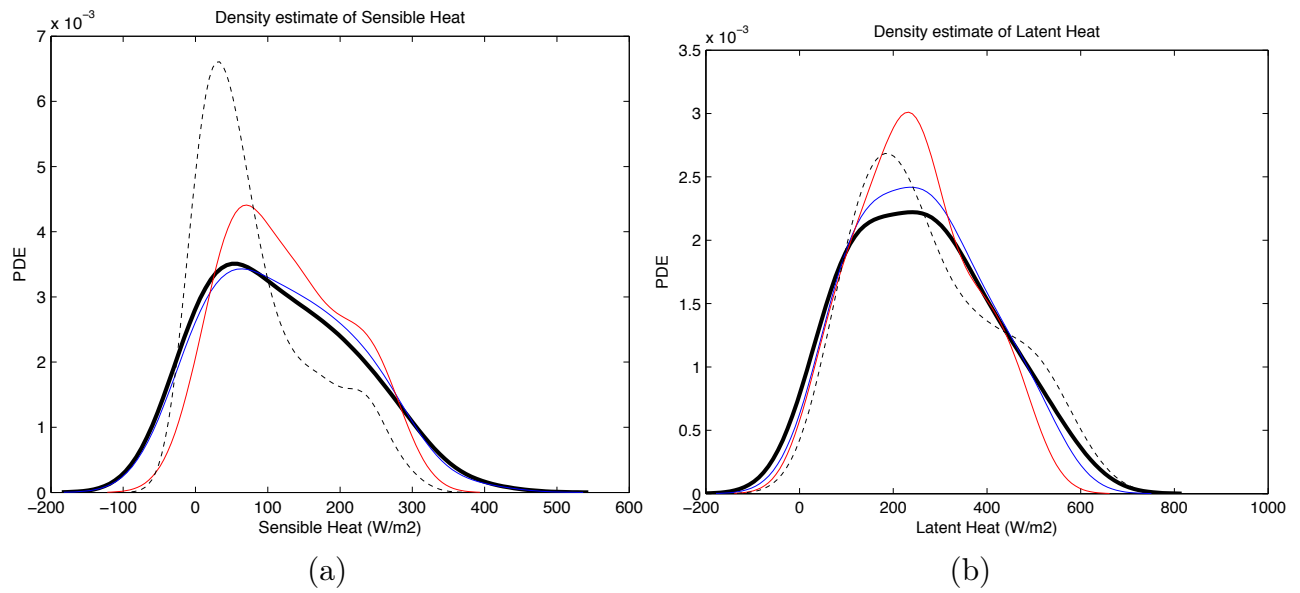


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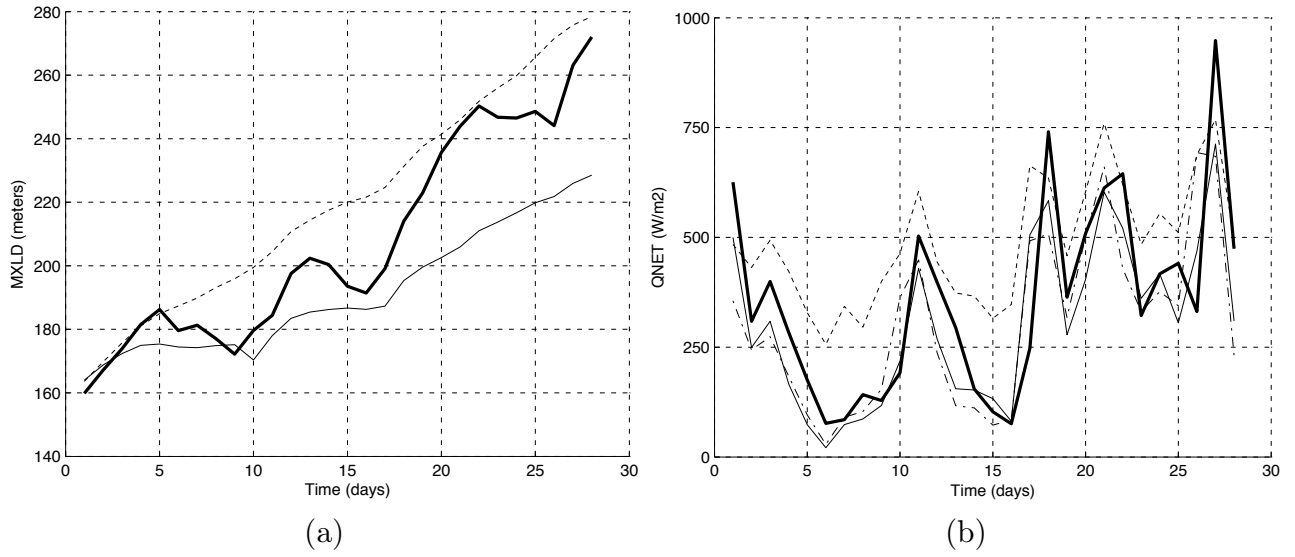


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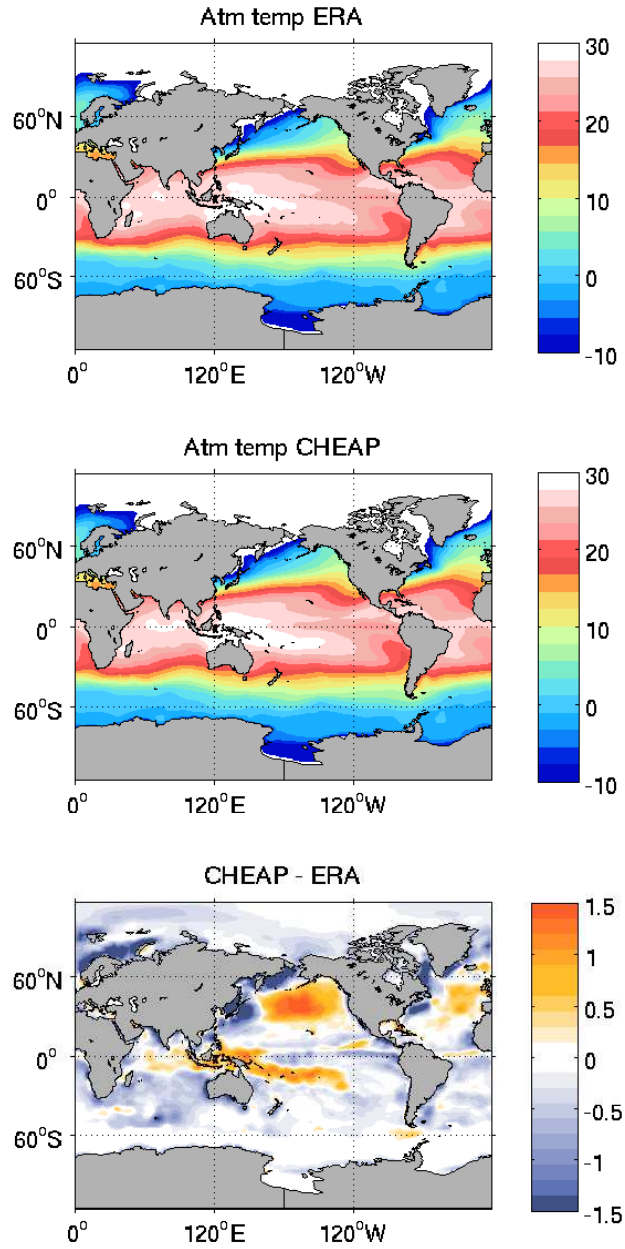


FIG. 6. Top: mean atmospheric lower layer temperature in January 2001 (data from ERA). Middle: reconstructed temperature in January 2001 with CheapAML starting in January 1st 2000. Lower panel: difference between the middle and top panel. (units are degrees C)

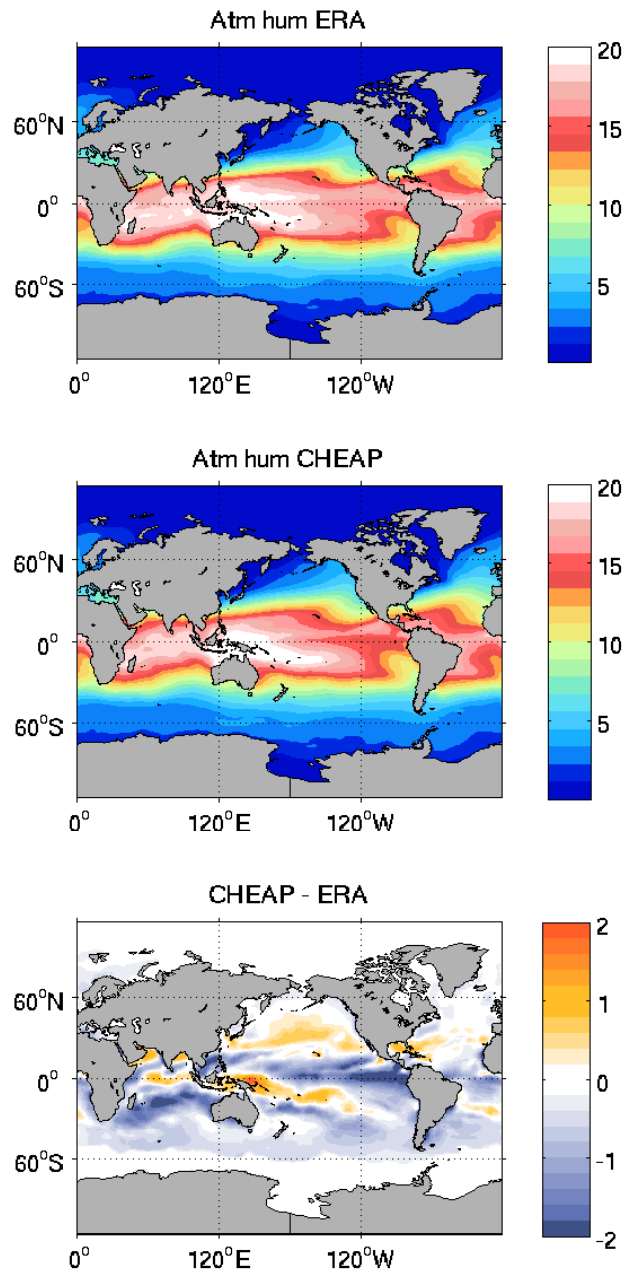


FIG. 7. Same as Fig. 6 but for humidity (units are g/kg).

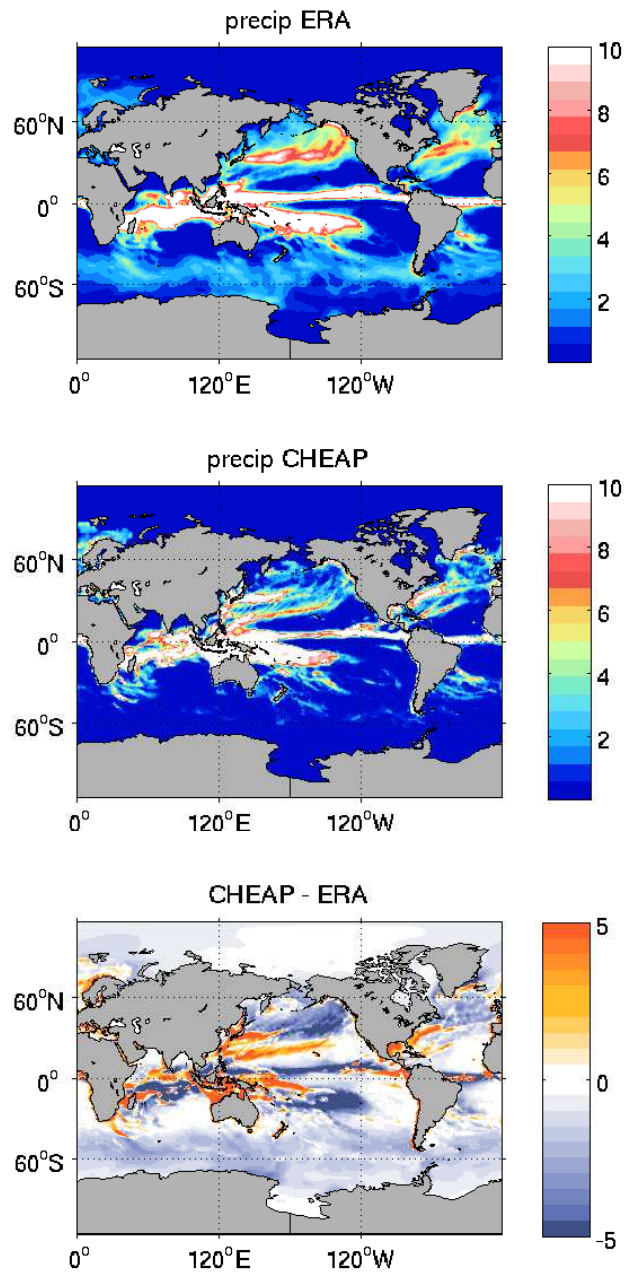


FIG. 8. Same as Fig. 6 but for precipitation (units: mm/day).

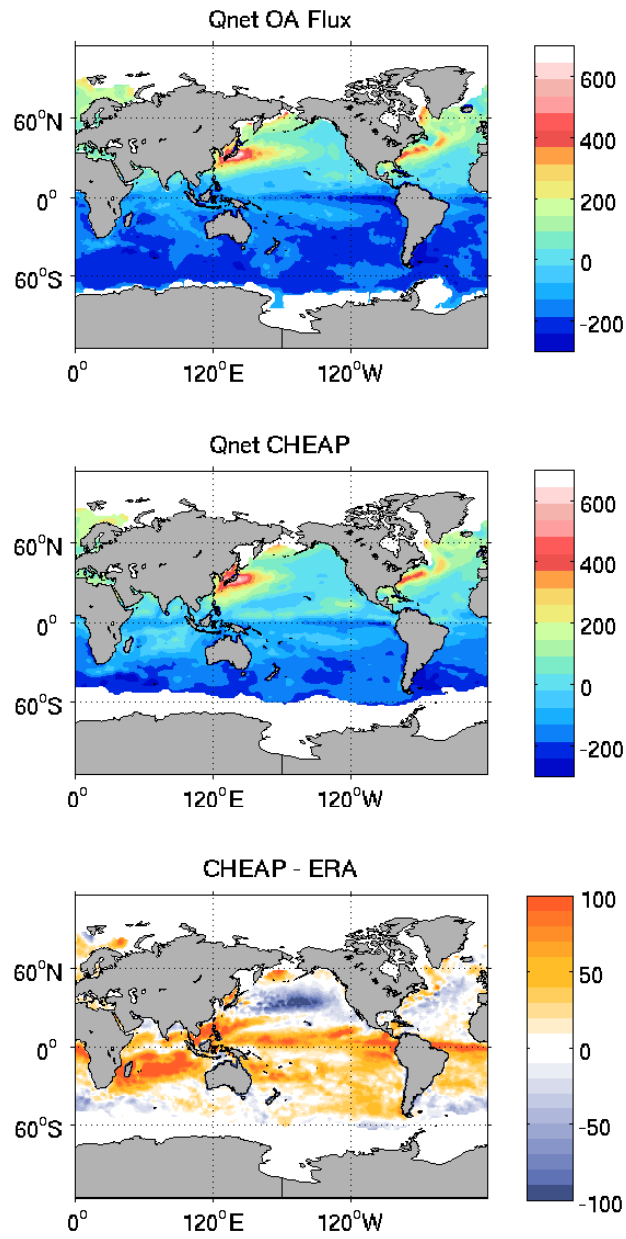


FIG. 9. Same as Fig. 6 but for Qnet (units: Wm^{-2}).

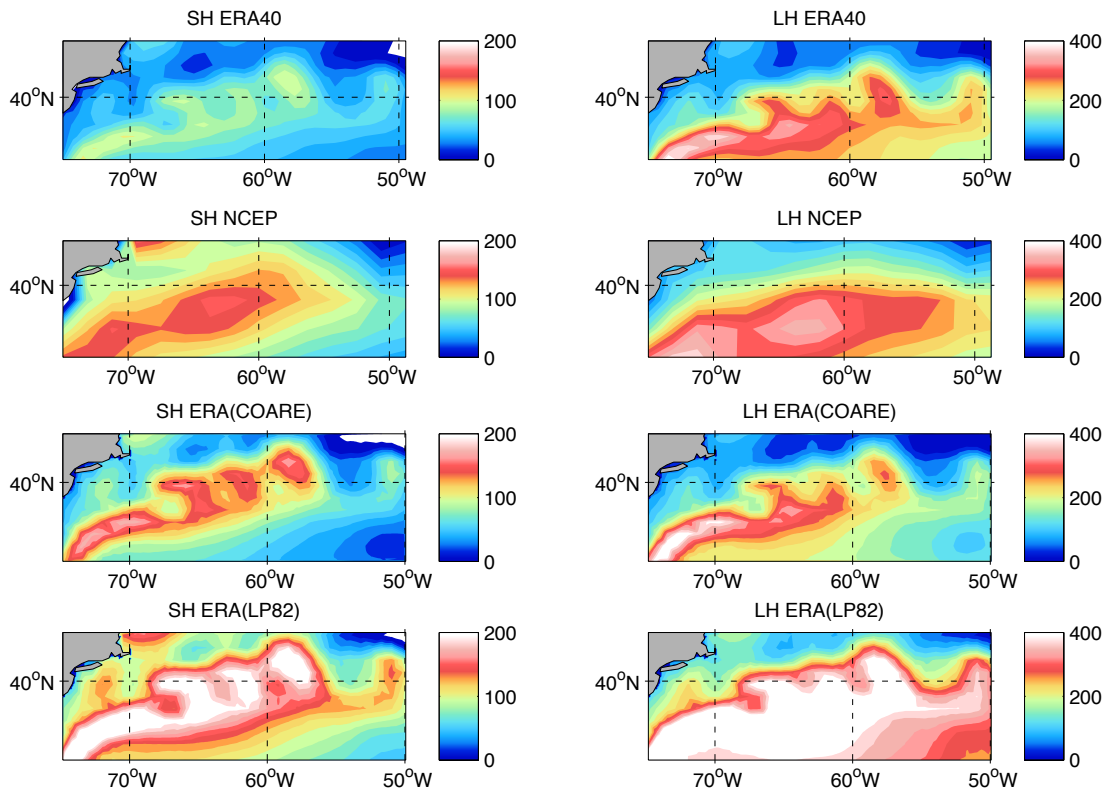


FIG. 10. Comparison of sensible (left column) and latent (right column) heat fluxes computed using different bulk formulae in January 2007. First row: ERA values Beljaars (1994); second row NCEP/NCAR reanalysis; third row: COARE 3 using ERA atmospheric and oceanic fields; last row: LP82 using ERA atmospheric and oceanic fields. (units are W m^{-2})