



2 Tropical Atlantic sea surface temperature and heat flux simulations in 3 a coupled GCM

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5 Received 4 April 2006; revised 5 June 2006; accepted 12 June 2006; published XX Month 2006.

6 [1] This paper contrasts the SST and heat flux errors in the
7 Tropical Atlantic simulated by the CPTEC Coupled ocean-
8 atmosphere General Circulation Model and its oceanic
9 model forced by momentum and heat estimates. Comparisons between solar radiation estimated by satellite
10 and measurements of PIRATA buoys have been made with
11 the purpose of analyzing the impact of solar radiation in the
12 simulation of SST in the tropical Atlantic. The radiative
13 transfer model (ISCCP DX) has shown higher correlation
14 with the buoys data than ECMWF ERA40 with larger
15 differences over the eastern tropical Atlantic, where the
16 numerical prediction models present difficulties in
17 simulating the appearance of stratus clouds. The use of
18 solar radiation based on satellite estimates and
19 parameterized heat flux generated the best SST and
20 surface heat fluxes. The stronger surface stresses
21 generated by the CGCM contributed to generating an
22 oceanic thermal structure in closer agreement with
23 observations than the OGCM runs. **Citation:** Siqueira, L.,
24 and P. Nobre (2006), Tropical Atlantic sea surface temperature
25 and heat flux simulations in a coupled GCM, *Geophys. Res. Lett.*,
26 33, LXXXXX, doi:10.1029/2006GL026528.
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30 1. Introduction

31 [2] Sea Surface Temperature (SST) in the tropical oceans
32 are influenced by the heat flux across the ocean surface,
33 horizontal advection, upwelling, and mixing processes. A
34 change in the balance among these processes causes SST
35 variations, that on interannual and seasonal time scales
36 reflect profound changes in the circulation of the entire
37 tropical oceans [Philander, 1990]. The seasonal cycle of the
38 atmosphere-ocean system is determined by complex inter-
39 actions and feedbacks between elements of the system.
40 Many ocean properties show strong links to overlying
41 atmospheric variability, suggesting that much of the
42 observed ocean variability is driven by the atmosphere.

43 [3] The amplitude of the tropical Atlantic SST annual
44 cycle is almost an order of magnitude larger than SST
45 interannual variability [Merle and Hisard, 1980], suggesting
46 that the later might depend on SST annual cycle. On the
47 other hand, the simulation of SST annual cycle by a coupled
48 ocean-atmosphere GCM (CGCM) is sensitive to the
49 strength/deficiencies of the CGCM's component models.
50 Therefore, understanding these sensitivities is useful to
51 achieve further insight into mechanisms at work for
52 ocean-atmosphere interactions.

[4] A possible way to look into the sensitivities of the
coupled system is to compare simulations performed by a
CGCM and by its oceanic component model (OGCM)
forced by observational estimates of heat and momentum
fluxes. This study presents such a comparison in the context
of the annual evolution of SST and surface heat flux
simulated by the CPTEC CGCM and its OGCM (GFDL
Modular Ocean Model version 3). Section 2 describes the
models used, simulations performed, and data sets used for
forcing the OGCM and for model validation. Section 3
compares the annual mean evolution of surface heat flux
and SST produced by the CGCM, OGCM, and observa-
tional estimates. Section 4 focuses on the temporal evolu-
tion of SST and surface heat flux simulations on specific
locations. Section 5 summarizes the results and conclusions.

2. Models, Simulations, and Data Sets

[5] The CGCM used in this study consists of a low
resolution version of the CPTEC/COLA Global AGCM
[Cavalcanti et al., 2002] coupled to GFDL's MOM3
OGCM. The AGCM has 28 layers in the vertical (with
top at 50mb) and triangular horizontal truncation at wave
number 42, which corresponds to a horizontal resolution of
 $2.815^\circ \times 2.815^\circ$ (T42L28).

[6] The ocean model used in the CPTEC CGCM is the
Modular Ocean Model (MOM) version 3 [Pacanowski and
Griffies, 1998], from the Geophysical Fluid Dynamics
Laboratory (GFDL) where global tropical oceans were
considered, with the ocean basins limited at 40°N and
 40°S . For the vertical resolution, 20 levels were adopted,
7 of them in the first 100m, spaced by 15m. The longitu-
dinal resolution is 1.5° , and the latitudinal resolution varies
gradually from $1/2^\circ$ between 10°S and 10°N to almost 3° at
 40°S and 40°N . The coupling area is the global tropics,
between 40°S and 40°N .

[7] Model's results intercomparison were conducted for
the year 1998 only, due to limitations in the solar radiation
data set estimated by satellite imagery available at the time
of this research. Therefore "Root Mean Squared Errors"
(RMSE) presented throughout this article are calculated for
12 monthly values for 1998. A set of three numerical
simulations was performed: one coupled CGCM run and
two uncoupled OGCM runs. The initial conditions for the
two OGCM simulations are taken from a 30 years long
OGCM integration (1969–1998) forced by ECMWF
ERA40 wind stress, climatological solar radiation
[Oberhuber, 1988], and surface heat fluxes parameterized
following Rosati and Miyakoda [1988]. After the spin-up
process, two OGCM forced runs were made during the year
1998, both of which used ECMWF ERA40 wind forcing:
one used solar radiation fields estimated from satellite

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t1.1 **Table 1.** Standard Deviation, Mean Error and Correlation Coefficients, for 5 PIRATA Buoy Locations

t1.2	PIRATA Buoy	PIRATA Std. Dev.	ISCCP Std. Dev.	ERA40 Std. Dev.	ISCCP Mean Error	ERA40 Mean Error	ISCCP CC	ERA40 CC
t1.3	15°N38W	50.59	53.81	48.94	52.11	10.77	0.89	0.67
t1.4	8°N38°W	56.19	67.73	68.57	57.31	-5.71	0.87	0.51
t1.5	0°N35°W	48.18	52.56	47.38	36.37	-27.98	0.90	0.57
t1.6	0°N0°E	42.04	47.41	54.77	57.14	-25.23	0.81	0.25
t1.7	10°S10°W	52.04	49.36	36.24	39.09	9.4	0.80	0.4

103 imagery (ISCCP DX) [Rossow and Zhang, 1995; Pinker et
 104 al., 1995] and parameterized surface heat flux following
 105 Rosati and Miyakoda [1988], and the other used total
 106 surface heat flux fields from ECMWF ERA40 reanalysis.

107 [8] The coupled simulation started in December 1997
 108 from the same oceanic IC as the forced OGCM runs and
 109 atmospheric IC from CPTEC AGCM forced by observed
 110 global SST. During the coupled simulation, observed global
 111 SSTs were used poleward of the coupling region.

112 [9] The verification data sets for surface flux are derived
 113 from the Comprehensive Ocean Data set (COADS), $2^\circ \times 2^\circ$
 114 spatial resolution and the in situ data sets from 9 PIRATA
 115 buoys. The SST verification data set corresponds to
 116 the monthly fields analyzed from the NOAA Optimum
 117 Interpolation Sea Surface Temperature Analysis project
 118 [Reynolds et al., 2002].

119 3. Simulations of Surface Heat Flux and SST

120 [10] The standard deviation, mean error, and correlation
 121 coefficients for 5 PIRATA buoy locations were computed
 122 and are presented in Table 1 in order to quantify the
 123 differences between the solar radiation fields used to force
 124 the two OGCM simulations. The higher correlation of
 125 ISCCP data with the PIRATA observations is partly due
 126 to the higher spacial resolution of the ISCCP solar radiation
 127 field. Yet, such correlation differences between the ISCCP
 128 and ERA40 solar radiation fields are larger over the eastern
 129 tropical Atlantic, where the numerical prediction models
 130 present difficulties in simulating the appearance of stratus
 131 clouds over cold waters.

132 [11] Figure 1 shows the net surface heat flux RMSE for
 133 the three simulations. The OGCM simulation produces the
 134 largest net heat flux errors in the northern and southeastern
 135 tropical Atlantic basin when using reanalysis fields
 136 (Figure 1a). The OGCM RMSE when using de satellite
 137 estimates of solar radiation (Figure 1b) presents comparably

smaller magnitudes than the other two simulations shown in 138
 Figure 1, mainly over the regions mentioned above. The 139
 CGCM (Figure 1c) shows RMSE spatial distribution similar 140
 to the OGCM forced by reanalysis fields (Figure 1a), except 141
 over the northern tropical Atlantic, where the CGCM RMSE 142
 are smaller, and off the coast of Guinea where CGCM RMSE 143
 are larger. As the next section will show, the major 144
 contributions to the RMSE shown in Figure 1 are 145
 deficiencies in the latent and radiative fluxes; sensible heat 146
 flux (figures not shown) are important only at higher 147
 latitudes and will not be discussed further. 148

[12] The comparison between the surface solar radiation 149
 RMSE fields for the ECMWF ERA40 and CPTEC CGCM 150
 relative to the ISCCP estimates (Figure 2) shows that both 151
 RMSE fields are of the same order of magnitude, with the 152
 ERA40's errors (Figure 2a) generally higher than CPTEC's. 153

[13] Figure 3 shows latent heat RMSE maps for the two 154
 OGCM and the coupled simulations. It is noteworthy in 155
 Figure 3 that the smallest latent heat RMSE values are 156
 found for the OGCM simulation forced with ISCCP solar 157
 radiation (Figure 3b). Both OGCM ERA40 and CGCM 158
 latent heat RMSE fields present the same order of 159
 magnitude, with the exception of the larger CGCM 160
 errors over the northern subtropics and equatorial Atlantic 161
 (Figures 3a and 3c). The combination of the large evapora- 162
 tive and solar flux errors over the northern tropical 163
 Atlantic and southeastern equatorial Atlantic (Figures 3a 164
 and 3c and Figures 2a and 2b respectively) suggest that 165
 these are the main contributors to the errors in the net 166
 surface heat flux shown in Figure 1a and 1c. 167

[14] Figure 4 shows the SST RMSE for the three simu- 168
 lations performed. The net heat flux errors over the northern 169
 tropical Atlantic and eastern equatorial Atlantic causes the 170
 largest SST errors over these regions for the OGCM 171
 simulation using ERA40 fields (Figure 4a). The reduced 172
 errors in SST are noteworthy in Figure 4b, when forcing the 173

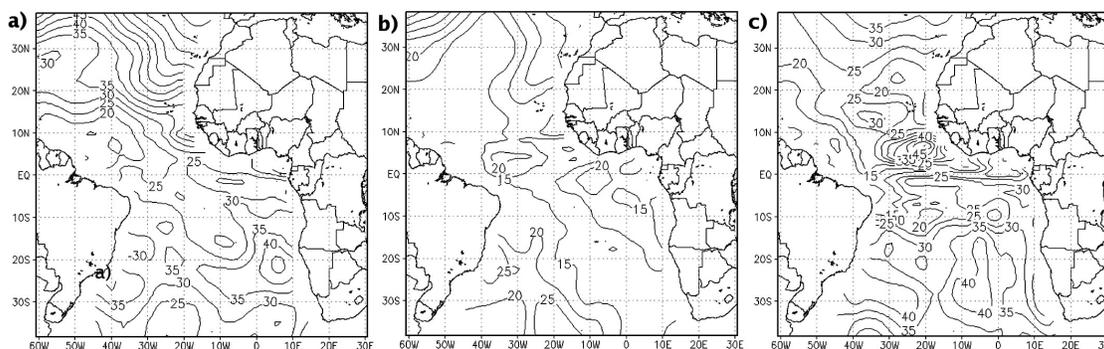


Figure 1. Net surface heat flux RMSE (Wm^{-2}): (a) ECMWF ERA40; (b) ISCCP DX; (c) CPTEC CGCM.

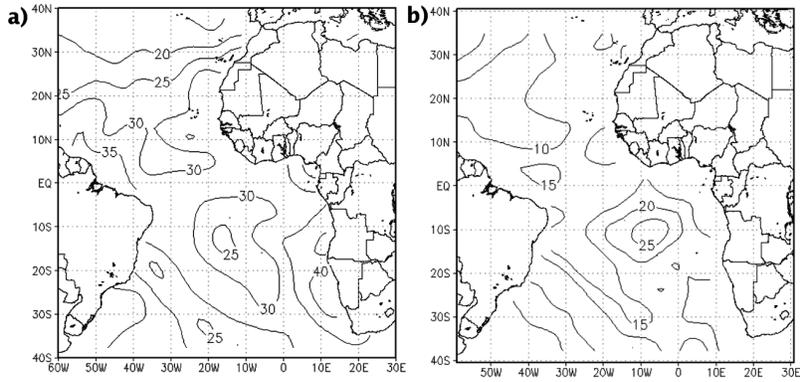


Figure 2. Surface solar radiation RMSE (Wm^{-2}): (a) ECMWF ERA40; (b) CPTEC CGCM.

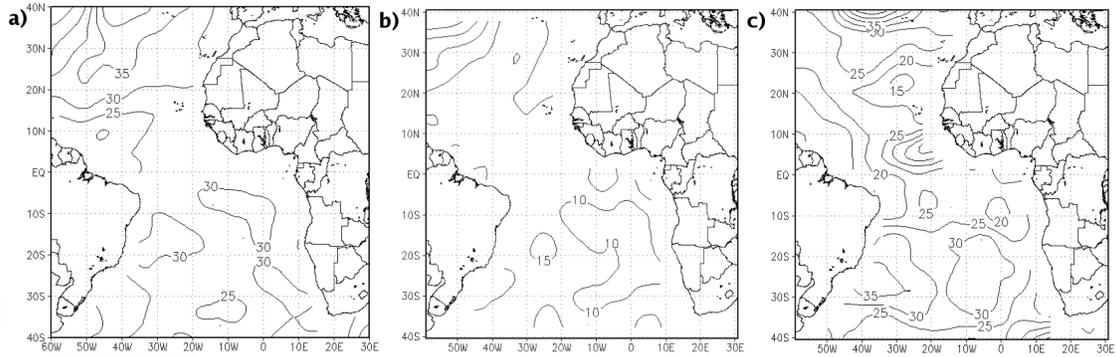


Figure 3. Surface latent heat RMSE (Wm^{-2}): (a) ECMWF ERA40; (b) ISCCP DX; (c) CPTEC CGCM.

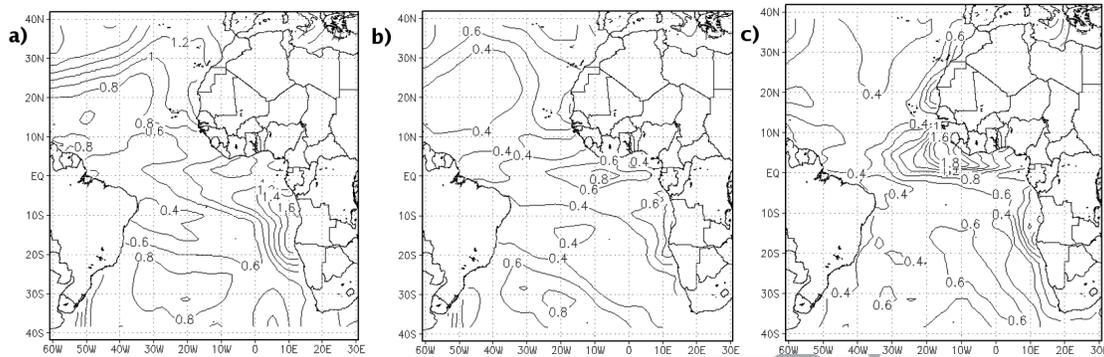


Figure 4. SST RMSE ($^{\circ}\text{C}$): (a) ECMWF ERA40; (b) ISCCP DX; (c) CPTEC CGCM.

174 OGCM with solar radiation fields computed from satellite
 175 estimates and parameterized latent, longwave, and sensible
 176 heat fluxes. This simulation shows a strong error reduction
 177 over the northern tropical and eastern equatorial Atlantic.
 178 The CPTEC CGCM (Figure 4c) presents a similar error
 179 pattern of that in the OGCM simulation forced by satellite
 180 estimates (Figure 4b), except over the central basin where
 181 the larger CGCM errors in latent heat loss (Figure 3c)
 182 contributes to the large magnitudes in the SST errors shown
 183 in Figure 4c.

184 [15] The main results of the three numerical simulations
 185 are summarized in Table 2, in the form of RMSE spatial
 186 mean over the entire tropical Atlantic. Overall, the main
 187 contributors for the net heat flux errors in the presented
 188 simulations are the shortwave and latent heat for all simu-
 189 lations (Table 2). Accordingly, OGCM simulations showed
 190 the best overall results when surface heat fluxes are param-
 191 eterized, but the solar fluxes are estimates derived from
 192 satellite IR imagery. Surprisingly, CGCM heat fluxes errors
 193 are only “marginally” larger than the ISCCP forced OGCM
 194 simulation (with the exception to the latent heat, for which
 195 the OGCM ERA40 simulation shows smaller mean error);
 196 while the ERA40 forced OGCM simulation presents the
 197 largest errors. Here, two processes might be at play; one is
 198 the expected improvement of simulated surface heat fluxes
 199 due to the presumably better estimate of shortwave solar
 200 radiation inferred from satellite IR data, as compared with
 201 the ERA40’s solar fluxes. The other is the possibility that
 202 surface momentum fluxes from the ERA40 reanalysis are
 203 worse than the CGCM stresses, thus impacting in the wind-
 204 induced evaporation and equatorial upwelling. Figure 5a
 205 shows a longitude-depth cross section of the second deriva-
 206 tive of temperature with depth along the equatorial Atlantic
 207 (as an estimate of thermocline slope and depth) for the three
 208 numerical experiments and Levitus climatology [Levitus
 209 and Boyer, 1994].

210 [16] It is remarkable to observe in Figure 5a that the
 211 CGCM thermocline is shallower in the east and presents a
 212 steeper east-west inclination than the thermoclines of both
 213 OGCM forced runs. This is an indication that the surface
 214 stress product generated by the CGCM is likely to be more
 215 energetic than the ERA40 stress products in the equatorial
 216 area, where the coupling is stronger. Such supposition is
 217 confirmed by the annual mean difference stress field shown
 218 in Figure 5b, confirming our supposition that ERA40
 219 stresses are too weak, resulting both; an excessively flat
 220 thermocline and less evaporative cooling of surface waters.

The root of such deficiencies might be in the very nature of
 two-tier approach of reanalysis. In regions like the eastern
 equatorial Atlantic, where stratus cloud decks form over
 cool waters, the reanalysis process uses observed SST, and
 generally produces subsidence that may not occur over
 these regions, increasing the solar flux and consequently,
 in our numerical experiments, the SST, which leads to
 greater surface flux errors.

[17] Heat transport mechanisms in the equatorial region,
 such as vertical entrainment, zonal and meridional heat
 advection also play an important role in the SST’s determi-
 nation. In order to access to what degree such transport
 mechanisms contribute to the mean error fields shown
 above, the zonal, meridional, and vertical heat transport
 differences between CGCM and OGCM-ERA40 simula-
 tions are computed following the heat storage rate equation
 in the work by Moisan and Niiler [1998], and shown in
 Figure 6. The examination of these components of the heat
 transport over the equatorial Atlantic reveals strong differ-
 ences between the CGCM and the OGCM simulations in
 the zonal advection and vertical entrainment (Figures 6a and
 6c). The greater magnitudes of CGCMs westward zonal
 heat advection in the central portion of the equatorial
 Atlantic together with the stronger vertical entrainment in
 the central and eastern portion indicates that the CGCM
 ocean dynamics promotes a better representation of the
 thermocline slope and depth due to the greater mixed layer
 heat loss provided by these two process.

4. Seasonal Cycle at Specific Locations

[18] To quantify the time evolution of the RMSE fields
 shown in the previous section, time series at two PIRATA
 sites are examined with respect to the seasonal evolution.
 The chosen points are at 15°N , 38°W and at 0°N , 0°E ,
 because of the differences in ocean dynamics and atmo-

Table 2. RMSE Spatial Mean: SST and Heat Flux Components

	Era40	ISCCP	CPTEC CGCM T42L28	
SST	0.89	0.52	0.53	t2.3
Net heat flux	33.81	17.53	27.90	t2.4
Short wave	31.79	13.76	19.45	t2.5
Latent heat	28.72	19.08	34.35	t2.6
Sensible heat	8.12	5.03	5.4	t2.7
Long wave	6.40	8.21	5.93	t2.8

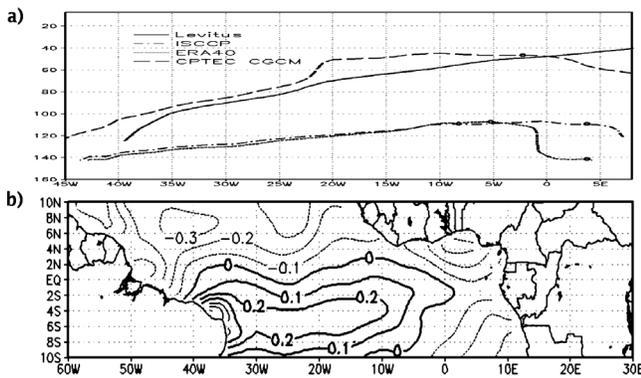


Figure 5. (a) Longitude-depth cross section of maxima temperature vertical gradient along the equator (as an indication of the positioning of the thermocline) for both OGCM forced runs (dash-dotted and dotted lines) and for the CGCM run (thick dashed line), and Levitus climatology (thick continuous line). (b) Annual mean difference CGCM – ERA40 wind stresses (dynes/cm²).

255 spheric forcing between these two locations. The seasonal
 256 evolution for monthly averages for SST, net heat, solar, and
 257 latent heat fluxes are shown in Figure 7.

258 [19] The CGCM seasonal evolution of SST at 15°N,
 259 38°W (Figure 7a) presents the best resemblance with the
 260 PIRATA observations of all simulations, albeit the general
 261 bias of all simulations (see RMSE values on the panels of
 262 Figure 7). On the other hand, the SST simulations at 0°N,
 263 0°E (Figure 7b) show a discrepant behavior related to the
 264 observations, as both the CGCM and the ocean simulation
 265 forced by reanalysis fields were unable to represent both the

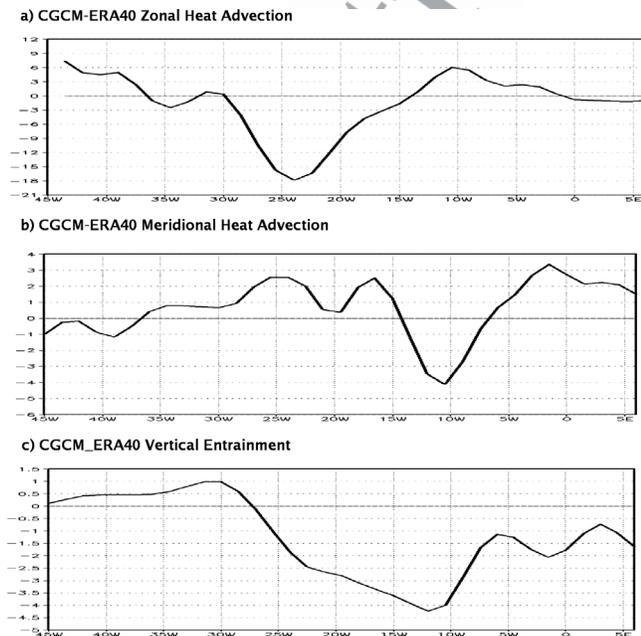


Figure 6. Longitudinal cross section at 0°N of annual mean difference: (a) CGCM-ERA40 Zonal Advection (W/m²); (b) CGCM-ERA40 Meridional Advection (W/m²); (c) CGCM-ERA40 Entrainment (W/m²).

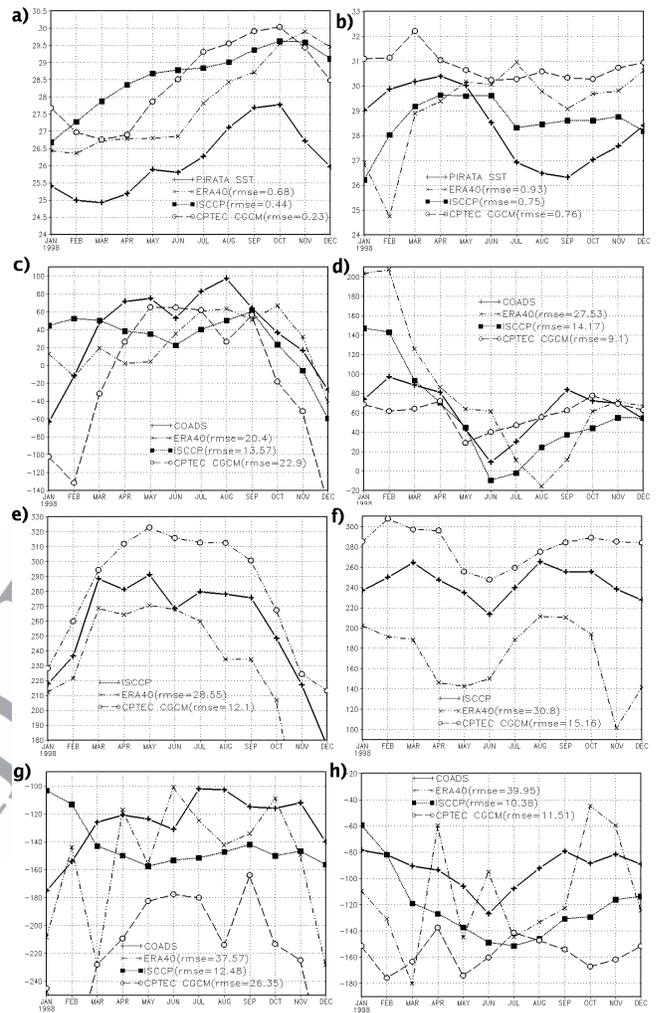


Figure 7. (a and b) Time series of SST (°C); (c and d) net heat (Wm⁻²); (e and f) solar heat (Wm⁻²); and (g and h) latent heat (Wm⁻²) at the PIRATA sites 15°N, 38°W (left column) and 0°N, 0°E (right column), respectively.

amplitude and phase of the observed SST time evolution at 266
 this site. 267

[20] The annual march of the net heat flux (Figures 7c 268
 and 7d) simulations show smaller discrepancies with 269
 observations than the simulated latent heat loss, shown in 270
 Figures 7g and 7f. The smaller discrepancies of the net heat 271
 fluxes indicate that compensation between the solar and 272
 latent heat are in place, as it can be verified by comparison 273
 of Figures 7e and 7g and Figures 7f and 7h over both 274
 PIRATA sites. The positive CGCM bias of solar heating is 275
 partially offset by the larger evaporative cooling bias. Such 276
 compensation is not so evident for the ERA40 simulation, 277
 resulting the larger discrepancies of the OGCM simulations 278
 shown in Figures 7c and 7d. 279

5. Discussion 280

[21] In this work comparisons have been made between 281
 two surface solar radiation products with in situ measure- 282
 ments of the PIRATA buoys, with the purpose to analyze the 283
 impact of solar radiation fluxes estimated by different 284

- 285 methods and heat flux parameterization in determining SST
 286 variations in the tropical Atlantic.
- 287 [22] The radiative transfer model (ISCCP DX - NOAA/
 288 NASA PATHFINDER) has shown higher correlation with
 289 the buoys data than ECMWF ERA40 fields. The differences
 290 are larger in regions where the numerical prediction models
 291 shows difficulties in simulating the appearance of stratus
 292 clouds over cold waters such as the eastern equatorial
 293 Atlantic.
- 294 [23] Two oceanic simulations forced with estimates of
 295 solar heat and momentum fluxes and a coupled ocean-
 296 atmosphere simulation were done. Based on the simulations
 297 results with different solar radiation inputs and heat flux
 298 parameterization, significant differences in SST and heat
 299 flux fields were detected suggesting that solar heat flux is of
 300 primordial importance to reduce SST errors on forced model
 301 simulations.
- 302 [24] The use of solar radiation fields based on satellite
 303 estimates and parameterized heat flux generated the best
 304 SST and surface heat fluxes simulations. The CGCM SST
 305 simulations were second best, due in part to latent and solar
 306 heat fluxes bias compensation, and in part to its better
 307 oceanic thermal structure. The examination of the oceanic
 308 heat transport over the equatorial Atlantic revealed strong
 309 differences between the CGCM and the OGCM forced runs.
 310 The surface stress generated by the CGCM has shown to be
 311 more energetic in the equatorial area than the ERA40
 312 reanalyses. The stronger surface stresses generated by the
 313 CGCM contributed to generating an oceanic thermal struc-
 314 ture in closer agreement with observations, thus suggesting
 315 the importance of the wind stress quality to correctly
 316 simulate oceanic advection and evaporative processes.
- 317 [25] The validation of the model results still requires
 318 systematic comparisons for longer periods of time. To
 validate the CGCM results against observation is a neces- 319
 sary task, and it is part of our current research undertakings. 320
- [26] **Acknowledgments.** The authors acknowledge R. Pinker for the 321
 solar radiation data and Marta Malagutti and Emanuel Giarolla for their 322
 help with the modelling work. This work is a result of the first author's 323
 Master Thesis at INPE, and was partially funded by a grant from CAPES 324
 and CNPq. 325
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