Atmosphere Feedbacks during ENSO in a Coupled GCM with a Modified Atmospheric Convection Scheme

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ABSTRACT

The too diverse representation of ENSO in a coupled GCM limits one’s ability to describe future change of its properties. Several studies pointed to the key role of atmosphere feedbacks in contributing to this diversity. These feedbacks are analyzed here in two simulations of a coupled GCM that differ only by the parameterization of deep atmospheric convection and the associated clouds. Using the Kerry–Emanuel (KE) scheme in the L’Institut Pierre-Simon Laplace Coupled Model, version 4 (IPSL CM4; KE simulation), ENSO has about the right amplitude, whereas it is almost suppressed when using the Tiedke (TI) scheme. Quantifying both the dynamical Bjerknes feedback and the heat flux feedback in KE, TI, and the corresponding Atmospheric Model Intercomparison Project (AMIP) atmosphere-only simulations, it is shown that the suppression of ENSO in TI is due to a doubling of the damping via heat flux feedback. Because the Bjerknes positive feedback is weak in both simulations, the KE simulation exhibits the right ENSO amplitude owing to an error compensation between a too weak heat flux feedback and a too weak Bjerknes feedback. In TI, the heat flux feedback strength is closer to estimates from observations and reanalysis, leading to ENSO suppression. The shortwave heat flux and, to a lesser extent, the latent heat flux feedbacks are the dominant contributors to the change between TI and KE. The shortwave heat flux feedback differences are traced back to a modified distribution of the large-scale regimes of deep convection (negative feedback) and subsidence (positive feedback) in the east Pacific. These are further associated with the model systematic errors. It is argued that a systematic and detailed evaluation of atmosphere feedbacks during ENSO is a necessary step to fully understand its simulation in coupled GCMs.

1. Introduction

ENSO is the strongest mode of interannual climate variability with significant impacts both in the tropics and at higher latitudes. Because it involves many different feedbacks (Wang and Picaut 2004), understanding and predicting both ENSO occurrence and amplitude is still a scientific challenge (McPhaden et al. 2006). For instance, the Intergovernmental Panel on Climate Change Fourth Assessment Report (IPCC AR4) sce-

narios do not show any clear evolution of ENSO properties in a warmer climate (Meehl et al. 2007; Guilyardi et al. 2009). Part of this uncertainty is due to model systematic errors. Indeed, and even though much progress has been achieved in the last decade, coupled ocean–atmosphere GCMs still have errors in reproducing the observed characteristics of El Niño events (AchutaRao and Sperber 2006; van Oldenborgh et al. 2005; Guilyardi 2006; Capotondi et al. 2006). Improving the next generation of state-of-the-art IPCC-class models is therefore a prerequisite to reliable projections of the likely evolution of ENSO in the future.

Several studies have pointed out the dominant role of the atmospheric component in setting ENSO
characteristics in state-of-the-art models (Schneider 2002; Guilyardi et al. 2004; Kim et al. 2008; Neale et al. 2008; Wu et al. 2007; Sun et al. 2008). It is nevertheless unclear why, and further analysis of the ocean–atmosphere feedbacks is a key to understanding and eventually correcting ENSO biases. Even though such an analysis is complicated by the fact that the origins of the biases are usually model dependent, one can nevertheless derive diagnostics to assess in detail the role of atmosphere processes during ENSO and compare them with observations, when available. Together with the estimation of ocean processes (Fedorov 2007; Dewitte et al. 2007; Jin et al. 2006), one should then be able to fully understand the specific ENSO characteristics of any particular coupled GCM (Guilyardi et al. 2009).

The atmospheric feedbacks during ENSO are of two types (Battisti and Hirst 1989; Jin et al. 2006; Lin 2007). First, the dynamical feedback (μ, or coupling strength), also called Bjerknes feedback (Bjerknes 1969; Neelin and Dijkstra 1995), is the intensity of the wind stress response to a given near-equatorial/tropical sea surface temperature (SST) anomaly; it is a positive feedback and maintains a large-scale east–west asymmetry across the equatorial Pacific. Second, the heat flux feedback (α) usually represents a damping of SST anomalies by both radiative and turbulent heat fluxes (Waliser et al. 1994; Jin et al. 2006). At interannual time scales, the two dominant terms in the tropical Pacific are the shortwave heat flux (SHF) feedback (α_{SW}) (see Bony et al. 1997) and the latent heat flux (LHF) feedback (α_{LH}). Indeed, in this region, the longwave heat flux is mostly related to the SST and changes little and the sensible heat flux is weak as the near-surface air temperature is close to the SST. The mean annual α is a negative feedback over most of the tropics, although its components can in time become positive. For instance, α_{SW} is negative over the warm pool where deep convection prevails (higher SST lead to increased deep convection, hence increased cloudiness and reduced shortwave heat flux, Ramanathan and Collins 1991), but positive over the cool tongue where higher SST reduce the stability of the boundary layer and thus the marine stratiform clouds amount (Wallace et al. 1989; Klein and Hartmann 1993; Norris and Leovy 1994; Li and Philander 1996; Li 1997b). The SST–SHF (LHF) relation described by these feedbacks encompasses both local (i.e., pointwise) thermodynamical mechanisms and remote processes (Stephens 2005). For instance, several studies have pointed to the key role of the large-scale atmospheric circulation in these feedbacks and, in particular, the vertical motions (Fu et al. 1994; Lau et al. 1997; Bony et al. 1997, 2004). Heat flux feedbacks have also been shown to have a key role in the tropical Atlantic Ocean, explaining a large fraction of model errors and diversity (Frankignoul et al. 2004).

The high sensitivity of the modeled atmosphere to the parameterization of deep convection has been documented by a number of studies (e.g., Zhang and McFarlane 1995; Hourdin et al. 2006; Bracnot et al. 2007, hereafter B07). The parameterization of deep convection and the associated clouds acts to mimic the way the atmosphere is destabilized by surface heating, how its properties are mixed vertically, and indirectly what surface fluxes are generated. In the real atmosphere, individual convective events can happen at very small horizontal scales (several hundred meters at most) and involve rapidly ascending and descending motions and complex thermodynamics (Emanuel 1994). These events then become organized into clusters at larger scale. The challenge of the different convection schemes devised over the years has been to represent the mean effect of these convective events over grid cells of several hundreds of kilometers in size.

The deep convection parameterization, which interacts with the clouds scheme, the boundary layer, and the dynamics, is of central importance in modeling of the tropical atmosphere. Indeed, deep convection is the primary heat source driving the large-scale circulation, affects wave disturbances through the release of latent heat, and drives the vertical redistribution of heat, moisture, and momentum. The representation of deep convection is therefore central in defining both the dynamical and the heat flux atmosphere feedbacks during ENSO. Several recent studies have documented the impact of a modified convection scheme on ENSO in coupled GCMs. Kim et al. (2008) and Neale et al. (2008) have shown that including convective momentum transport (CMT) redistributes momentum vertically toward the surface during ENSO, shifting mean and interannual structures back toward the east. In the model used by Kim et al., the ENSO amplitude strongly increases when the CMT is added, owing to a 50% increase of the Bjerknes feedback μ. In Neale et al. (2008) and Wu et al. (2007), adding the CMT leads to striking improvements in ENSO simulation and alleviates the strong 2-yr period ENSO in the control simulation. Neale et al. (2008) argue that the improvements come from a stronger role of intraseasonal variability and a redistribution of zonal momentum that leads both to increased μ and α. Even with the same atmospheric physics, deep convection can significantly affect the evolution and termination of ENSO in the east Pacific via local feedbacks (Zhang and McPhaden 2008; Lengaigne and Vecchi 2009).

Hence, the impacts of the atmosphere deep convection on ENSO can be due to changes in the background
state, changes in the intraseasonal variability, and change in local thermodynamic feedbacks—all three aspects playing a key role in ENSO. For instance, adding the CMT acts to reduce wind stress easterly biases, which were shown to be related to too weak ENSO amplitude (Guilyardi 2006). The change in the simulation of intraseasonal variability (Neale et al. 2008; Wu et al. 2007) can lead to changes in the triggering and amplification of ENSO as shown by Lengaigne et al. (2004).

Here we analyze the different atmospheric ENSO feedbacks in two simulations of the L’Institut Pierre-Simon Laplace (IPSL) Coupled Model, version 4–(CM4) model that differ by the deep convection and the associated cloud scheme used: the standard IPCC version with the Kerry–Emanuel (KE) scheme and the Tiedke (TI) scheme, the rest of the atmosphere and the ocean being exactly the same. The Atmospheric Model Intercomparison Project (AMIP) atmosphere-only simulations were described by Hourdin et al. (2006), and the mean state and seasonal cycle of the coupled simulations were described by B07. The striking feature of these simulations is that, while the control KE simulation does exhibit ENSO with about the right amplitude, ENSO in the TI simulation is almost suppressed (Fig. 1). We show that this suppression is largely due to modified atmospheric feedbacks and investigate the reasons why.

Section 2 describes the model, the KE and TI simulations, and the reference observation used. Section 3 evaluates the ENSO properties of the two simulations, and section 4 investigates the details of the different atmospheric feedbacks and unravels the reasons for ENSO suppression in the TI simulation. Section 5 identifies the model systematic errors responsible for the atmosphere feedback deficiencies. A summary and a discussion are presented in section 6.

2. Model and simulations

   a. Model and simulations

   The model used in this study is the IPSL CM4 model (Marti et al. 2009), which couples the Laboratoire de Météorologie Dynamique (LMDz) AGCM (Hourdin et al. 2006) and the Océan Parallélisé (OPA) OGCM (Madec et al. 1998) via the Ocean–Atmosphere–Sea Ice–Soil (OASIS) coupler (Valcke et al. 2000). The standard configuration, used for the IPCC AR4 (Solomon et al. 2007), uses the Kerry–Emanuel convection scheme and the associated cloud scheme in LMDz. A sensitivity simulation using the Tiedke convection scheme and the associated cloud scheme is also made. The KE scheme gives a unique treatment of both deep and shallow convection and has a more complex representation of

![Figure 1](https://example.com/figure1.png)

**Fig. 1.** Monthly SST anomaly standard deviation for (a) 1900–99 HadISST1.1 and 100 years of IPSL CM4 with the (b) Kerry–Emanuel convection scheme, (c) the Tiedke convection scheme, and (d) the KE convection scheme using an interannual SW heat flux correction (KE-mod); see text. Contours every 0.2°C; Niño-3 (to the east) and Niño-4 (to the west) regions are indicated.
downdrafts, which helps to limit an otherwise overactive convection, especially over the oceans. The representation of clouds associated with cumulus convection differs between KE and TI. In KE it is coupled to the deep convection scheme by using the prognostic in-cloud water content and the degree of saturation of the large-scale environment (Bony and Emanuel 2001), whereas in TI an assumed homogenous cloud fraction is a function of the vertically integrated moistening tendency predicted by the Tiedke convection scheme (Hourdin et al. 2006). In TI, the radiative effect of convective clouds is also treated differently. A more complete description of these schemes can be found in Hourdin et al. (2006) (sections 2.3 and 2.4) and B07 (section 2b). The only adjustment made in TI (when compared to KE) affected the radiation scheme [so as to have a balanced top of atmosphere (TOA) heat flux] and the differences seen in TI also existed in a nonadjusted simulation. The AMIP atmosphere-only simulations (AMIP-KE and AMIP-TI) span the 1979–2002 period (which encompass the 1982–83 and 1997–98 El Niño events) and are described in Hourdin et al. The setup, the mean state, and seasonal cycle of the KE and TI simulations, of which 100 years are used here, are described for the tropical Pacific in B07. It is worth noting that distinguishing the detailed relative impacts of both changes (convection scheme versus cumulus cloud scheme) is not possible since the convective scheme change will also modify the latent heating and the vertical distribution of moisture, hence the cumulus cloud properties. Nevertheless, the sensitivity experiments made in AMIP mode by Hourdin et al. suggest that the change of convection scheme has a dominant effect.

b. Validation datasets

Several datasets are used to evaluate the performance of the model. In all cases, monthly means are used. SSTs are compared to the Met Office’s Hadley Center’s Sea Ice and SST dataset version 1 (HadISST1) (Rayner et al. 2003) and the Reynolds (1988) dataset used for the National Centers for Environmental Prediction (NCEP) reanalysis. Precipitations are compared to the Climate Prediction Center’s Merged Analysis of Precipitation (CMAP) (Xie and Arkin 1997). Wind stress comes from the 40-yr European Centre for Medium-Range Weather Forecasts Re-Analysis (ERA-40) (Uppala et al. 2005), NCEP/Department of Energy Global Reanalysis 2 (NCEP-2) (Kalnay et al. 1996) and a combination of European Remote Sensing Satellite (ERS) and TAO (Menkes et al. 1998). Ocean surface heat fluxes are compared to ERA-40, the International Satellite Cloud Climatology Project (ISCCP) (Zhang et al. 2004) and objectively analyzed air–sea fluxes (OAFlux) (Yu et al. 2007), and cloud cover from ISCCP (Rossow and Schiffer 1991). The three-dimensional structure of the atmosphere is compared to ERA-40 and NCEP-2. In the ocean, two reference-forced OGCMs are used. They are made from the same OGCM [Nucleus for European Modelling of the Ocean (NEMO)–OPA] but at higher resolution (0.5° instead of 2°–0.5° in KE/TI) and use two sets of Coordinated Ocean and Sea Ice Reference Experiment (CORE) (Griffies et al. 2009) forcing: one based on NCEP-2 and the other on ERA-40 (Barnier et al. 2006; Molines et al. 2006). In the comparison, and whenever possible, two references are used to illustrate the observations/reanalysis uncertainty, keeping in mind that this only provides a lower bound to the uncertainty.

c. Mean state and seasonal cycle in the tropical Pacific

ENSO is defined as an anomaly to the mean seasonal cycle. It is therefore key to describe this seasonal cycle, especially in coupled models where it can exhibit quite different behaviors in the tropical areas (Covey et al. 2000; Guilyardi 2006; Wu et al. 2008). Moreover, errors in ENSO simulations can often be traced back to errors in the background state, as will be shown in section 5.

In the tropical Pacific, B07 described significant changes in the seasonal cycle between KE and TI. We detail here the aspects relevant to ENSO mechanisms. Figure 2 presents the mean seasonal cycle at the equator in the Pacific for the zonal wind stress (shading), the SST (contour), and the 3 and 8 mm day\(^{-1}\) precipitation contours. In the observations and reanalysis (Figs. 2a and 2b), the trade winds (easterlies) are strong during Northern Hemisphere (NH) summer and autumn and are associated with the upwelling of colder SST in the east Pacific via strong coupling (Bjerknes feedback, \(\mu\)). In NH spring, the coupled system relaxes, the trade winds weaken, and the SST in the east Pacific warms up. The bulk of the atmospheric deep convection (as delimited by the 3 mm day\(^{-1}\) precipitation contours) stays west of the date line, and the seasonal cycle in the west Pacific is weak.

KE exhibits a quite well-simulated seasonal cycle in the equatorial Pacific (Fig. 2e), especially when compared to other IPCC-class models (Guilyardi 2006). In the east Pacific, the upwelling is timed right (even though it does not last long enough) and the NH spring relaxation is present. In the west Pacific, the annual cycle is weak, like in observations, and the deep convection mostly stays west of the date line. A common bias of many coupled GCMs is that the patterns are shifted westward: the upwelling does not reach the South American coast and the easterlies extend too far into the warm pool, especially during December–February (DJF), when westerlies prevail there in observations.
The feedbacks between convection, large-scale circulation, SST, and clouds maintain the ITCZ in a realistic position in KE (not shown, see B07). As already pointed out by B07, the origin of these qualities and systematic errors can partly be traced back to the atmospheric GCM (see also Lin 2007; Sun et al. 2008). When forced by observed SST, AMIP-KE produces easterlies and precipitation that broadly agree
with observations (Fig. 2c). Nevertheless, the easterlies are too strong in NH winter/spring and the structures are shifted westward by about 10°–20°, especially over the warm pool. These biases are then amplified in the coupled model (Fig. 2e).

The seasonal cycle in the equatorial Pacific is degraded in TI when compared to KE (Fig. 2f). In the east, this involves a weaker annual cycle, a stronger semiannual cycle (an additional spurious relaxation of the easterlies is seen in the NH autumn), and reduced upwelling. In the west Pacific, there is a much stronger annual cycle, and the intensification of the easterlies in NH winter (weakly present in KE) is dramatically increased here and reaches the amplitude of the NH summer east Pacific easterlies. The deep convection spuriously migrates east of the date line and reached the South American coast during the two easterly relaxation periods (April–June and November–December). Accordingly, the mean zonal SST gradient across the equatorial Pacific is reduced by 1.4°C in TI when compared to KE. Meridionally, the ITCZ is too far south, close to the equator, and the spurious semiannual cycle is associated with a strong spurious double ITCZ (B07). Again, these biases can be traced back to the AMIP-TI simulation (Fig. 2d). Even forced by observed SST, the spurious secondary maximum of easterlies is seen extending west of the date line from October to March. The deep convection also moves much too far to the east and reaches the South American coast during the NH spring relaxation.

B07 attributes the east Pacific biases in TI to the impact of the modified convection scheme on the South American monsoon regime. In KE, the added representation of downdrafts allows a similar intensity of convection over land and over the ocean whereas, in TI, the deep convection over the ocean is much stronger than over land (Fig. 14 in B07). This results in intensified monsoon winds in the east Pacific that disrupt the seasonal cycle, especially during spring and autumn. Indeed, the equinoctial asymmetric precipitation monsoon mode (which measures the April–May minus October–November mean precipitation difference, Wang and Ding 2008) is much too strong in AMIP-TI when compared to AMIP-KE, whereas the solstitial mode (measuring the winter minus summer precipitation difference) is not changed much (not shown). These biases during the equinoxes modify the global circulation of the tropical Pacific and contribute to the basinwide biases seen both in AMIP-TI and TI during these periods.

In the ocean, the depth of the mean thermocline, as measured by the zonal 20°C isotherm depth, is not changed much between KE and TI (Fig. 3). The near-equatorial thermocline is slightly deeper for TI (about 10 m), which will somewhat weaken ENSO-related thermocline feedback (Li 1997b) but not to the point of suppressing ENSO (Fig. 1). As a comparison, the mean thermocline depth of two forced ocean models differing by the wind product used (ERA-40 and NCEP) shows similar differences.

### 3. Simulated ENSO properties

As already seen in Fig. 1, the ENSO amplitude is diminished by a factor of ~3 from KE to TI [the Niño-3 (5°S–5°N, 150°–90°W) SST anomaly (SSTA) standard deviation has decreased from 1° to 0.34°C]. To understand the mechanisms that led to such a dramatic change, El Niño composites are devised by using events for which the SSTA in Niño-3 is larger than 1.5 standard deviation (four events in HadISST1.1, five in KE, and nine in TI; Fig. 4). This criterion only picks up the large El Niño events; this is justified as the change in ENSO properties between KE and TI is very large. It was checked that relaxing the threshold to 1.3 standard deviations does not change the results discussed below. The El Niño composite for KE exhibits a correct amplitude and time evolution when compared to observations, even though the amplification phase in the NH spring of year 0 is stronger than in observations and the decay also starts earlier. In TI, the amplitude is very weak, as expected from Fig. 1. For both simulations, the interevent differences are small enough so that the composite is a good representation of each event (not shown). Composites are also devised for the AMIP simulations, which include the 1982–83 and 1997–98 El Niño events, and for the different reference data. Both atmospheric feedbacks (dynamical and heat flux) are then analyzed.
The equatorial anomalies of the composite zonal wind stress $\tau_x$ (shading) and SST (contours) in the Pacific are shown in Fig. 5. Here, reference fields include both the ERA-40 and the NCEP reanalysis (Figs. 5a and 5b). In both cases, the westerly wind stress anomaly that propagates eastward in year 0 is associated with the El Niño warming through the Gill (1980) atmosphere response and via the Bjerknes feedback. Both AMIP-KE (Fig. 5c) and AMIP-TI (Fig. 5d) exhibit a similar wind stress response to the forced SST. The magnitude of the response is slightly larger in KE (the next section will quantify these differences), and both AMIP simulations show an increase of easterlies in the east of the basin, associated with the westward shift of the wind structure (Figs. 2c and 2d). In KE, the zonal wind stress also exhibits a westerly anomaly associated with weakened easterlies, albeit much weaker than in AMIP-KE (Fig. 5e). The SST anomaly structure is shifted westward, as also seen in Fig. 1. Unlike the observations and the AMIP simulations, the eastward propagation of the wind stress anomaly envelope is not very clear. This may be partly due to the dominance of the SST mode in this model, whereas observations (especially the 1982–83 and the 1997–98 events) also exhibit a thermocline mode, which involves eastward propagation (Trenberth and Stepaniak 2001; Guilyardi 2006; Lengaigne and Vecchi 2009). In TI, the interannual anomalies are quite small since ENSO is mostly suppressed.

To describe the heat flux feedback, the equatorial anomalies of the composite total downward surface heat flux (shading) and SST (contours) in the Pacific are shown in Fig. 6. Here reference fields include both ERA-40 and OAFlux (Figs. 6a and 6b; the radiative components of the OAFlux dataset are taken from ISCCP). Even though surface fluxes in the different analysis available exhibits marked differences (Wittenberg et al. 2006), both products show strong heat flux damping during El Niño, mostly collocated with positive SST anomalies. The amplitude is weaker for the ERA-40 composite, which may be due to the larger number of El Niño events included. For the OAFlux composite, which only includes the 1997–98 event, the heat flux anomaly reaches $-80$ to $-100$ W m$^{-2}$ in large areas of the central eastern Pacific in January–February of year +1. In the AMIP-KE composite, the heat flux feedback reaches $-60$ W m$^{-2}$ in the central Pacific, whereas it reaches $-100$ to $-120$ W m$^{-2}$ in AMIP-TI. In KE, the heat flux anomaly is not collocated with the SST anomaly but on the SST zonal gradient near the date line and is also weaker, reaching only $-40$ W m$^{-2}$ (Fig. 6e), indicating a different balance of physical mechanisms east of the date line. In TI, the interannual anomalies are also quite small since ENSO is mostly suppressed even though they are collocated with the SST anomalies.

4. Mechanisms for ENSO suppression in TI

The analysis presented so far strongly suggests that the increased heat flux feedback in TI is responsible for the ENSO suppression in this simulation. To verify this hypothesis, the Bjerknes (BJ) index and its components (Jin et al. 2006) are computed for KE and TI (Table 1). The BJ index measures the ENSO linear growth rate based on the recharged oscillator theory (Jin...
The strongly negative BJ index for TI confirms that the balances of processes damp the development of ENSO in this simulation. All components of the BJ index contribute to the overall weakening in TI when compared to KE (Table 1). Nevertheless, the largest contribution to the decrease is by far the negative feedback due to the thermodynamic damping ($\alpha$), which is multiplied by 3, from $-0.45\text{ yr}^{-1}$ to $-1.33\text{ yr}^{-1}$. The ocean feedbacks (including the zonal advective feedback, the thermocline feedback, and the Ekman feedback; for details see Jin et al. 2006; P. Kim and F.-F. Jin (2009, personal communication) as well as the dynamical
damping (associated with the background state) also play a role, but of lesser amplitude. Most of the ocean feedback change is due to the weakening of the thermocline positive feedback, associated with the deeper mean thermocline (Fig. 3).

A rapid heat budget analysis further shows that the AMIP-KE/AMIP-TI heat flux difference of $-50 \text{ W m}^{-2}$ (Figs. 6c and 6d) over a 50-m-thick mixed layer (typical of the central/east Pacific) modifies the SST tendency by $-1^\circ\text{C month}^{-1}$ (equivalent to an additional damping
rate of (6 weeks)$^{-1}$; see also Burgers and van Oldenborgh 2003), a value large enough to cancel the development of ENSO, as suggested by the BJ index analysis above.

a. Heat flux feedback components

To understand the detail of these heat flux feedback changes, the components of the surface heat flux are now analyzed. Figures 7 and 8 show the composite evolution at the equator in the Pacific Ocean of the SHF (and total cloud cover in contours) and the LHF (and the meridional wind stress in contours), both of which are the dominant contributors to the total heat flux changes in the tropics. In ERA-40 and ISCCP (Figs. 7a and 7b), the SHF reduction reaches about $-60 \text{ W m}^{-2}$ and is associated with the increased cloud cover due to the eastern migration of the deep atmosphere convection over warm waters. This feedback is mostly occurring at the end of the warming episode due to the time required to build up a large enough SST to trigger convection. In AMIP-KE, the associated negative shortwave feedback is weaker (in an integral sense, that is, both the maximum and the extension are weaker), and becomes positive in the eastern part of the basin where heating destabilizes the boundary layer and reduces the stratiform cloud cover (Norris and Leovy 1994; Li and...
Philander 1996). In AMIP-TI the negative SHF feed-
back is much stronger since the convection in TI is also
stronger, leading to a much increased cloud cover for the
same SST anomaly (Fig. 7d, see also B07).

The latent heat flux feedback in ERA-40, and to
a lesser degree in OAFlux (Figs. 8a and 8b), exhibits
a zonal dipole with reduced latent heat flux loss (hence
the positive anomalous LHF) in the central and west
Pacific and increased latent heat flux east of 150°W. The
reduction in the central and west Pacific is due to the
relaxation of the trade winds that occurs during ENSO
in this region. In the east Pacific, the increase is due to
the increased meridional wind that dominates the mean wind speed in this region (Chelton et al. 2001). These features are well simulated by both
AMIP simulations, and the positive anomalous LHF is
of comparable intensity in the central and west Pacific.

On the other hand, the AMIP-TI feedback is larger in
the east Pacific (+20 W m$^{-2}$ in the Niño-3 region). The
increased meridional wind change in AMIP-TI is re-
sponsible for this change (Fig. 8d). This feature is asso-
ciated with the convergence due to increased convection
in the ITCZ in AMIP-TI and the too southerly location of
the ITCZ, reaching the equatorial waveguide (see B07).
Hence, both the SHF and LHF feedback in AMIP-TI
are stronger deep convection over the equatorial ocean compared to AMIP-KE.

b. Quantification of atmospheric feedbacks

To further quantify the dynamical ($\mu$) and heat flux
feedback ($\alpha$) in the east Pacific, the corresponding in-
terannual anomalies are regressed onto the Niño-3 SST
anomalies (Figs. 9 and 10). The slope of the linear re-
gression measures the intensity of the feedback. For $\mu$
[measured here as the Niño-4 zonal wind stress anomaly]
regressed onto the Niño-3 SST anomaly, both regions being the maximum variability zone for the corresponding fields, according to Gill (1980); not shown], the reference values are $8.6 \times 10^{-3}$ N m$^{-2}$ C$^{-1}$ for NCEP-2 and $12.8 \times 10^{-3}$ N m$^{-2}$ C$^{-1}$ for ERA-40 (Figs. 9a and 9b). Both AMIP simulations have values within this range, as suggested by Fig. 5, with the AMIP-TI being somewhat weaker (Figs. 9c and 9d). With a value of $4.6 \times 10^{-3}$ N m$^{-2}$ C$^{-1}$, the KE simulation (Fig. 9e) exhibits a much weaker dynamical feedback than its AMIP counterpart. Similarly the TI simulation has a much weaker dynamical feedback, although the range of SST anomalies is much reduced due to the suppressed ENSO (Fig. 9f).

The average total heat flux feedback ($\alpha$) in the Niño-3 region is $-19$ W m$^{-2}$ C$^{-1}$ in ERA-40 and $-17$ W m$^{-2}$ C$^{-1}$ in OAFX (Figs. 10a and 10b). Note that this value drops to about $-10$ W m$^{-2}$ C$^{-1}$ for

**Fig. 9.** Scatterplot of monthly Niño-4 zonal wind stress anomaly as a function of monthly Niño-3 SSTA for (a) NCEP2, (b) ERA40, (c) AMIP-KE, (d) AMIP-TI, (e) KE, and (f) TI. The slope of the linear regression measures the coupling strength $\mu$, as indicated in units of $10^{-3}$ N m$^{-2}$ C$^{-1}$. Colors indicate different seasons: Dec–Feb (black), Mar–May (blue), Jun–Aug (red), and Sep–Nov (green).
NCEP-2 (not shown), whose surface heat fluxes are subject to larger errors in the east Pacific (Cronin et al. 2006). In AMIP-KE $\alpha$ is only $-10$ W m$^{-2}$ C$^{-1}$, whereas it reaches the reference value of $-18$ W m$^{-2}$ C$^{-1}$ in AMIP-TI, as also suggested by Fig. 6d. Again, $\alpha$ in KE is much reduced ($-4.9$ W m$^{-2}$ C$^{-1}$; Fig. 10e) when compared to AMIP-KE. On the other hand, TI exhibits an $\alpha$ similar to AMIP-TI ($-20$ W m$^{-2}$ C$^{-1}$), although still with a much reduced range of SST anomalies (Fig. 10f). As discussed in the introduction, the mean resulting $\alpha$ in the east Pacific is a mix of several regimes (in which $\alpha$ varies and even changes sign), as suggested from the nonlinearity of the SHF–SST anomalies relation (Fig. 10). This aspect will be investigated in more detail in the next section.

Even though most theoretical studies use an annual mean value for $\mu$ and $\alpha$, these exhibit a marked seasonal cycle (Fig. 11). In both reanalyses, the positive dynamical feedback is strongest from July onward, when ENSO is amplified via the Bjerknes feedback, and peaks during October–November just before ENSO itself peaks

Fig. 10. As in Fig. 9 but for the monthly Niño-3 total heat flux anomaly as a function of the Niño-3 SST anomaly. The slope of the linear regression measures the heat flux feedback $\alpha$, as indicated in units of W m$^{-2}$ C$^{-1}$. (a) ERA40, (b) OAFlux/HadISST1.1, (c) AMIP-KE, (d) AMIP-TI, (e) KE, and (f) TI.
On the other hand, the heat flux feedback is strongest from about January to May (when its value doubles from \(215 \text{ W m}^{-2} \text{ C}^{-1}\) during the rest of the year to \(330 \text{ W m}^{-2} \text{ C}^{-1}\), when ENSO anomalies are reduced and/or ENSO is not yet developing. This seasonal evolution of the atmosphere feedbacks can therefore partly explain the ENSO seasonal phase lock and suggests a physical explanation to the results obtained from linear stability analysis in which ENSO is damped in NH winter and spring, whereas it is amplified during the rest of the year (Li 1997a; Li and Hogan 1999; Fedorov and Philander 2001).

In the AMIP simulations, \(\mu\) also exhibits a maximum in NH autumn with AMIP-KE having about the correct magnitude then and AMIP-TI having too weak values, especially in NH spring (Fig. 11a). The dynamical feedback in both models is reduced in coupled mode, and the seasonality disappears. Nevertheless and as described before, the intensity of the dynamical feedback is similar with both convection schemes, and \(\mu\) changes cannot explain the drastic change of ENSO behavior seen in TI. The heat flux feedback also has the correct seasonal behavior in the AMIP simulations (Fig. 11b).

Nevertheless, while the AMIP-TI follows the reanalysis quite well, the AMIP-KE \(\alpha\) is too weak by 50% year round. This bias is strongly amplified in KE, whereas \(\alpha\) in TI remains closer to the reanalysis value, albeit with a loss of seasonality. This loss of seasonality also results in a stronger negative feedback in NH summer and autumn, which further damps ENSO development. The analysis of the SHF and LHF components of \(\alpha\) (Fig. 12) shows that the difference between AMIP-KE and AMIP-TI is mostly due to \(\alpha_{\text{SW}}\), although \(\alpha_{\text{LH}}\) contributes as well to the difference in NH spring when \(\alpha\) is strongest. In coupled mode, \(\alpha_{\text{SW}}\) accounts for all the difference between KE and TI (Figs. 12a and 12b).

To verify that changes in \(\alpha_{\text{SW}}\) can indeed explain the large change of ENSO amplitude between TI and KE, a sensitivity simulation is made with the KE atmospheric physics. In this simulation, called KE-mod, an interannual flux correction is performed to artificially change \(\alpha_{\text{SW}}\) to a value close to that of TI (\(-15 \text{ W m}^{-2} \text{ C}^{-1}\), see green dashed curve in Fig. 12a). The interannual flux correction is imposed throughout the tropics between 30°S and 30°N, using the following relation:

\[
\text{SHF}_O = \text{SHF}_O^{\text{SCref}} + \alpha_{\text{SWmod}}(T_O - T_O^{\text{SCref}}),
\]

where \(\text{SHF}_O\) is the SHF received by the ocean in KE-mod, \(\text{SHF}_O^{\text{SCref}}\) is the mean seasonal cycle of SHF in the KE simulation, \(\alpha_{\text{SWmod}}\) is the modified \(\alpha_{\text{SW}}\), \(T_o\) is the SST in KE-mod, and \(T_O^{\text{SCref}}\) is the mean seasonal cycle of SST in KE. The mean state and seasonal cycle of KE-mod are very close to that of KE (not shown), as can be expected from the fact that the time integral of both SHF are equivalent. As seen in Fig. 1d, the standard deviation of SST in KE-mod has dramatically decreased and the Niño-3 average is now 0.4°C, a value close to that of TI. The total \(\alpha\) in Niño-3 in KE-mod is now \(-21 \text{ W m}^{-2} \text{ C}^{-1}\), close to that of TI. This further confirms that the decrease of ENSO amplitude is largely due to the changes in \(\alpha_{\text{SW}}\).

In summary, the correct ENSO amplitude in KE is obtained from an error compensation: it results from a too weak dynamical positive feedback associated with a too weak heat flux negative feedback. In TI, the heat flux feedback is closer to that of the reanalysis, whereas the dynamical feedback stays too weak and results in a suppressed ENSO. The analysis presented here, in Hourdin et al. (2006), and in B07 further suggests that...
the “improved” $\alpha$ in AMIP-TI and TI is most likely due to, again, error compensation as the deep convection is much too strong and the cloud cover too extended when the Tiedke scheme is used.

5. Radiative feedbacks over the cold tongue

To further understand the change of $\alpha_{SW}$ between AMIP-KE and AMIP-TI, the radiative feedbacks in the east equatorial Pacific are now analyzed in more detail during the ENSO amplification period (NH summer–autumn), when TI exhibits the largest error in the Niño-3 $\alpha_{SW}$ (Fig. 12a). A closer examination of maps of $\alpha_{SW}$ for the ISCCP satellite observations during the July–December period shows that the two regimes discussed in the introduction coexist in the Niño-3 region (Fig. 13b): a regime of negative $\alpha_{SW}$ to the west, associated with deep convection, and a regime of positive $\alpha_{SW}$ in the east Pacific, associated with the cold tongue/stratocumulus system. ERA-40 and ISCCP exhibit marked differences in this eastern region (Figs. 13a and 13b). Cronin et al. (2006) show that ERA-40, indeed, has a too strong surface cloud forcing in the ITCZ area when compared to in situ moorings observations, whereas ISCCP is broadly in agreement. The AMIP-KE simulation (Fig. 13c) has a structure closer to that of ISCCP, whereas AMIP-TI (Fig. 13d) has a structure closer to ERA-40, with too strong negative feedback over the ITCZ, and extending toward the equator (this may be due to the AGCM used in ERA-40 also having a Tiedke-type convection scheme). Nevertheless, the region of positive $\alpha_{SW}$ in AMIP-KE extends too far west and exhibits too strong values when compared to ISCCP. This explains the weaker negative feedback during ENSO amplification. In KE and TI (Figs. 13e and 13f), these biases are again amplified as the cold tongue regime (in KE) and the convective regime (in TI) take over the entire east Pacific all the way to the date line.

Next we look at the influence of the large-scale atmospheric circulation. A given SST can be associated with different regimes of vertical motions (Bony et al. 2004). It is therefore necessary to distinguish these regimes to further understand the origin of the weaker $\alpha_{SW}$ in KE. The time evolution of the anomalous SHF, the vertical velocity at 500 hPa ($v_{500}$), the SST anomaly, and the SST is shown for the far east Pacific ($5^\circ$N–$5^\circ$S, $110^\circ$–$90^\circ$W) for the El Niño composites in Fig. 14. The vertical velocity at 500 hPa is a good proxy for large-scale deep convection ($v_{500}$) and subsidence ($v_{500}$.20 hPa day$^{-1}$) regimes (Bony et al. 2004; Bony and Dufresne 2005). In ISCCP, the increase in SHF during increasing SST (hence a positive $\alpha_{SW}$) from July to December peaks at 15 W m$^{-2}$ in September–October of year 0 (black solid curve in Fig. 14a). The SHF rapidly decreases to become a negative feedback as SST reaches the deep convection threshold in November–December. As evidenced by a $\omega_{500}$ on average >20 hPa day$^{-1}$, the positive $\alpha_{SW}$ until October of year 0, indeed, occurs in a predominantly subsident regime over the cold tongue (Fig. 13b; see also Bony and Dufresne 2005): when the SST increases, the stratocumulus deck is destabilized and the SHF increases. In AMIP-KE, the SHF anomaly reaches 30 W m$^{-2}$ in December of year 0 (red solid curve in Fig. 14a) and its subsequent decrease lags that of ISCCP by about three months, allowing much more heat to enter the ocean mixed layer during this key El Niño amplification phase. In coupled mode (KE), as the cold tongue is too well developed, the deep convection regime is never reached in the east Pacific, even at the peak of El Niño ($\omega_{500}$ stays positive; see Fig. 14b).
Consequently, $\alpha_{SW}$ stays positive throughout the event (Fig. 14a), and even through the year (Fig. 12a). In sharp contrast, TI almost always exhibits a dominant deep convection regime in the far east Pacific (Figs. 14b), leading to the strongly negative $\alpha_{SW}$ seen all year round (Fig. 12a). This difference between KE and TI was also noted by B07, who show that the largest difference of surface cloud forcing occurred in the subsidence regime, which includes the eastern part of the Pacific (B07, their Fig. 13).

Several studies have shown that deep convection was roughly occurring for SSTs higher than 26°–27°C (Lau et al. 1997; Bony et al. 1997, 2004). To identify this threshold in the far east Pacific, $\omega_{500}$ is binned as a function of SST in Fig. 15. For ERA-40, the deep convection threshold is near 27.6°C, a value usually not
reached during the seasonal cycle in this region (Fig. 2a). AMIP-KE has a higher SST threshold (28.2°C) than that of AMIP-TI (27.2°C). This low value AMIP-TI explains the spurious NH spring incursion of convection in the east Pacific, noted in Fig. 2d. In coupled mode, KE maintains the 28.2°C threshold of AMIP-KE, whereas the deep convection threshold for TI becomes even lower (26.3°C, Fig. 15). When using these thresholds, in Fig. 14, one can readily explain the regime changes both in time and between the simulations and the observations. As seen from Fig. 14d, the two convective events in the ERA-40 composite (April of year 0 and January–May of year 1; Fig. 14b) correspond to times when the deep convection threshold (27.6°C in Fig. 14d) is reached. Similarly, the higher threshold for AMIP-KE when compared to AMIP-TI explains the longer subsidence regime for AMIP-KE (November–January) discussed above. Figure 14d also clearly shows that convection cannot be reached during El Niño in KE because the SST almost never reaches 28.2°C. Conversely, because of the very low threshold in TI, deep convection almost always occurs in the east Pacific in this simulation (Fig. 2f). Such a low deep convection threshold provides a very effective mechanism to prevent the development of El Niño in this simulation.

6. Summary and discussion

The atmosphere feedbacks during ENSO in two coupled GCMs that differ only by the atmosphere convection parameterization used are analyzed. Using the Kerry–Emanuel scheme (KE), ENSO has about the right amplitude; however, this is shown to be due to an error compensation between a too weak Bjerknes feedback ($\mu$) and a too weak heat flux feedback ($\alpha$). Using the Tiedke scheme (TI), a too strong deep convection leads to a doubling of $\alpha$. Any SST anomaly in the east Pacific is therefore strongly damped and, since the Bjerknes feedback remains weak, ENSO is suppressed in TI. The mechanisms are unraveled by analyzing the atmosphere-only equivalent simulations (AMIP-KE and AMIP-TI).

The detailed analysis of the heat flux feedback in the east Pacific, where the damping of SST is key to ENSO evolution, shows that the shortwave heat flux (SHF)
feedback $\alpha_{SW}$ is the dominant term, especially during the ENSO amplification phase in NH summer and autumn. The SHF feedback is further analyzed as a blend of two different regimes, both influenced by the large-scale circulation: the deep convection regime, where $\alpha_{SW}$ is negative, and the subsidence/cold tongue/marine stratocumulus clouds regime, where $\alpha_{SW}$ is positive (Fig. 13b). In AMIP-KE, the subsidence regime is too extended and explains the too weak absolute value of $\alpha_{SW}$ in this simulation. In AMIP-TI, the deep convection regime is stronger and explains the increased absolute value of $\alpha_{SW}$ (Fig. 12 and Figs. 13c,d). These different regime distributions can be traced back to different deep convection thresholds in the east Pacific (28.2°C versus 27.2°C, Fig. 15). In KE and TI these biases are amplified as the subsidence regime (in KE) and the deep convection regime (in TI) take over the entire east Pacific. Interestingly, east of the date line, the structure of $\alpha_{SW}$ for AMIP-TI is closer to that of the ERA-40 reanalysis (see also Hourdin et al. 2006), whereas the structure of $\alpha_{SW}$ for AMIP-KE is closer to that of the ISCCP satellite observations (Fig. 13). As shown by Cronin et al. (2006), the ISCCP radiative feedbacks compare better with that of in situ observation than ERA-40, which has a too strong convection over the ITCZ, like AMIP-TI. Hence, the correct value for $\alpha_{SW}$ in the TI physics is again due to an error compensation, this time between a too strong deep convection and a too weak subsidence regime—both associated with a too low SST deep convection threshold.

Many of these errors in KE and TI can be traced back to errors in the background state (Fig. 2), most of which are already present in the AMIP simulations, pointing to a predominantly atmosphere origin. The westward shift of the cold tongue/trade winds system in KE (already present in AMIP-KE) will enhance the regions of positive SHF feedback. The too strong deep convection in TI (and AMIP-TI) and a too warm SST in the east Pacific associated with a too low deep convection threshold will, on the other hand, enhance regions of negative SHF feedback. Interestingly, the mechanisms that lead to errors in the background state in TI—namely, a damping of the seasonal cycle (B07)—also operate at interannual time scales, that is, to damp ENSO. This gives hope that alleviating the background errors will also correct the interannual features of the model.

According to Hourdin et al. (2006), the marine stratuscumulus cloud cover in AMIP-KE is much too weak (their Fig. 19). Deser et al. (1993) have shown that, if the background amount of low clouds is underestimated, the radiative response to changes in low-level subtropical cloudiness is overestimated. This agrees with the analysis presented in Fig. 14a, where $\alpha_{SW}$ in AMIP-KE is more positive than that of ISCCP. Conversely, Hourdin et al. (2006) also show that AMIP-TI has more low-level clouds in the subtropics than AMIP-KE. This leads to a weaker positive $\alpha_{SW}$ in the subsidence regime over the cold tongue region (Fig. 13d) and an overall stronger SHF damping in the east Pacific (Fig. 14a). These differences between TI and KE physics are most prominent when the cold tongue is present, that is, from June to October—a key amplification period for ENSO and a time when AMIP-TI exhibits severe biases (Fig. 2f). The underestimation of low-level subtropical cloudiness is a classical and persistent bias of coupled GCMs (Karlsson et al. 2008). This suggests that most models will overestimate the SHF feedback in the east Pacific during ENSO, leading to an overall weaker $\alpha$ in this region, as noted in many models (Sun et al. 2006, 2008; Philip and van Oldenborgh 2006; Lin 2007; Lloyd et al. 2009). Although the low clouds are a prime candidate to explain these SHF biases, high cirrus clouds could also contribute to $\alpha_{SW}$ changes. A more detailed analysis will be required to fully conclude on this point.

The origin of the too weak Bjerknes feedback ($\mu$) both in KE and TI is more challenging to track down and involves both local and nonlocal mechanisms. That the AMIP simulations have about the correct value for $\mu$ (Fig. 9) but that this value is divided by about two in the coupled simulation points to errors in the mean state, that is, changes in the mean SST gradients that drive the
atmosphere circulation. Marti et al. (2009) show that, when the atmospheric meridional resolution is increased from 2.5° to 1.8° in KE, $\mu$ is also increased from 4.6 (Fig. 9) to 6 N m$^{-2}$ K$^{-1}$. Besides improved internal dynamics, the fact that the atmosphere can better “see” the equatorial ocean waveguide (2°N–2°S) and, hence, have a stronger dynamical response to the same SST anomaly can explain this result. Lin (2007) also argues that $\mu$ is a function of the background wind speed (since the zonal wind stress depends on it), which has errors in GCMs. In any case $\mu$ exhibits a large diversity among present models (Guilyardi 2006; Lloyd et al. 2009), and more work is needed to fully understand this diversity. These too low $\alpha$ and $\mu$ atmospheric feedbacks can explain both the too early decay of El Niño in KE (Fig. 4) and the shortlived upwelling in the seasonal cycle (Fig. 2c).

Several other studies explored the sensitivity of the ENSO amplitude to $\alpha$ and $\mu$. Battisti and Hirst (1989) showed that the amplitude of El Niño is anticorrelated with the intensity of $\alpha$. In their theoretical study and using a linear model, they estimated that the ENSO amplitude is divided by 5 if $\alpha$ doubles, a relation close to what is found between KE and TI. The weaker $\alpha$ in coupled models when compared to their equivalent AMIP configuration was also noted by Sun et al. (2008), who attribute it to a weakened precipitation response in the coupled mode. The results presented here suggest that this may also be associated to shortcomings in the representation of marine stratocumulus cloudiness and/or the deep convection threshold.

Wu and Kirtman (2007) pointed out that AMIP-type simulations have shortcomings in the representation of air–sea interaction in regions and periods when the atmospheric forcing dominates over the SST forcing. This occurs mostly during the warm/rainy season in convective regimes and can explain the stronger $\alpha$ in AMIP-KE and AMIP-TI when compared to that of KE and TI. That the feedbacks are better represented in the coupled mode can also be seen from Fig. 13, where the structure of the KE and TI feedbacks are closer to that of ISCCP (the bow-shaped negative feedback surrounded by positive feedback), whereas the AMIP-KE and AMIP-TI feedback structures are closer to that of ERA-40, which is also based on a model forced by observed SSTs.

The coexistence of two regimes of opposite feedback sign for $\kappa$ possibly has implications for the response of ENSO to climate change. Higher SSTs can lead to more occurrence of the deep convection regime (provided that the SST deep convection threshold remains the same) and an overall stronger negative feedback, as shown by Philip and van Oldenborgh (2006) for several coupled models. Nevertheless, the other feedbacks relevant to ENSO (background state, ocean feedbacks, . . .) will also evolve. In the models used for the IPCC AR4, the resulting balance of feedbacks is such that no clear trend in ENSO change is seen for the next century. As shown here, a systematic and detailed evaluation of atmospheric feedbacks during ENSO in the present-day climate simulations, for which observations are available for validation, is a necessary step to fully understand the ENSO simulation in coupled GCMs, and its likely change in the future. A first attempt to evaluate these atmospheric feedbacks in the CMIP3 models is presented in Lloyd et al. (2009).

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