

Climate Model Forecast Experiments for TOGA-COARE

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Abstract

Short term (1 - 10 days) forecasts are made with climate models to assess the parameterizations of the physical processes. The time period for the integrations is that of the Intensive Observing Period (IOP) of the Tropical Ocean Global Atmosphere Coupled Ocean - Atmosphere Response Experiment (TOGA-COARE). The models used are the National Center for Atmospheric Research (NCAR) Community Climate Model (CAM) version 3.1 (CAM 3.1), CAM 3.1 with a modified deep convection parameterization and the Geophysical Fluid Dynamics Laboratory (GFDL) Atmospheric Model (AM) version 2 (AM2). The models were initialized using the state variables from the ERA40 reanalysis.

The CAM deep convective parameterization fails to demonstrate the sensitivity to the imposed forcing to simulate precipitation patterns associated with the Madden Julian Oscillations (MJO) present during the period. The AM2 and CAM 3.1 modified exhibit greater correspondence to the observations at the TOGA-COARE site ,suggesting that convective parameterizations that have some type of limiter (as do AM2 and the modified CAM3.1) simulate the MJO rainfall with more fidelity than those without. None of the models are able to fully capture the correct phasing of westerly wind bursts with respect to precipitation in the eastward moving MJO disturbance. Better representation of the diabatic heating and effective static stability profiles is associated with a better MJO simulation.

Because the models' errors in forecast mode bear a resemblance to the errors in climate mode in simulating the MJO, the forecasts may allow one to better dissect the reasons for model error.

1. Background

For the purpose of validating the parameterizations used in climate models, the models are used to produce weather forecasts. Over the course of a short term forecast (say 3 to 10 days) the errors emerging from individual parameterizations are not as intertwined as in long climate simulation and thus a source of trouble might be more apparent. The forecast might also shed light on the etiology of the climate biases if these manifest themselves in the short integration. The utility of a short term forecast to diagnose errors requires that the model be started from as realistic conditions as possible. Deviations of the model forecasts from the subsequent observed weather can be documented and the causes diagnosed in manner more transparent than in a statistical evaluation of a climate simulation.

The model integrations are run over the intensive observing period (IOP) of the Tropical Ocean Global Atmosphere Coupled Ocean - Atmosphere Response Experiment (TOGA COARE) which spanned the period from 1 Nov 1992 to 28 Feb 1993. This time period is rich in varied tropical phenomena, Gutzler et al. (1994), and has a comprehensive amount of observational data. Many aspects of the TOGA COARE period have been studied extensively, permitting a useful framework for model verification. Intraseasonal variability was quite pronounced during the IOP. This afforded examples of the Madden-Julian Oscillation (MJO) and the wind bursts embedded within them, Lin and Johnson (1996). In this work extensive use will be made of the observations taken on the Intensive Flux Array (IFA) of TOGA-COARE centered at about 2°S, 155°E.

The well documented MJO activity occurring during the TOGA-COARE IOP

affords a critical evaluation of the models' parameterizations driven by this very important form of tropical variability. The MJO is a convectively coupled and highly viscous Kelvin-Rossby wave which moves along the Equator eastward from the Indian Ocean to the Dateline with a phase speed of about 5-10 m/s, Lin and CoAuthors (2006). The MJO is the dominant tropical intraseasonal mode but its poor simulation is endemic among general circulation models, Slingo and CoAuthors (1996), Lin and CoAuthors (2006). Generally, the oscillations tend to be too weak and too fast in those models where they have some signal. Previous modeling studies indicated that the MJO depiction is sensitive to the representation of convection. Slingo and CoAuthors (1996) indicates that those models with CAPE closure tended to be more realistic. It was also found that adding moisture triggers to the convective schemes aided the MJO signal. The vertical profile diabatic heating also could be significant in driving MJO variability, Lin et al. (2004). The forecast framework permits the evaluation of many aspects of the model in the context of observed MJO dynamic forcing.

The structure of the paper is as follows. The next section will present the climate models used. This will be followed by the experimental design. Section 4 documents the verification data which is cited and this is followed by the presentation of the results. The paper's conclusions are presented in the final section.

2. Models Used

Since this study is focused on short time scale processes in the Tropics, the convective parameterizations of each model will be the only aspect to be described in any detail.

a. NCAR Community Atmosphere Model (CAM) 3.1

The CAM was the unmodified version 3.1, Collins et al. (2004), distribution. The model was run using the eulerian, spectral dynamics with a spectral resolution of T42 and 26 vertical levels (L26) and 30 vertical levels (L30). The deep convection parameterization uses the CAPE (Convective Available Potential Energy) based closure described by Zhang and McFarlane (1995). The scheme is based on a plume ensemble approach where it is assumed that an ensemble of convective scale updrafts (and the associated saturated downdrafts) may exist whenever the atmosphere is conditionally unstable in the lower troposphere. The closure condition assumes that deep convection acts to consume CAPE within a fixed relaxation time of 2 hours. The cloud base mass flux is directly proportional to the value of CAPE at each gridpoint. The scheme is triggered if CAPE exceeds 70 J/kg. The CAM also has a parameterization to characterize the convective forcing associated with shallow and middle level convection, which is not treated by the deep convection scheme. For convenience, this shallow/middle convection parameterization will be referred to as the Hack scheme, Hack (1994).

b. CAM Convection Parameterization Modification

The modification to the Zhang-McFarlane (ZM scheme) deep convection parameterization described by Zhang and Mu (2005) was implemented in the CAM 3.1, T42 L26. Hereafter this modified version will be referred to as the ZMO. The closure of the modified scheme assumes the cloud base mass flux is proportional to the CAPE change due to the temperature and moisture changes above the boundary layer that result from the large-scale processes. In this case the free troposphere is the region above the boundary layer. The modified scheme includes a relative humidity threshold as a convection trigger to suppress convection in conditions when the boundary layer is too dry. The relative humidity of the air at the parcel lifting level must be greater than 80 percent for deep convection to occur. Additionally, the new scheme allows convection initiating above the PBL top to be included in the deep convective parameterization. In the default CAM this upper level instability is for the most part addressed by the Hack convection parameterization. The Hack scheme itself is unmodified from the default CAM formulation.

c. GFDL Atmospheric Model(AM) 2

The AM2, GAMDT (2004), was run in the bgrid ($2.0^\circ \times 2.5^\circ$) configuration with 24 vertical levels. The AM2 cumulus parameterization is a relaxed Arakawa-Schubert formulation, Moorthi and Suarez (1992). Closure is determined by relaxing the cloud work function for each cloud type in the spectrum back to a critical value over a fixed time scale. The cloud work function is reduced to zero for shal-

low updrafts that detrain below 600 hPa. Deep convection is prevented in updrafts with a lateral entrainment rate lower than a critical value determined by the depth of the sub-cloud layer, (Tokioka limiter, Tokioka et al. (1988)). This entrainment limiter is only applied to deep convection(tops above 500 hPa). Use of this limiter increases the amplitude of tropical transient activity to bring it closer to observed values. In the AM2 implementation of the relaxed Arakawa-Schubert, there are no parameterized downdrafts.

Convective momentum transport (CMT) is parameterized by a downgradient diffusive formulation. The impact of cumulus convection on the horizontal momentum is represented by adding to the vertical diffusion coefficient for momentum a term directly proportional to the cumulus mass flux and depth of convection. The current (3.1) version of the CAM does not have any parameterization of cumulus momentum transport.

3. Experiment Design

a. Initialization

The availability of high quality analyses from NWP centers affords an excellent starting point for providing initial data to the climate model. The crucial path is that of taking the analyses which are in balance for the model used in the data assimilation and producing data that is in near balance for the climate model and yet as near to the observed atmosphere as possible.

The models were initialized using the state variables (winds, temperature, spe-

cific humidity, and surface pressure) from the ERA40 reanalysis, Uppala et al. (2005). The ERA40 data were on the original η normalized pressure coordinates (60 levels) and were interpolated to the model η coordinates using procedures based on the ECMWF 'full pos' procedures, White (2002) which are used for initializing the ECMWF models using data from other assimilation systems. The methods employed here mimic the forecast analysis cycle of the NWP centers. At the centers, the analyses are carried out every six hours using all the observations available. A six hour forecast from the previous analysis time supplies the new background for the analysis variables, and provides updates to model fields that are not analyzed such as prognostic clouds and land variables. The cycle then proceeds with another six hour forecast and analysis.

In modern data assimilation systems the state produced by the analysis is usually suitable for the starting point of the six hour forecast. In the past, the analyzed state was not in balance with the forecast model and techniques, such as Non-linear Normal mode initialization (NMI), were employed so that the model could produce as good a short term forecast as possible. Our situation is analogous to the past problems, in that the model we are initializing is not the model used in the analysis nor is it on the same horizontal and vertical discretization. This mismatch will result in some initial imbalance.

Fillion et al. (1995) demonstrated that use of a properly designed digital filter, Lynch and Huang (1992), could produce initial states that were actually superior to NMI for the purpose of model short term forecasts. The methods of Fillion et al. (1995) were applied to the CAM3.1 forecasts with effective results, in that the

technique effectively removed the initial imbalances observed in the first few hours of the integration from the ERA40 initialized state. These experiments indicated that beyond 24 to 36 hours the forecast converged to virtually the same solution regardless of the initialization used.

The present work requires different models run on many computer systems. Since the digital filter introduced a complication that had minimal effect on the forecast projections which were of interest (2 -3 days), the pragmatic approach was to not use it. Therefore, the only initialization used for the experiments reported here is a careful interpolation to the model grid with a smoothing commensurate with the model horizontal resolution. The results are reported for times beyond 36 hours.

The model forecasts are started at 00 GMT and run for up to 10 days over TOGA-COARE period (1 Nov 1992 - 28 Feb 1993 , 120 days). Thus there are 120 forecasts. Tests indicated that running from other times (06, 12, 18) do not affect the results presented here. In order to generate the time series shown in the figures, chunks of data of 24h in duration are extracted from each forecast and concatenated to produce a series of data sections starting at a particular forecast projection. Thus the day 3 forecast time series is made up of the 120 forecasts from hours 48 to 72. The model data are saved every timestep. Day 3 is used extensively in the results section. This period is selected since it is clear of initialization problems as described above, yet it is close enough to the initial time to expect the model to be reasonably close to the observed atmosphere.

b. Specified Sea Surface Temperature

The fluxes of heat and moisture from the ocean surface play a key role in almost all tropical atmospheric systems. In this work the SST are prescribed. The default SSTs are monthly mean values based on observations and linearly interpolated to the required times between values at mid-month. The monthly means used in CAM and AM2 are based on the weekly values of the NOAA OI SST analysis, Reynolds et al. (2002). These weekly data are the highest frequency available suitable for a GCM in the TOGA-COARE period. Experiments were carried out comparing the results of using the weekly and monthly SST data input to the CAM. The differences between using the two data sets were very small for the analyses carried out in this work. We will present the integrations based on the monthly mean SST.

4. Verification Data

Extensive surface and upper level data were available centered at the Intensive Flux Array (IFA) of TOGA-COARE, Johnson and Ciesielski (2000) and Ciesielski et al. (2003). The IFA is centered at about 2°S, 155°E, and spans about 5 degrees in latitude and longitude. For rainfall verification three data sets were used. The TOGA-COARE IFA Budget estimated rain , Johnson and Ciesielski (2000) and Ciesielski et al. (2003), provides good time resolution (6h) but for a single point. The CMAP Pentad rainfall, Xie and Arkin (1997) provides global coverage on a 2.5° x 2.5° grid, but for 5 day means. The SSMI rainfall estimates, Wentz and Spencer (1998) are global on a 0.5° x 0.5° grid and provide nominal daily data.

The SSMI precipitation estimates are not available over land due to the nature of the retrieval algorithm.

For the state variables outside the IFA location the data used for verification were the ERA 40 reanalysis, Uppala et al. (2005).

Note that the IFA data used here has had a correction applied to the humidity sounding data as described by Ciesielski et al. (2003). Ciesielski et al. (2003) maintain that these corrections produce moisture fields free of artifacts due to humidity sensor anomalies. The ERA40 analysis was carried out using rawinsonde data that were not corrected. Thus there will be some discrepancy between the IFA and ERA40 data. Ciesielski et al. (2003) provide an extensive discussion of how these corrections affect the moisture profiles and the variables derived therefrom for the IFA. The possible effects that the difference in rawinsonde data between the IFA and ERA40 will be discussed when the variables affected by this difference are presented in the results.

5. Results

The presentation of results will expand outward from point data at the IFA, to a band about the Equator from 5°S to 5°N and finally to the entire Tropics. The IFA data allows a detailed examination of many processes in the vertical but this analysis is confined to a single point. Figure 1 is provided to place the IFA in the context of the larger space/time variability in the tropics during the TOGA-COARE IOP. Figure 1 is a time-longitude plot of daily SSMI rainfall estimates averaged

over 5°N to 5°S . The missing values, mostly over land, are grey shaded. From about 100°E to 130°E the Maritime continent imposes what is probably an artificial minimum due to the presence of some land in the latitude band. Nonetheless, there are two very obvious MJO events which propagate from the Indian Ocean, through the IFA (155°E), and proceed to the Dateline. One begins on about 28 Nov at 75°E and passes the IFA on about 18 Dec. The other originates on about 7 Jan near 30°E and reaches the IFA about 27 Jan. There are more localized events at the IFA near 8 Dec and 6 Feb.

a. IFA

• Rainfall at IFA

Figure 2 shows the time series of rainfall at the IFA for the observations and the model's day 3 forecasts. The propagating features evident in Fig. 1 contribute to the rain events in the middle of December and January. Fig. 1 also indicates that the episodic, locally intense nature of the rain in the larger region is shared by the IFA.

Comparison to observations illustrates that the CAM tends to rain not as intensely as observed. The observations are averaged over an area comparable to a model grid cell so the comparison in this sense is fair to the model. Although there are variations, all the observational precipitation data sets point to the CAM systematically underestimating the intensity of rain events. This type of behaviour can be linked to the CAPE trigger/closure of the convective scheme. At the IFA the observed values of CAPE (not shown) exceed the threshold specified in the

model convective scheme for the entire period. This leads to the persistent rain. The failure of CAPE to be a useful indicator of Tropical rainfall has been noted before, McBride and Frank (1999). The scheme is triggered almost every day, and the constant consumption of the available CAPE inhibits a buildup for more explosive events.

The ZMO clearly exhibits more intense events than the CAM, although these do not always coincide with the observations and it misses some significant events. ZMO's correlation with the observations is 0.27 compared to 0.34 for the CAM. The ZMO rainfall is markedly more episodic than the CAM, the variance is 84 mm^2d^{-2} versus 24 mm^2d^{-2} for the CAM (Obs 55 mm^2d^{-2}). This variability is probably due to the restrictions imposed by the RH trigger and using the large scale forced CAPE changes in place of CAPE trigger. Experiments made with the RH trigger disabled indicate that its use yields a higher frequency of intense events in the Tropics.

The precipitation time series for day 3 of the AM2 forecasts, Fig. 2, has maxima at about the same level as the CAM but with slightly more defined minima between events. The AM2 variance is 37 mm^2d^{-2} and the correlation with the obs is 0.35.

Lin and CoAuthors (2006) comment that the IPCC models demonstrating a high level of persistence in the rainfall events tend to do poorly in MJO simulations. Fitting an AR1 model to all the time series in Fig. 2 yields day 1 autocorrelation values of 0.74, 0.82, 0.62, and 0.81 for the Observations, CAM, ZMO and AM2, respectively. Lin and CoAuthors (2006) theorize that the use of limiters, such as the RH trigger in the ZMO, can be useful in preventing excessive persistence in

modeled rain.

Temperature and moisture at IFA

Figure 3 presents the time average profiles of the differences of temperature between the ERA40 and IFA observations and the models' day 3 forecasts. These data are presented for averages over the active, inactive, and entire TOGA/COARE periods. The inactive and active periods are listed in Table 1. The periods are chosen subjectively by considering the IFA precipitation(Fig. 2), vertically integrated observed Q1 and observed (ISCCP) deep convective cloud. The active conditions have a substantial contribution from the MJO passages as seen in Fig. 1.

The ERA40 is the data source from which the models are initialized. There are a number of reasons why the two 'observational' data, ERA40 and IFA, will disagree. The first is that the data assimilation system of ERA40 has much different techniques and input sources from the IFA analysis. The second is that the ERA40 analysis is carried out on a vertical and horizontal grid different from the IFA defined domain. The interpolation methods to put the ERA40 values onto the IFA defined region will introduce some differences. The ERA40 values evaluated in Figs. 3 and 4 are those of a grid with a structure similar to the CAM and AM2. The third reason is that the moisture values for the rawinsonde data have been corrected in the IFA analysis, Ciesielski et al. (2003). The input data for the ERA40 did not have these corrections. The corrections are only for moisture.

As seen in Fig. 3 IFA is a slightly colder than the ERA40. This amounts to about 0.25 K with a bump of 0.5 K at 600 hPa. The 600 hPa feature is near the melting layer, 0°C, in the IFA region. The temperature differences from ERA40,

Fig. 3, for the models have very similar shapes across the various averaging periods.

It appears that the models establish a characteristic temperature field distinct from the ERA40 which does not vary rapidly as conditions change. The CAM and ZMO have similar shapes, but above 900 hPa the ZMO has an almost uniform translation with respect to the CAM with a cold shift of about a degree. The AM2 is more consistently cold with respect to the ERA40 through the column. This feature is seen in the AM2 multi-decadal climate simulations, GAMDT (2004). Note that all the models have a cold bias at the upper levels (200 hPa), a defect common to the climates simulated by many GCMs.

Figure 4 is that same as Fig. 3 except that the variable is specific humidity. With respect to the ERA40, the IFA data are moist with respect to ERA40 with values of about 1 g kg^{-1} , and at 800 hPa of about 0.5 g kg^{-1} for the average over the whole period. This difference is quite consistent with the corrections presented by Ciesielski et al. (2003). Given that the models start from the ERA40, the model difference for this variable is calculated as the model minus the ERA40. It is not a certainty that the corrections imply that the ERA40 represent a fatally flawed version the the atmosphere. In any case, the possible shortcomings of ERA40 still permits a meaningful evaluation of the models. Note that despite the fact that the differences in the ERA40 and IFA specific humidity are somewhat larger for the inactive periods, little of this discrepancy is reflected in the model results. Additionally, the longer forecast projections show much the same character and magnitude of difference as the shorter forecasts. Examining the day 5 forecast differences of the CAM (not shown), it was seen that at this farther forecast projection the model exhibits much

the same behavior. This is true for the other models as well. This is a long enough time such that the initial moisture profile has substantially diminished influence. Generally, the magnitude of the model differences is large enough to establish its nature for both observational sets. This is certainly true for temperature but less so for moisture. Finally, the differing shape of the model profiles from the IFA-ERA40 differences indicates that the models have substantially altered the initial difference profiles, so much so as to have some confidence that the problems with the initial conditions are not dominating the answer. Furthermore, as the results depend on ensembles of 120 forecasts the model differences are more likely model errors.

The specific humidity curves, Fig. 4, have a similar character across all the models. There is a tendency for the levels just below 900 hPa to be too dry and then above this level to be too wet. The CAM exhibits the largest differences. For all models the differences are similar across averaging periods. Interestingly, the difference at the upper levels tends to be less for the active periods. This is at least partly due to the fact that the models active/inactive periods do not wholly coincide with the observations. The difference profile for the CAM is quite similar to that seen in climate simulations of the model in this region.

• Q1 and Q2 at IFA

Some insights into the formulation of the profiles in Figs. 3 and 4 can be seen from an examination of the Q1 (apparent heat source) and Q2 (apparent moisture sink) profiles for the observations, and models averaged over the active, inactive, and entire TOGA/COARE period. The Q1 and Q2 values for the observations are computed as residuals from temperature and moisture budgets. Thus the Q1 in-

cludes effects of radiation and moist processes. The model values are computed to coincide as close as possible to what the observations are attempting to estimate. Unlike the temperature and specific humidity, the Q1 profiles generally show distinct differences for the various averaging periods. As seen in Fig. 5 during the active periods the observed profiles display a heating rate maximum around the 450 hPa level. There is some indication that the heating peak occurs lower in the atmosphere for the inactive period. The CAM Q1 data, Fig. 5, has the heating maximum at a lower level and is sharply peaked. This sharp heating peak coincides with the warm bias for this model shown in Fig. 3. The rapid decrease above and below the peak compared to the observed might contribute to the relative cooler regions in Fig. 3 on either side of the warm bias. The CAM Q1 shows significantly greater heating than the observations during the inactive periods. This is at least partially due to the fact that the periods were chosen on the basis of the observed atmosphere behavior and the CAM convective scheme is invoked when the observations indicate little activity(Fig. 2).

The ZMO Q1 curves, Fig. 5, show a good agreement to the observed upper level maximum in heating, especially during the active periods. The sharper than observed gradients on other side of the maximum in the ZMO, might contribute to the cold bias on either side of the heating maximum. The inactive Q1 for the ZMO shows virtually no upper level heating indicating that the limiter is a bit too severe in its effects. The improvements at the upper level for the ZMO heating profile are accompanied by some anomalies at the lower levels. The ZMO shows a layer of cooling for all periods below 700 mb which is not present in the observations.

The AM2 has the heating maximum at about the same level as the observations during the active periods but with somewhat diminished amplitude with respect to the observed. The overall low amount of heating in the profile is consistent with the cold bias of this model. However, the shape of the curve is like the observed. The AM2 also shares the anomalous cooling below 700 hPa with the ZMO.

Figure 6 displays a breakdown of the contributions to Q1 by various model processes. For simplicity, only averages over the entire TOGA-COARE experiment are shown. The relative relationships of the components are for the most part the same as the average in the active and inactive periods, the active periods having greater amplitude than the inactive.

Not surprisingly, for the Tropics at this time scale, the peak of the heating due to moist processes very much determines the Q1 profiles. In both CAM versions the upper level maximum in the moist processes is formed by an interplay of the ZM and Hack convective schemes. In both models the ZM contribution dominates below 600 hPa and the Hack above this level. The cooling maximum at about 900 hPa in the ZMO can be seen to be due to the long wave cooling, which is offset by the moist heating in the CAM. The ZM heating is a factor of two greater in the CAM compared to the ZMO below 850mb. The elevation of the Q1 peak in the ZMO with respect to the CAM is seen to be due to an enhanced contribution by the Hack scheme at the upper levels. The calling sequence in the CAM3.1 is deep convection, shallow/middle (Hack) convection and stratiform precipitation. Since the deep convection is called so often in the CAM is is not surprising that the Hack has so little impact since the ZM removes most of the instability. With the less

frequent deep convection in the ZMO, Hack convection is more active.

The Q1 profiles for the AM2 are shown in Fig. 5. For the active periods the AM2 Q1 resembles that of the ZMO although the amplitude of the upper level heating is much smaller in the AM2. Both have a similar low level cooling below 750 hPa, at odds with the observations. The average Q1 for the AM2 does not have the well defined peak of the observations, but rather a slight local maximum at 400 hPa. Examining the components of Q1 in the AM2, Fig. 6, it can be seen that the heating maximum for the AM2 is due to the interplay of the large scale and convective heating. On the average the large scale has a relatively smaller amplitude and this results in the modest maximum at the upper levels.

Lin et al. (2004) indicate that stratiform precipitation contributes to a top heavy vertical heating profile seen during MJO events. Experiments were run in which the Hack parameterization was shut off in the CAM and ZMO. The results to these experiments showed that the large scale precipitation in the models took over the role of the Hack scheme in the distribution of heating and the timing of the precipitation. The ZMO had a slightly more top heavy profile with the Hack turned off. Thus all the models showed the tendency of the large scale precipitation to contribute to the top heavy heating profile seen by Lin et al. (2004), with the ZMO, with Hack turned off having the composition of the profile most like the observations. In any case the lower level of the CAM heating would make it less like the observations of MJO active periods. The ZMO has a heating profile at the upper levels most like the observed.

Figure 7 is the observed apparent moisture sink (Q2) averaged over the same

periods as the Q1 data. Ciesielski et al. (2003) compute the impacts that the moisture corrections have on the estimates of Q2 for the IOP. The change is positive at all levels and decreases with height. The change has a maximum at the surface of about 0.75 Kday^{-1} , decreasing to 0.5 Kday^{-1} at 900 hPa and further decreasing to near zero at 700 hPa. Given the magnitudes of the differences in Fig. 7, it is probably safe to assume that the ERA40/IFA data differences will not be a determining factor on the interpretation of the results.

The observed mean values, Fig. 7, evince a drying throughout the column with a slight maximum at about 450 hPa for the active periods. The magnitude of the drying is fairly constant between 800 and 400 hPa. Both CAM models show more structure than the observed data with distinct extrema at the upper and lower levels. Both CAM versions have a drying peak at about 400 hPa, the ZMO peak is broader and at a higher level than that of the CAM. The CAM has a drying peak at 900 hPa which has an amplitude comparable to the upper level feature. This is a factor of four greater than observations, and could account for the CAM lower level dry bias. The ZMO has a moistening maximum at about 950 hPa, in sharp disagreement with the CAM and observations. This moistening would help explain the reduction of the dry bias seen in the CAM but at the price of introducing behavior apparently at odds with observations. The large bias of ZMO in Q2 at the lower levels is largely offset by advection (not shown), resulting in the modest bias specific humidity (Fig. 4). The AM2 at the upper levels, above 600 hPa, compares fairly well with the observed Q2. The AM2 maximum of drying is about $2 \text{ g kg}^{-1}\text{day}^{-1}$ while the observations have a value of $3.5 \text{ g kg}^{-1}\text{day}^{-1}$. The AM2 is noisy below 850 hPa,

but there is a tendency toward a bit more moistening than seen in the observations.

Figure 7 present the profiles of the main components making up the Q2 for the models. The much reduced role of drying by the deep convection in going from CAM to the ZMO accounts for the anomalous low level moistening in ZMO. As in the Q1 plots, the Hack scheme plays a more prominent role in ZMO as compared to CAM. In the lower levels the strong moistening is at odds with the observations. As with the Q1 profiles, the AM2 and ZMO Q2 profiles have a similar shape, the AM2 having a reduced amplitude with respect to the ZMO. In the AM2 the small low level moistening appears to be a small imbalance between very large contributions from the convective drying and vertical diffusion moistening.

• **Zonal Wind at IFA**

Figure 9 displays the profiles of the time average zonal wind at the IFA for the observations and the Day 3 forecasts of the CAM, ZMO and AM2. The ERA40 zonal wind (not shown) agrees well with the IFA data. The CAM and observations show fair agreement except for the sharp decrease in amplitude in the CAM below 850 hPa. The model tends to underestimate the zonal wind in the lowest levels. The model slightly overestimates the westerly wind above 800 hPa. The ZMO has a similar behavior for this variable but with a bit more easterly bias at the lowest levels. The 26 level CAM has rather coarse resolution at the lowest levels. An experiment with a 30 level version of the CAM was run to ascertain whether the poor low level resolution might account for the bias in the zonal wind simulation. The 30 level CAM used here has 8 levels below 800 hPa as compared to 4 in the 26 level CAM. The 30 level model winds (not shown) present some small differ-

ences in structure attributable to the finer resolution of the model but the increased resolution produces essentially the same profile as Fig. 9.

Figure 10 displays the time series of the zonal wind at 900 hPa for the observations and the day 3 CAM forecasts at the IFA. This figure clearly shows the strong easterly bias of the model which especially evident during periods of stronger observed westerlies. This underestimation of the westerly zonal wind can also be seen in climate simulations of the CAM in the IFA region. The climate DJF values of the lowest level zonal wind in the CAM display an underestimate of $2\text{-}3 \text{ ms}^{-1}$ over the TOGA-COARE region compared to the NCEP reanalysis.

Figure 10 shows that the AM2 produces a very credible simulation of the zonal wind at this level without the significant problems exhibited by both versions of the CAM used here. The AM2 has more model layers (9) in the boundary layer than the CAM (4). However, the 30 level CAM experiment indicates that the number of levels in the boundary layer is not a sufficient condition to ensure a good simulation. The AM2 does have a representation of momentum transport by convective processes which the CAM does not. The AM2 has a mechanism for relating momentum transport to the cumulus mass flux predicted by the relaxed Arakawa-Schubert (RAS) convective scheme, GAMDT (2004). The vertical gradient of the zonal wind is strongly positive in the AM2 during the wind burst events. Given this gradient the CMT scheme of the AM2 will contribute to an increase in the zonal wind at 900 hPa.

The forcing of the wind during MJO events is rather complex. The wind is a result of the interplay of the large scale pressure gradient, mesoscale and cumulus

scale momentum fluxes and pressure gradients. It is the latter two forcings that the climate models attempt to parameterize. As the subsequent very brief discussion attempts to point out, it is still uncertain as how to best capture these forcings in a climate simulation. Gutzler et al. (1994) remark that the TOGA COARE IOP has anomalously strong westerly flow over the IFA and relate this observation to the MJO passages. Lin et al. (2005) by a residual technique indicate the importance sub-grid scale processes in decelerating the zonal flow during an MJO passage. Houze et al. (2000) show that mesoscale circulations act to slow down the westerly wind in the onset region of the the MJO and transported westerly momentum downward in the strong westerly region thus accelerating the lower level wind burst after onset. Mecham et al. (2006), using a cloud resolving model, indicate that the convective and stratiform regions are both important in describing the momentum transport for the TOGA-COARE MJO passages. They show that sometimes these sub-grid scale mechanisms work in concert and sometimes oppose the sense of the large scale momentum transport. Furthermore, the convective and stratiform fluxes can have different roles depending on the phase of the MJO. Such complications make the parameterization of the momentum budget quite problematical during these events. Indeed, it was in part due to these uncertainties that the AM2 decided to use the simpler (simpler than mass-flux formulations, e. g. Gregory et al. (1997)) diffusive scheme.

Experiments were run using a version of the CAM 3.1 modified as described by Wu et al. (2003). This version of the model includes a parameterization of cumulus momentum transport and a modified deep convection similar to the modified

ZM used here. The Wu et al. (2003) CMT is a mass-flux formulation taking into account the momentum transported by the inferred cloud mass flux and also momentum changes forced by the convection induced pressure gradient force. The results for the zonal wind were very similar to the CAM. The simulation with CMT exhibited a wind profile very close to that of the CAM and ZMO in Fig. 9 and also underestimated the zonal wind similar to ZMO in Fig. 10. These results indicate a non-critical role at least for this particular parameterization at this location. GAMDT (2004) comment that the diffusive coefficients for the AM2 CMT are larger at low levels than an equivalent mass-flux formulation. The AM2 results would give some evidence that perhaps the simpler diffusive scheme is superior in this very complex situation. It should be noted that no experiments were conducted turning off CMT in the AM2, so there is no definitive proof that this parameterization played a major role in the AM2's wind simulation.

b. Equatorial Tropics

• **200 hPa Velocity Potential**

A very commonly used diagnostic to monitor MJO activity is a longitude-time plot of band -passed filtered 200 hPa velocity potential. Figure 11 displays the time-longitude plots of the band pass (30 - 70 days) 200 hPa velocity potential for the ERA40 and days 3 and 6 of the CAM, ZMO and AM2 simulations averaged from 5°S to 5°N. The band pass filter is a 2nd order tangent Butterworth filter, applied in the frequency domain, Otnes and Enochson (1978). This analysis provides a useful overview of the ability of the models to capture and maintain the intrasea-

sonal variability present in the upper tropospheric circulation. The ERA40 data (top panel, Fig. 11) shows the two MJO events which passed through the central TOGA-COARE array (155°E , 2°S) during the experiment. The sign convention of the figure is such that local minima correspond to active (upper level divergence) regions. The good correspondence between the top panel of Fig. 11 with Fig. 1 increases confidence in both fields.

Figure 11 demonstrates the ability of all the models to maintain the basic MJO feature for 3 days. Both the AM2 and ZMO maintain a fair representation out to 6 days but the CAM shows a drastic falloff in amplitude. Comparison of the ERA40 data and the CAM in Fig. 11 indicates that even at day 3 the CAM had begun to diminish significantly. Zhang and Mu (2005) demonstrate that this the ZMO can produce realistic MJO variability in coupled model climate simulation although it appears to overestimate the amplitude in Fig. 11 Some possible reasons for this behavior will be put forth in a later section. Keep in mind that the models are initialized by the ERA40 state at the start of each forecast. Thus the possible diminution of the MJO in the CAM is consistent in all phases and locations of the oscillation. Agudelo et al. (2006) performed 30 day forecasts using the ECMWF climate model for the TOGA-COARE period. Agudelo et al. (2006) noted that the forecasts started before the transition period, that is between the suppressed and active phases of the MJO, did a poor job in simulating the convective activity. The short term forecasts reported here indicate that the CAM systematically underestimates the MJO amplitude during all the phases of the oscillation. The decrease in amplitude at 6 days is more than can be attributed to the fact that the model does

not have an interactive ocean, Rajendran and Kitoh (2006).

• Rainfall and 850 Wind

Figure 12 presents time-longitude plots of pentad rainfall data from the CMAP and SSMI observations and ERA40 850 hPa zonal wind averaged from 5S to 5N. The two rainfall data sets agree rather well on the location (in space and time) of the rain events, but the SSMI events consistently have somewhat more amplitude. The westerly wind burst associated with the first MJO is prominent as the MJO approaches the Dateline at about 28 Dec. The phasing and relative locations of the wind and rain maxima are in close agreement with the careful analysis of Lin and Johnson (1996). The maximum in the rainfall precedes the westerly wind peak, the maximum winds occurring just as the active period rain passes. The second MJO which reaches the Dateline at the end of January has some indications of this wind and rain structure but it is not as well delineated as the first. Lin and Johnson (1996) only use the first MJO as an archtypical model. For the first MJO the IFA is to the west of the most intense rain and strongest wind bursts.

Figure 13 is the same as Fig. 12 except for the model rain and winds for day 3 of the forecast. The CAM does capture the pattern of the rainfall but it is evident that the model underestimate of intensity is not confined to the IFA. The precipitation events of the ZMO appear to be too intense compared to CMAP but are more in line with the SSMI. However, these short-lived intense precipitation events in ZMO do not evince the eastward progression seen in the observations. The ZMO does have an increase in precipitation activity in the Indian Ocean with respect to the CAM which a significant improvement since Waliser et al. (2003) cite the lack of

precipitation activity in the Indian Ocean is common to many AGCMs.

The plot of the AM2 in Fig. 13 indicates that the model is more active than the CAM but less than the observations. The pattern of the rain with respect to the westerly wind burst at the end of December is in fairly good agreement with the observations, albeit the amplitudes of the rain and wind are too small. Although the AM2 underestimates the wind burst just west of the Dateline, it's westward extension is such that at the location of the IFA a good simulation is achieved (Fig.10).

The pattern correlations between the rain CMAP observations and CAM forecasts decreases from 0.7 at Day 3 to 0.59 at Day 6 (not shown). The average rainfall over the figure remains at 4.6 mm day^{-1} for Day 3 and Day 6 compared to an observed value of 4.98. The ZMO has a larger rainfall amount than the CAM, the averaged for Day 3 and Day 6 are 5.3 and 5.6, respectively. The increased activity in the ZMO does not capture the observed variation too well, the pattern correlation for Day 3 is 0.5 and this decreases to 0.48 by Day 6. The pattern correlation of the AM2 rainfall with CMAP falls from 0.7 on day 3 to 0.64 on day 5.

In an attempt to get at some possible causes of the differences seen in Fig. 13, simulations were run for a different set of variations on the CAM. The models used are (1) the ZMO with the RH limiter criterion set to zero, (2) the CAM using the modifications of Wu et al. (2003) which is very similar to the ZMO deep convection parameterization but with a complex cumulus momentum transport added, and (3) the CAM with 30 vertical levels. The elimination of the RH limiter in the ZMO has a dramatic impact on the character of the rainfall. While the rainfall has more

intensity than the CAM, it is diminished considerably compared to the ZMO. The Wu et al. (2003) scheme produces a 850 hPa wind simulation (not shown) very similar to ZMO. This would indicate a negligible role for the CMT parameterization in this model for this set of integrations. The poor extension of the westerly wind maxima to the west is similar to the ZMO so that the wind at the IFA has an underestimate of the bursts much like the ZMO. A critical aspect of capturing the wind burst would appear to be the systematic intensification of the rain as the disturbance moves eastward which is not handled well by either model. The CAM with 30 levels is seen to behave in a very similar fashion to the 26 level model.

• **Regional Rain and 850 hPa wind** The preceding sections provided a complete view in time with respect to the model tropical performance but sacrificed spatial detail to achieve this completeness. In this section latitude-longitude plots will be presented to give a two dimensional perspective to the rain and low level wind field for specific time periods. An obvious candidate for such an examination is the MJO passage at the IFA in mid-December.

Figure 14 presents the rainfall averaged over days preceding and after the passage of the first MJO at the IFA. The CMAP rainfall is used, but the SSMI data showed the same character. The observations show the expected shift and strong intensification of rain from the west of the IFA to the east. Inspection indicates that the AM2 captures the eastward progression with the most fidelity of all the models. Keep in mind that the models do not forecast a large propagation in the 2 day forecast. Rather, this is a demonstration of how the models to respond to the varying dynamics presented to it as the MJO proceeds eastward. Note, that this

also shows the AM2 is capable of greater rainfall intensity than the CAM, which is not obvious from the IFA time series. The CAM shows almost no difference in the two plots. The ZMO also shows a good correspondence to the observations, except that overestimates the rain in the region of New Guinea in both time periods. This overestimate of rain over New Guinea and its environs is endemic to ZMO and is quite apparent in the Tropical error pattern for the entire TOGA-COARE period shown later in this section.

Figure 15 displays the 850 hPa wind and MSLP for the ERA-40 and the difference from these values for the models for the period 11 Dec - 1 Jan. This encompassed the period of the westerly wind maxima seen in Fig. 10. The CAM shows the largest differences with a large underestimate of the east-west pressure gradient along the Equator. The very weak rainfall response seen in Fig. 14, is consistent with the dynamics failure to establish the needed pressure gradient. The quality of the ZMO affected by the anomalously strong rainfall over and to the south of New Guinea. This sets up the anomalous circulation to the west of the IFA which disrupts the westerlies and sets up some northerly flow over the IFA. The AM2 has the smallest wind error in agreement with Fig. 10 at the IFA. This is consistent with this model's good representation of the observed rain pattern.

Looking over all the wind maxima and rainfall relationship in Fig. 12, it appears that the eastward intensification of rainfall is often followed by increase in the westerly wind. The ZMO progression of rain is in fits and starts, disrupting the establishment of the winds. The CAM has such small intensity as to be wholly ineffective. The AM2 while having a reduced intensity with respect to the observations

does capture the rain progression smoothly and produces a weaker wind maximum but well positioned.

The evaluation of the lower level zonal wind in the models presents a challenge. Figure 9 indicates clearly that on average the CAM and ZMO underestimate the strength of the zonal wind below 800 hPa, while the AM2 performs better. Figure 10 illustrates that at the IFA the manner in which the CAM and ZMO achieve the average values is quite different. In Fig. 10, ZMO is seen to be better than the CAM during the peak westerly winds in late December but has other discrepancies such that the mean is not that different in Fig. 9. It is seen in Fig. 15 that beyond the IFA, the ZMO displays much less widespread bias than the CAM for the period considered. Note that some of the largest wind differences in Fig. 15 for ZMO occur over the IFA. Just using the evidence from Fig. 9 it would seem that the CAM and ZMO share the same zonal wind properties but extending in time beyond the average and in space beyond the IFA, the ZMO appears to have a distinct advantage over the CAM. The AM2 consistently has the best simulation for all these aspects of the zonal wind.

6. Discussion

The work here indicates that given realistic initial states the CAM deviates from the observations rather quickly. This result demonstrates that the representation of shorter term processes and likely convection, is at fault in this model. The modification of the ZM scheme increased the activity of the model but at the expense of

generating some problematic heating profiles. The new scheme does increase the intensity of precipitation events but these events do not have a high correlation to observations. The correlation for the data in Fig. 13 drops to 0.5 by Day 3 of the forecast which is below the Day 6 level of the control model, CAM.

Lin and CoAuthors (2006) comment that the IPCC models demonstrating a high level of persistence in the rainfall events tend to do poorly in MJO simulations. It has been shown that the CAM rainfall is more persistent than the observations while the ZMO rainfall is a great deal more episodic and less persistent than the observations. The ZMO evinces a much stronger MJO signal than the CAM out to day 6 of the forecast. This is in line with Lin and CoAuthors (2006). However, the AM2 which displays a good MJO signal out to day 6, shows the same level of persistence in rainfall as the CAM. One possible explanation is that the ZMO produces episodic, intense events which are not highly correlated with the observations but can supply energy to the MJO band. The AM2 has events which are more intense than the CAM but are well correlated with the observations. The superior timing of the AM2 rain events allow it to maintain the MJO signal.

Lin and CoAuthors (2006) mention that the characteristics of the waves generated by the IPCC models indicate that the models consistently have equivalent depths that are too large. This can be the result of the models having an effective static stability that is too large. Raymond (2001) defines the effective static stability as:

$$\Gamma_{eff}(z) = \frac{\partial \theta}{\partial z} - \frac{\partial S_\theta}{\partial w} \quad (1)$$

where S_θ is the potential temperature source produced by the diabatic parameterizations, θ is the potential temperature, z is the height and w is the vertical velocity. The value of $\partial S_\theta / \partial w$ is estimated by fitting a line to the S_θ and w data for each level. The slope of this line is taken as the estimate of $\partial S_\theta / \partial w$. The effective static stability attempts to quantify the effect of latent heat release to diminish the negative buoyancy associated with large scale vertical ascent. Figure 16 shows the plots of $\partial \theta / \partial z$, $\partial S_\theta / \partial w$ and $\Gamma_{eff}(z)$ for the observations, CAM, ZMO and AM2 at the IFA averaged over the whole TOGA COARE period. As can be seen the $\Gamma_{eff}(z)$ profiles indicate that the model values are consistently larger than the observed. Since the initial state is specified, the $\partial \theta / \partial z$ curves are approximately the same as the models have not drifted too far. As Lin and CoAuthors (2006) surmised it is the underestimate of $\partial S_\theta / \partial w$ that drives the differences in $\Gamma_{eff}(z)$. The AM2 yields values which are in best agreement with the observations. At least in this case, the models with the lower values of $\Gamma_{eff}(z)$ do a bit better in maintaining the MJO amplitude. The AM2 appears to have the greatest fidelity to many physical characteristics which maintain the MJO, this overcomes some of its shortcomings, such as weak heating.

Figure 17 displays a time-pressure diagram of the band pass filtered (30 -70 day) apparent heat source, Q_1 for the observations and the models. This figure shows aspects of the heating profiles related to MJO forcing which can be obscured by time averaged profiles of Figs. 5 and 6. Of particular interest is the event centered on 21 Dec 1992. This time coincides with the maximum in observed rainfall at the IFA. In Fig. 17 the observations have a maximum at about 450 hPa at this

time. The CAM and AM2 both generally under estimate the magnitude of the Q_1 and especially for the 21 Dec event. The ZMO has a substantially larger magnitude than the other two models but its heating on this time scale is not very discriminate with strong extrema occurring where only weak are observed. Also the ZMO maximum in heating occurs about 4 days ahead of the observed, while the other models coincide almost exactly with the observations. Note that in the developing phase previous to 21 Dec, heating begins at the lower levels and afterwards in the decaying phase the heating ceases more uniformly with height. This picture is in line with the findings of Lin et al. (2004). The models have the opposite slope, with low level heating occurring after 21 Dec in the decaying phase. The AM2 is the only model with low level heating in the developing phase although it is somewhat less organized than in the observations. The contributions to the heating by radiation, convection and large scale process were examined for the models. In both CAM and ZMO the convection was very dominant for the 21 Dec event. In the AM2 the convection was dominant but there was a significant contribution to upper level heating and low level cooling by the large scale during the MJO passage. The AM2 had features which were quite similar to those estimated from observations by Lin et al. (2004). It would appear the the lack of amplitude in the heating in the AM2 can be partially compensated by a more accurate placement of the heating in the vertical (Fig. 6) thus maintaining the circulation as seen in Fig. 11. It should be noted that Zhang and Mu (2005), using the convective parameterization of ZMO, observe low level convective heating proceeding the MJO events diagnosed in coupled climate integrations. Apparently, the model running without constraint can

achieve more realistic aspects than when it is driven by observed fields. This indicates that the forecast technique needs to be used with other diagnostic methods to fully describe the behaviour of any parameterization in a model.

7. Conclusions

Using a climate model as a forecast model allows the forcing of specific events. This makes model assessment more exacting regarding parameterizations than statistics from a climatology. The following conclusions are drawn from results of forecasts over the TOGA-COARE period, using the CAM 3.1, the CAM 3.1 modified ZM parameterization and the AM2.

- The CAM rains too often and produces rain events of insufficient intensity.

The AM2 precipitation at the IFA has similar characteristics, although it has stronger rain events than the CAM away from the IFA.

- The CAM model has very little activity in the Indian Ocean, a shortcoming common to many AGCMs, Waliser et al. (2003).

- The CAM rainfall persistence, low level of maximum heating and increased effective stability are all factors noted by Lin and CoAuthors (2006) as being detrimental to a successful MJO representation in the models. The ZMO and AM2 have better characteristics in this aspect.

- The CAM is unable to sustain the observed MJO intensity over a 6 day forecast. This indicates that the poor MJO characteristics of the climate simulations are not due to poor representations of longer time scale air/sea interac-

tions or radiative processes

- The Zhang modification to the CAM deep convective parameterization results in a stronger, more realistic MJO and time distribution of precipitation. The rainfall events, although as intense as observed are not consistently produced with the same timing as the observations.
- The AM2 maintains a fairly realistic MJO. The IFA rainfall events are not much more intense than the CAM. The AM2 heating profile has a peak higher than the CAM, but with a smaller amplitude. This might indicate that the level rather than the magnitude of the heating is a critical factor.
- Westerly wind bursts are systematically underestimated by all versions of the CAM at the IFA. Increasing the number of model levels in the CAM PBL does not address this problem. This bias over the IFA is largely due to the CAM missing the westerly wind bursts associated with the two MJO events during the period. The underestimate of the westerly wind is also evident in the CAM climate simulations. The ZMO also has a poor representation of these wind events, but for a different reason. The CAM fails to capture the shift of precipitation from west to east of the IFA as the MJO propagates. The forcing driven by the convective heating is much the same for all phases of the MJO. The ZMO does capture the sense of the precipitation shift, but consistently overestimates the rainfall over land to the west of the IFA and does not have a smooth progression to the east.
- The effective static stability, Γ_{eff} , is overestimated by all the models with the AM2 having values closest to observed estimates.

- Overall, the AM2 appears to have the best simulation of the models and variables considered here.
- The inclusion of some type of limiter imposed on the deep convective scheme appears to substantially improve the representation of convective events in the Tropics. The CAM has no such limiters while both the AM2 and ZMO do.

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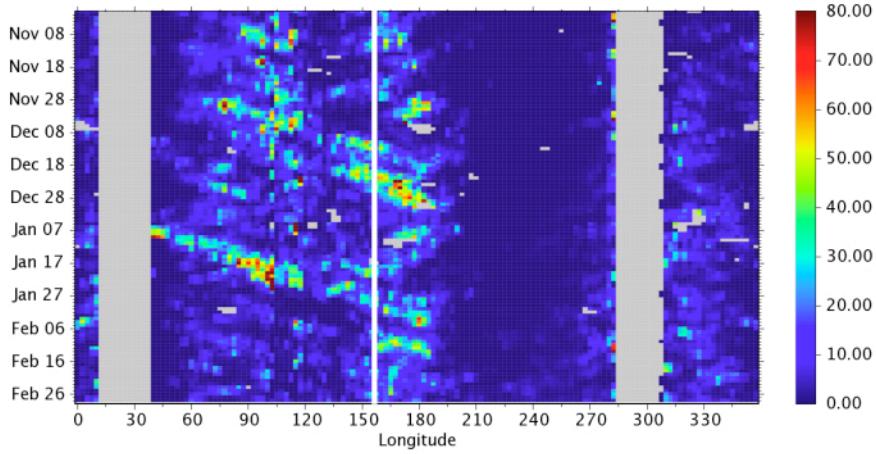


FIG. 1. Time-Longitude plot of rainfall estimates from SSMI data averaged over 5N to 5S.

Units are in mm day^{-1} . Missing data are grey. The retrieval algorithm does not yield values over land. Approximate longitude of IFA is indicated by the white line.

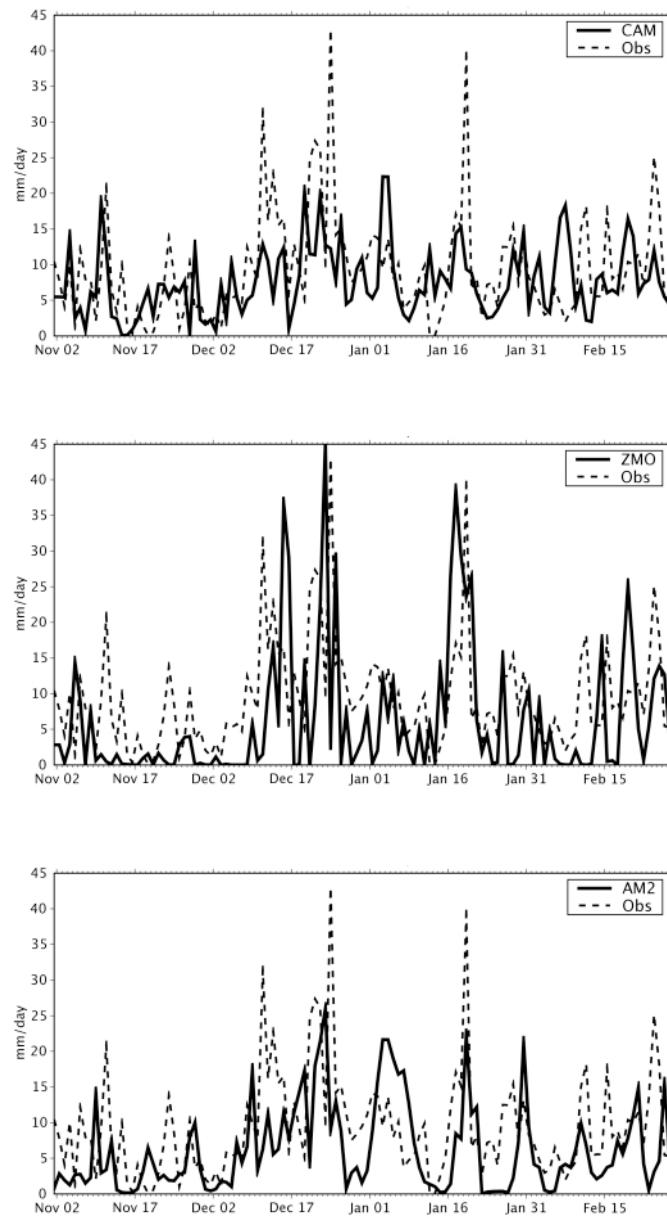


FIG. 2. Rainfall at the TOGA COARE IFA. Observational estimate from IFA array budget and CAM forecast Day 3 (top) , ZMO forecast Day 3 (middle) and AM2 forecast Day 3 (bottom)

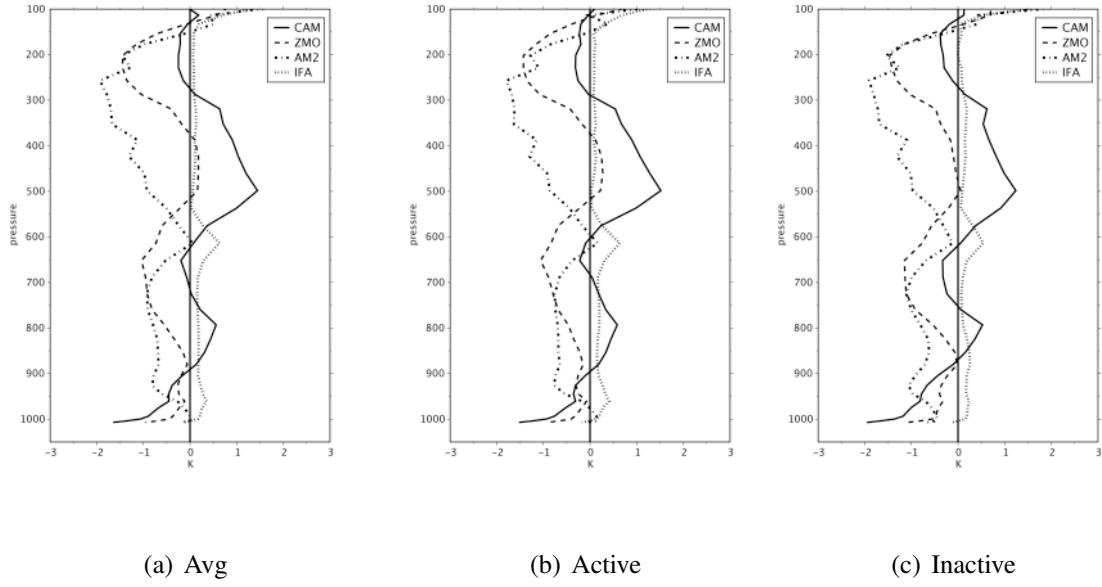


FIG. 3. Profiles of temperature difference from the ERA40 data for IFA Obs and day three forecasts of CAM, ZMO and AM2. Differences are presented for averages over the (a) Average over the entire TOGA-COARE IOP, (b) Active, and (c) Inactive, periods. The plots all indicate the value of the given data minus the ERA40 values.

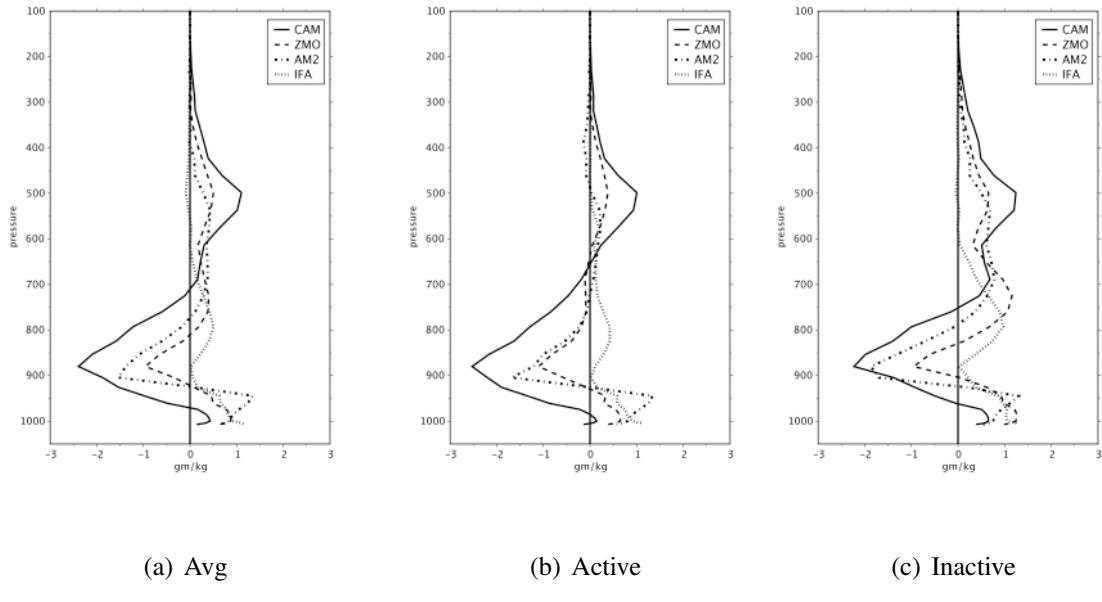


FIG. 4. Profiles of specific humidity difference from the ERA40 data for IFA Obs and day three forecasts of CAM, ZMO and AM2. Differences are presented for averages over the (a) Average over the entire TOGA-COARE IOP, (b) Active, and (c) Inactive, periods.

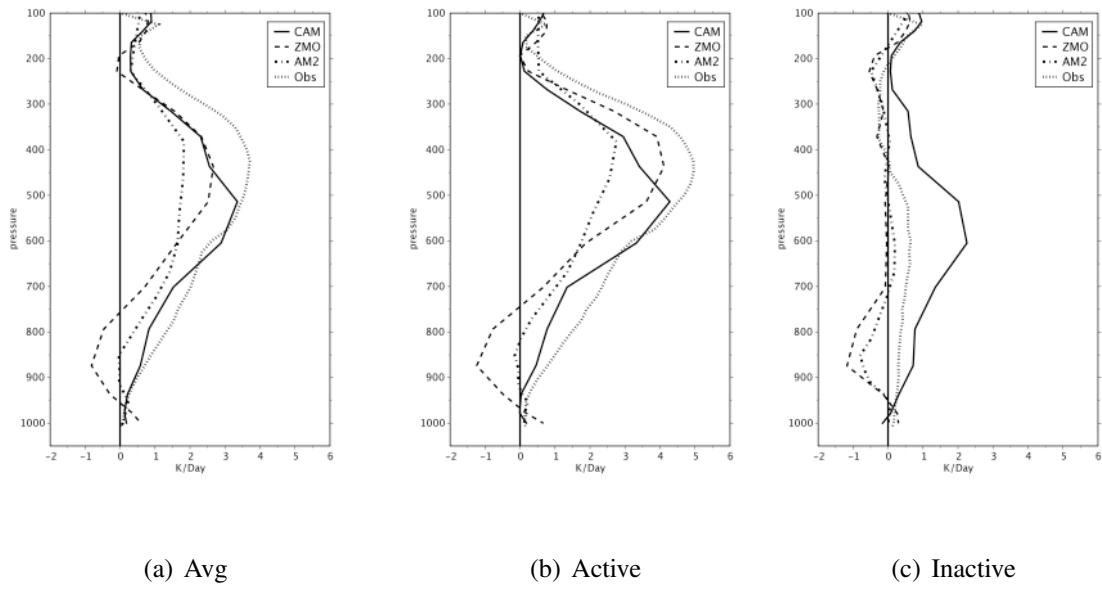


FIG. 5. Profiles of apparent heat source (Q1) at the IFA for the Observations, CAM, ZMO and AM2.(a) Averaged over the entire TOGA-COARE , (b) averaged over Active periods and (c) averaged over Inactive periods. Units are in degrees Kelvin per day.

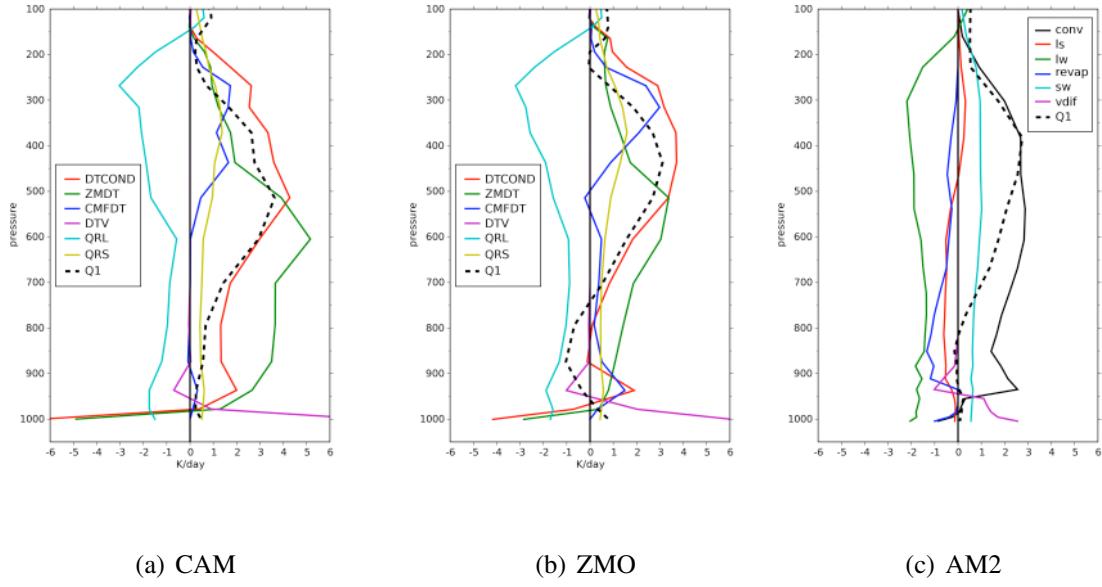


FIG. 6. Profiles of the components making up the apparent heat source (Q_1) at the IFA for the (a) CAM, (b) ZMO and (c) AM2. Units are in degrees Kelvin per day. DTCOND:moist processes, ZMDT: Deep convection, CMFDT: Hack (shallow) convection, DTV:vertical diffusion, QRL:long wave radiation,QRS:shortwave radiation. conv:convective,sw:short wave radiation, lw:long wave radiation, revap:re-evaporation of precipitation, ls:large scale precipitation, vdiff:vertical diffusion. Note for the CAM DTCOND includes both ZMDT and CMFDT plus some other processes not broken out. $Q_1(\text{CAM}) = \text{DTCOND} + \text{DTV} + \text{QRL} + \text{QRS}$, $Q_1(\text{AM2}) = \text{conv} + \text{sw} + \text{lw} + \text{ls} + \text{vdiff}$

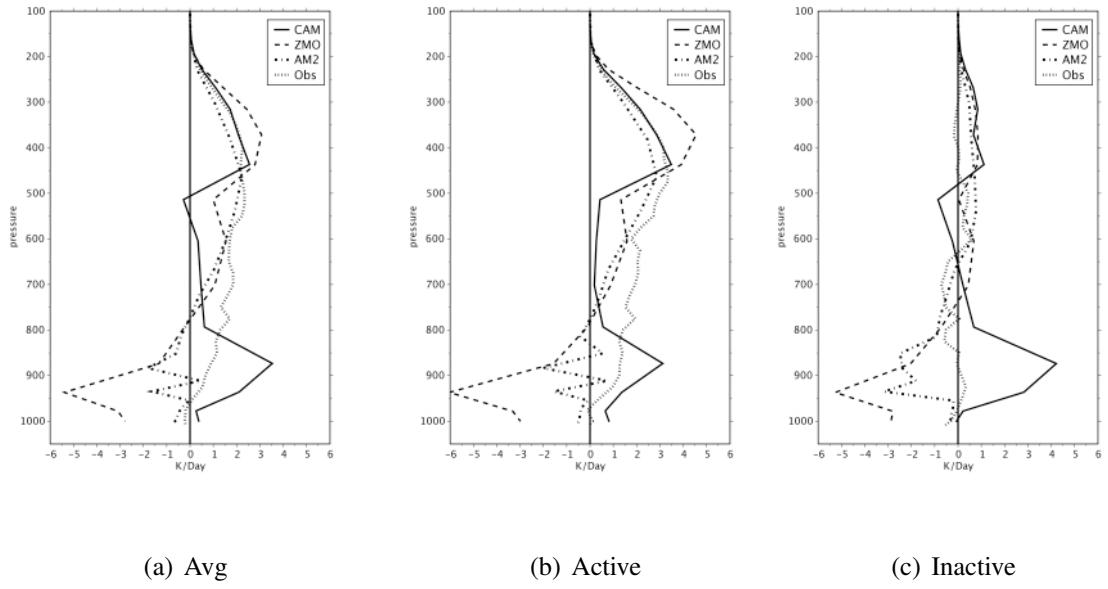


FIG. 7. Profiles of apparent moisture sink (Q_2) at the IFA for the Observations, CAM, ZMO and AM2.(a) Averaged over the entire TOGA-COARE , (b) averaged over Active periods and (c) averaged over Inactive periods. Units are in degrees Kelvin per day.

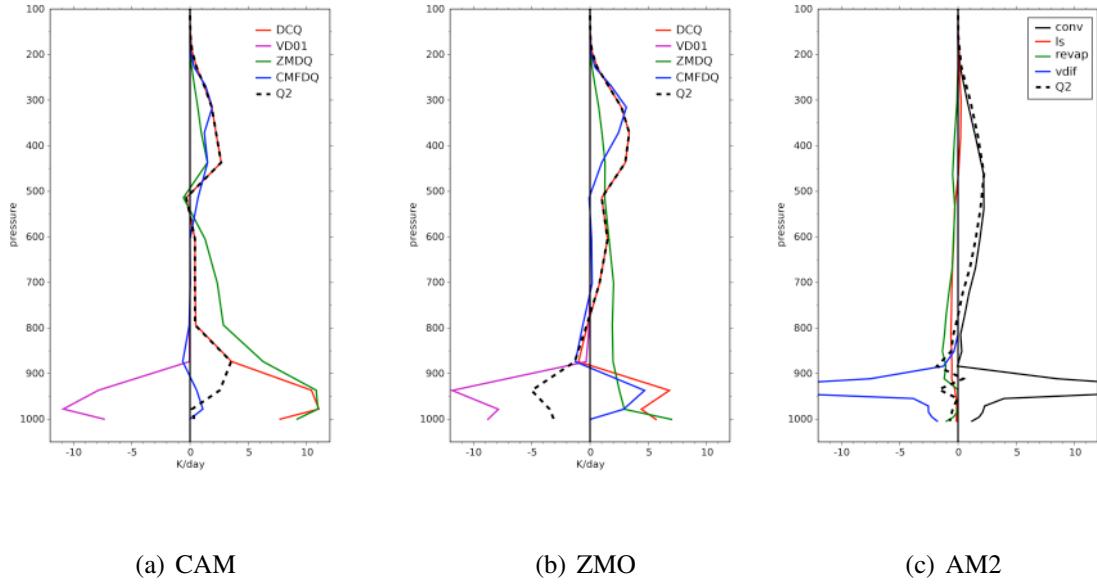
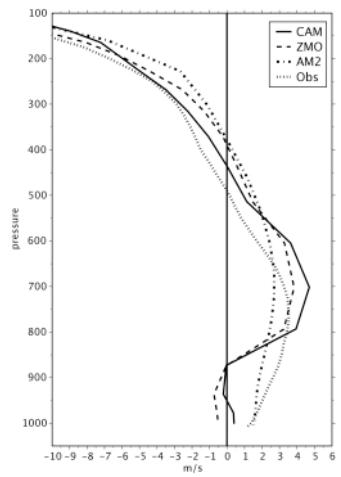
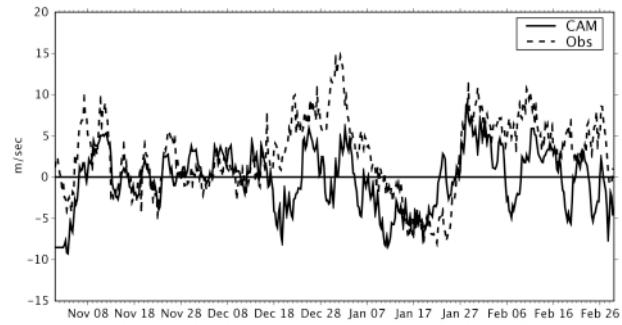


FIG. 8. Profiles of the components making up the apparent moisture sink (Q_2) at the IFA for the (a) CAM, (b) ZMO and (c) AM2. Units are in degrees Kelvin per day. DCQ:moist processes, VD01:vertical diffusion, ZMDQ: Zhang-McFarlane deep convection, CMFDQ: Hack shallow convection. conv:deep convection, ls:large scale precipitation, revap: re-evaporation of precipitation, vdiff:vertical diffusion. Note for the CAM DCQ includes both ZMDQ and CMFDQ plus some other processes not broken out. $Q_2(\text{CAM}) = \text{DCQ} + \text{VD01}$, $Q_2(\text{AM2}) = + \text{conv} + \text{ls} + \text{vdiff}$

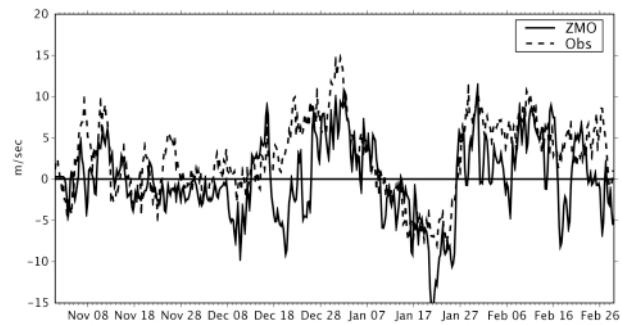


(a) Avg

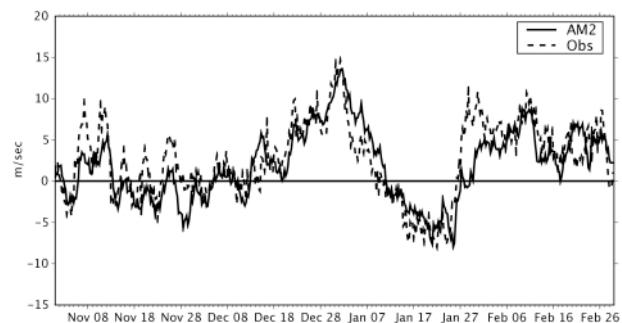
FIG. 9. Profile of avearge zonal wind (m/s) at the IFA for the Obs, CAM, ZMO and AM2.



(a) CAM

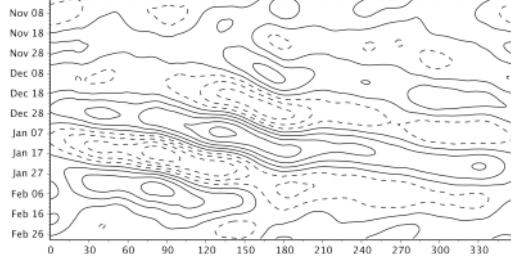


(b) ZMO

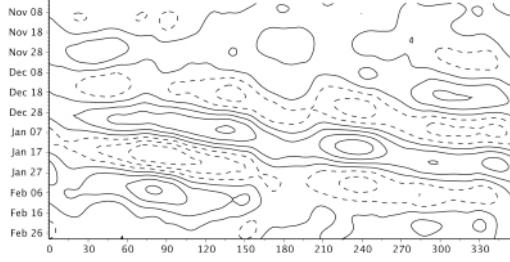


(c) AM2

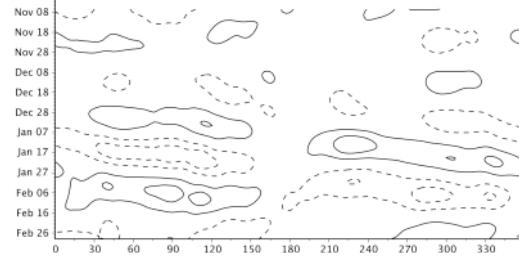
FIG. 10. Zonal wind at 900 hPa for the TOGA COARE IFA. Observational estimate from IFA array budget and (a) CAM forecast Day 3 , (b) ZMO forecast Day 3, (c) AM2 forecast Day 3



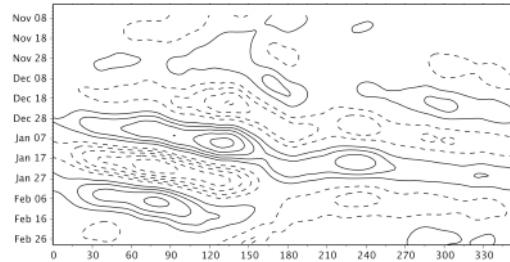
(a) ERA40



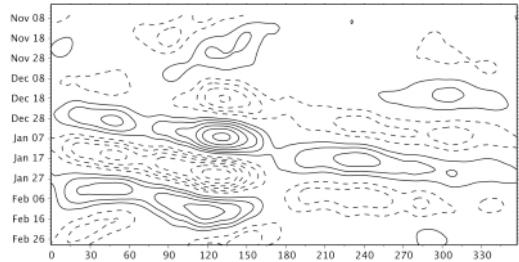
(b) CAM Day 3



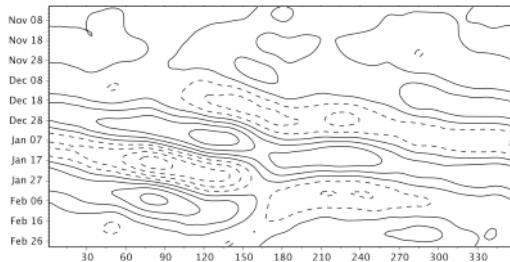
(c) CAM Day 6



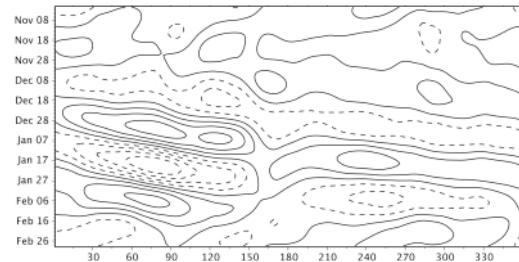
(d) ZMO Day 3



(e) ZMO Day 6

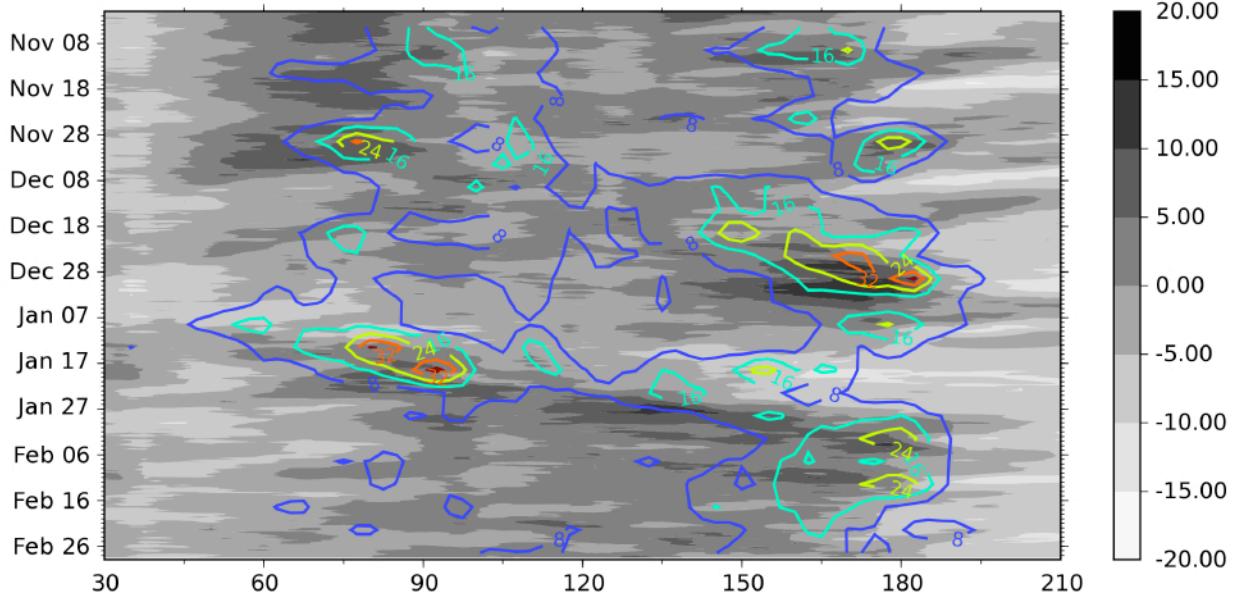


(f) AM2 Day 3

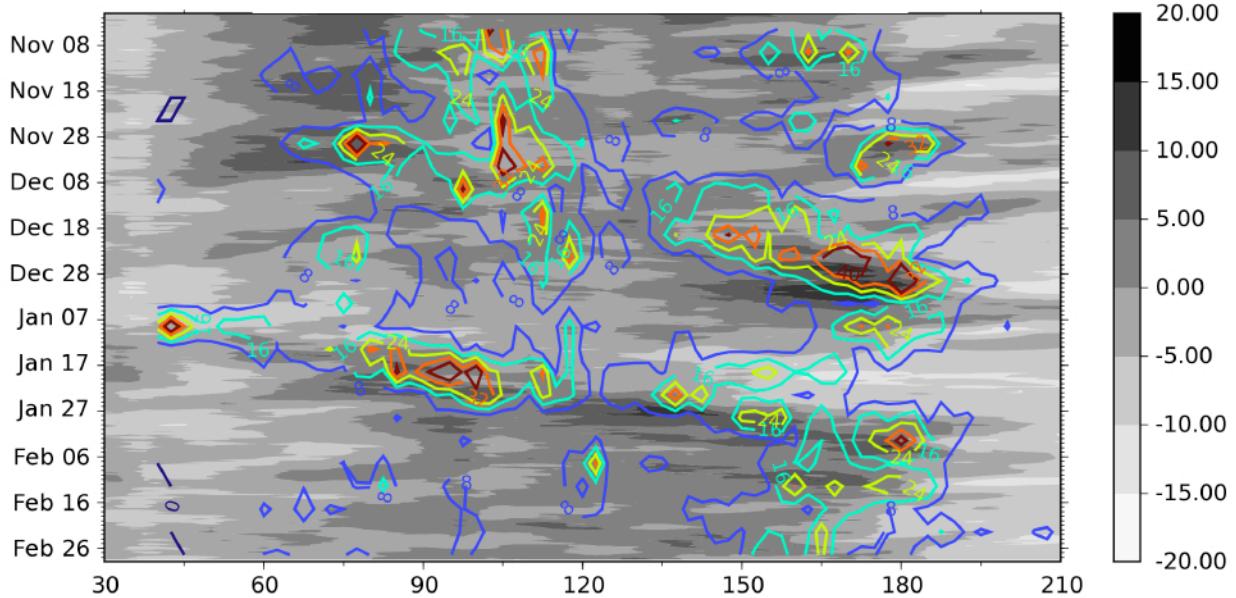


(g) AM2 Day 6

FIG. 11. Velocity potential, band passed filtered 30-70 days, averaged from 5S to 5N. Shown are the ERA40 observations and Day3 and Day 6 of the CM, ZMO and AM2. The contour interval is $2 \times 10^6 \text{ m}^2 \text{s}^{-1}$. Negative contours are dashed.

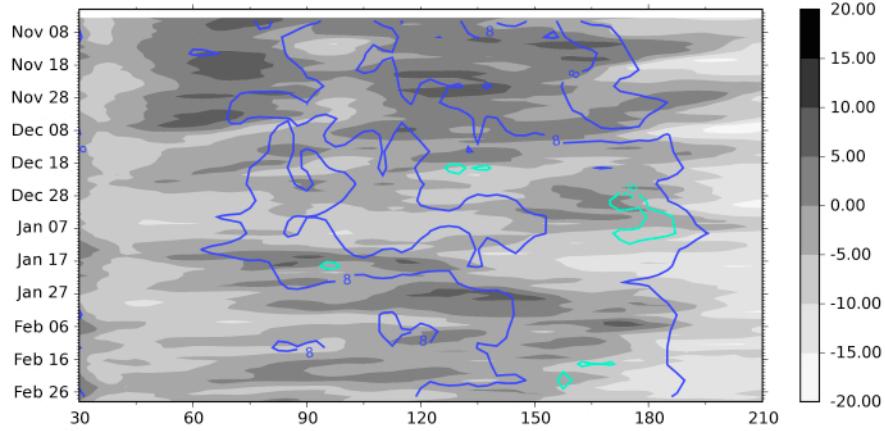


(a) CMAP rainfall, ERA40 850 hPa zonal wind

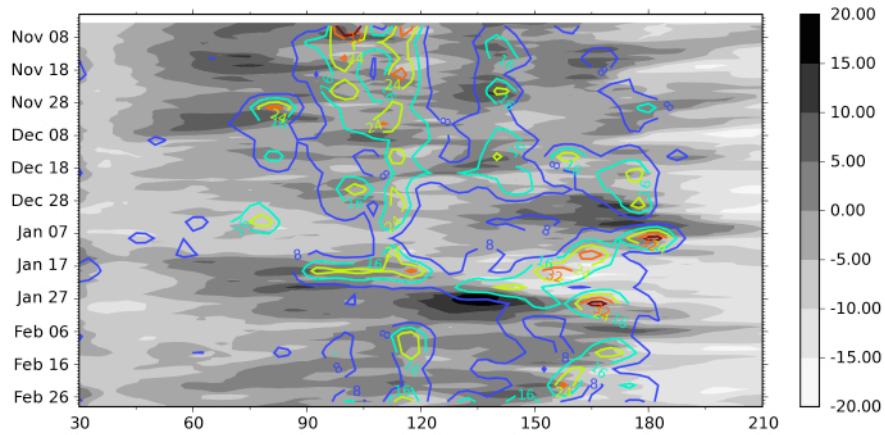


(b) SSMI rainfall, ERA40 850 hPa zonal wind

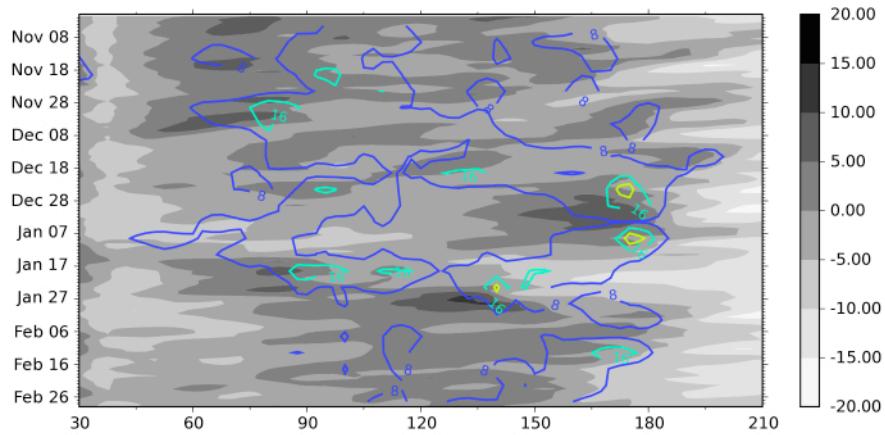
FIG. 12. Contours of CMAP (top) and SSMI (bottom) rainfall (mm/h) and shading of ERA40 zonal wind (m/s) at 850 hPa. The SSMI data have been put on the Pentad times. Contours for rain are in units of 8 mm/h.



(a) CAM



(b) ZMO



(c) AM2

FIG. 13. Contours of rainfall (mm/h) and shading of zonal wind (m/s) at 850 hPa for the (a top) CAM and (middle) ZMO and (bottom) AM2 Day 3 forecasts. Contours for rain are in units of 8 mm/h.

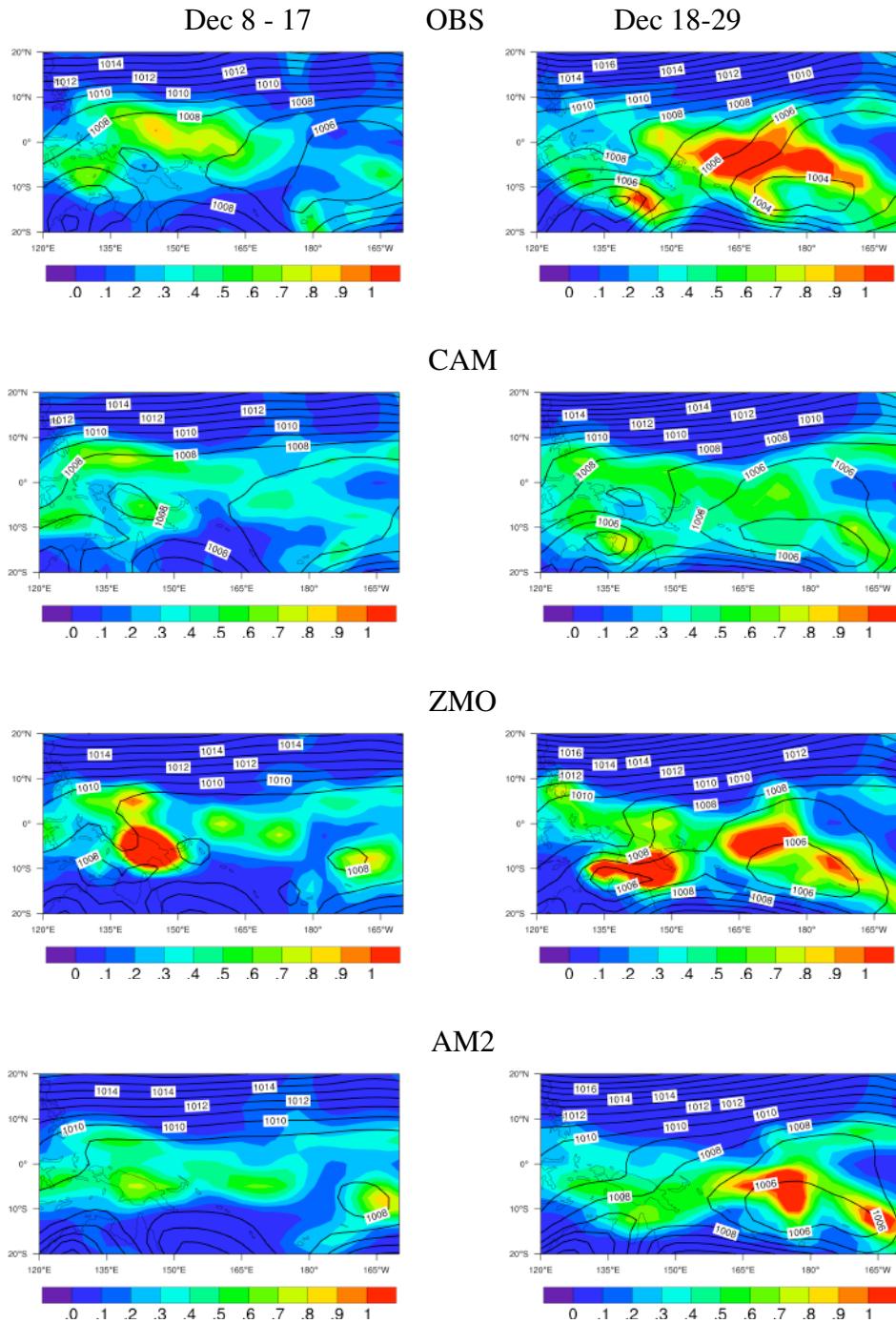


FIG. 14. Rainfall and MSLP for the period 8 Dec to 18 Dec (left column) and 18 Dec to 29 Dec (right column). Observational CMAP estimate (top row), CAM forecast Day 3 (second row from top) , ZMO forecast Day 3 (third row from top) , AM2 forecast Day 3 (bottom row)

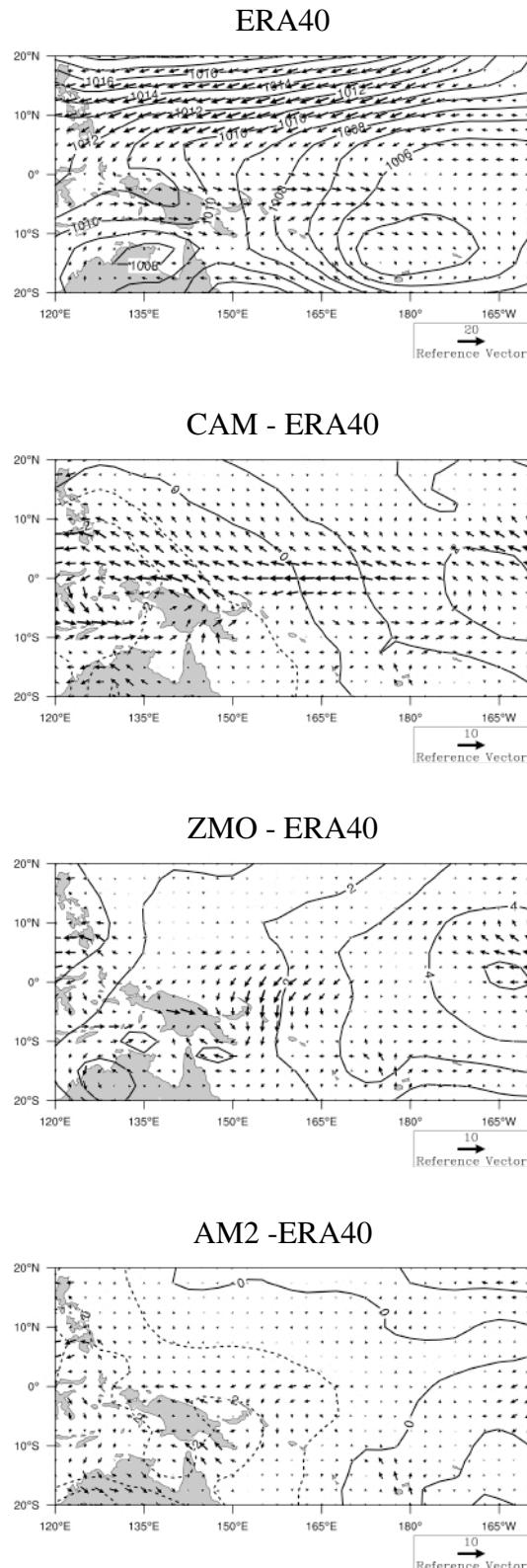


FIG. 15. 850 Winds and MSLP for period 11 Dec to 10 Jan. Observed ERA (top), Difference CAM forecast Day 3 - ERA (second from top) , Difference ZMO forecast Day 3 - ERA (third from top) , Difference AM2 forecast Day 3 - ERA (bottom)

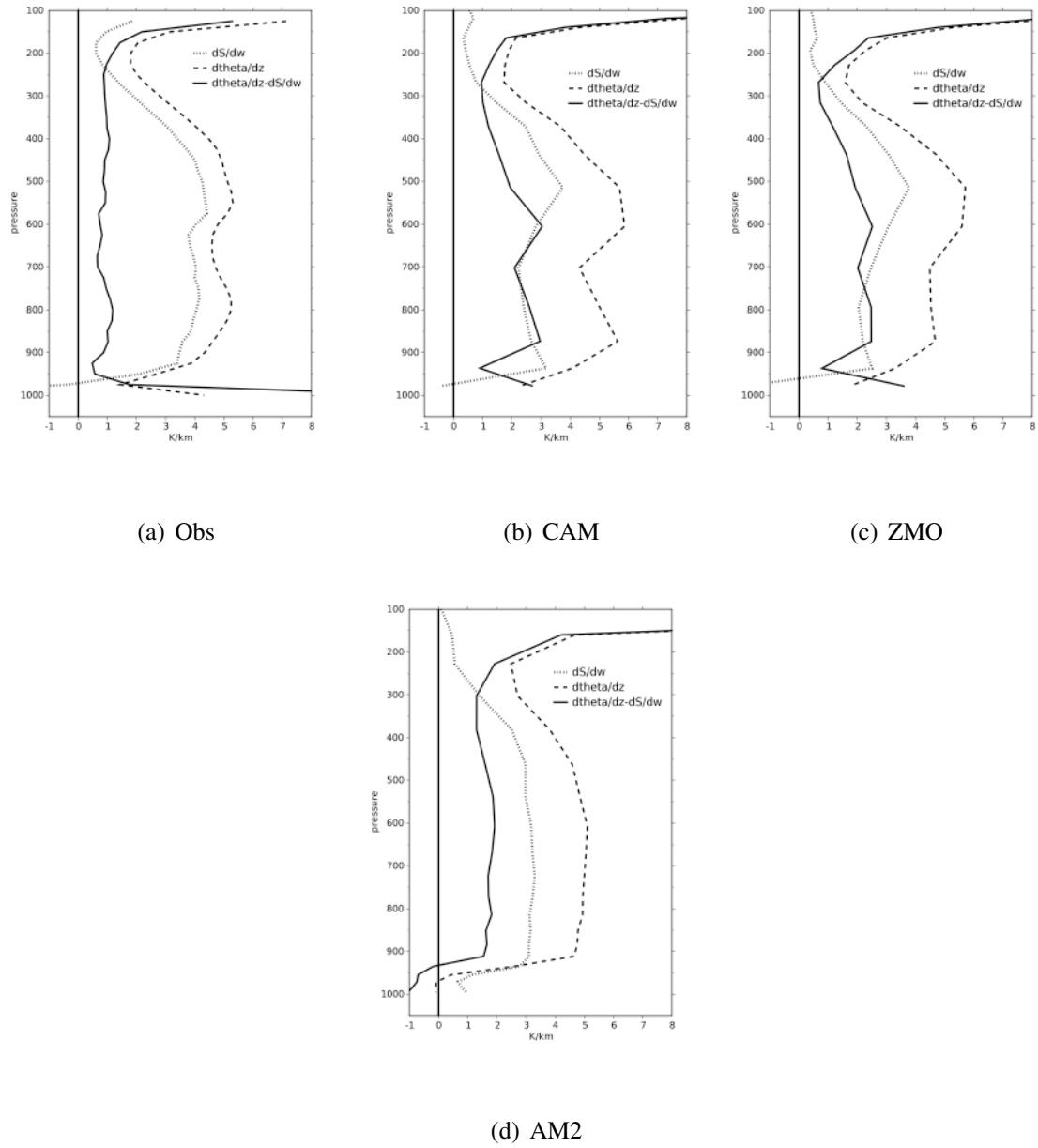
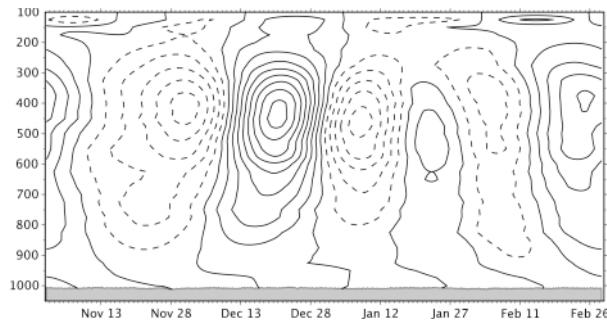
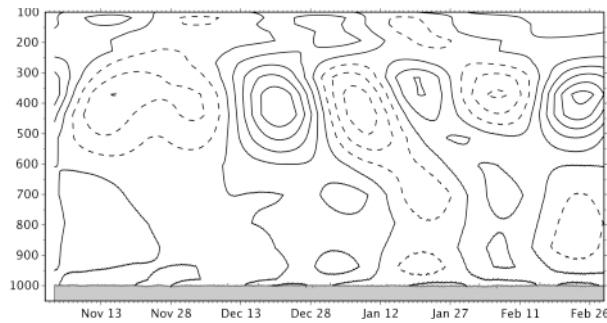


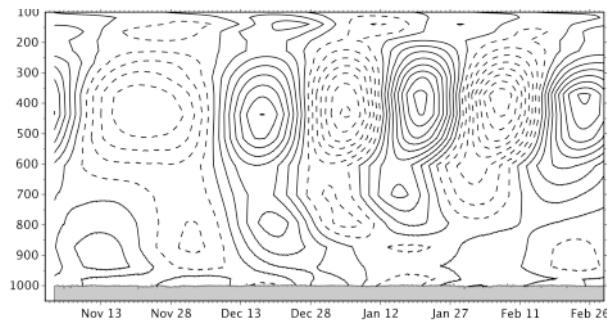
FIG. 16. Profiles of Effective Static Stability, $\Gamma_{eff}(z)$, (solid), $\partial S_\theta/\partial w$ (dotted), and $\partial\theta/\partial z$ (dashed) at the IFA for the (a) Observed, (b) CAM, (c) ZMO and (d) AM2.



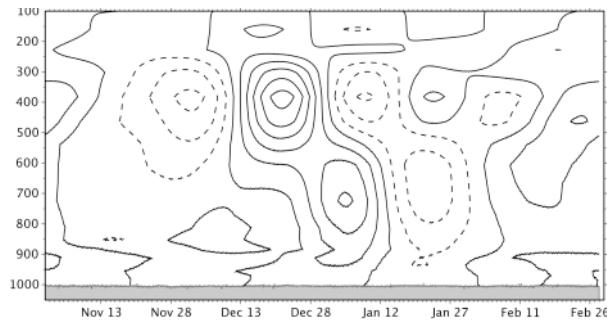
(a) Observations



(b) CAM



(c) ZMO



(d) AM2

FIG. 17. Band Pass (30 - 70 days) apparent heat source (Q1) at the IFA. Observational Estimate (top), CAM forecast Day 3 (second from top) , ZMO forecast Day 3 (third from top) , AM2 forecast Day 3 (bottom). Contours are 0.5 K day^{-1} . Dashed lines are negative. Zero and

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Active	Inactive
3 Nov - 12 Nov 1992	13 Nov - 21 Nov 1992
22 Nov - 27 Nov 1992	28 Nov - 6 Dec 1992
8 Dec - 29 Dec -1992	5 Jan - 16 Jan 1993
15 Jan - 22 Jan 1993	23 Jan - 28 Jan 1993
28 Jan - 1 Feb 1993	2 Feb - 9 Feb 1993
10 Feb - 21 Feb 1993	

TABLE 1. Time periods used for averaging over the Active and Inactive times of the TOGA-COARE IOP.