Evaluation of the eastern equatorial Pacific SST seasonal cycle in CMIP5 models

Z. Y. Song¹, H. L. Liu²,³, C. Z. Wang³, L. P. Zhang⁴, and F. L. Qiao¹

¹First Institute of Oceanography, State Oceanic Administration, Qingdao, China
²Cooperative Institute for Marine and Atmospheric Studies, University of Miami, Miami, FL, USA
³Atlantic Oceanographic and Meteorological Laboratory, NOAA, Miami, FL, USA
⁴Atmospheric and Oceanic Science, Princeton University, Princeton, NJ, USA

Correspondence to: F. L. Qiao (qiaofl@fio.org.cn)

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Abstract. The annual cycle of sea surface temperature (SST) in the eastern equatorial Pacific (EEP) has the largest amplitude in the tropical oceans, but it is poorly represented in the coupled general circulation models (CGCMs) of the Coupled Model Intercomparison Project Phase 3 (CMIP3). In this study, 18 models from CMIP5 are evaluated in terms of their capability of simulating the SST annual cycle in the EEP. Fourteen models are able to simulate the annual cycle fairly well, which suggests that the performances of CGCMs have been improved. The results of multi-model ensemble (MME) mean show that CMIP5 CGCMs can capture the annual cycle signal in the EEP with a correlation coefficient up to 0.9. Moreover, the CMIP5 models can simulate the westward propagation character of the EEP SST – in particular, EEP region 1 (EP1) near the eastern coast leading EEP region 2 (EP2) near the central equatorial Pacific by 1 to 2 months in spring. However, the models fail to reproduce the in-phase SST relationship between EP1 and EP2 in August and September. For amplitude simulations, the model SST in EP1 shows weaker seasonal variation than the observations due to the large warm SST biases from the southeastern tropical Pacific in the boreal autumn. In EP2, the simulated SST amplitudes are nearly the same as the observations while there is the presence of a quasi-constant cold bias associated with poor cold tongue simulation in the CGCMs. To improve CGCM simulation of a realistic SST seasonal cycle, local and remote SST biases that exist in both CMIP3 and CMIP5 CGCMs must be resolved at least for simulating the SSTs in the central equatorial Pacific and the southeastern tropical Pacific.

1 Introduction

The eastern equatorial Pacific (EEP) is a key region for El Niño–Southern Oscillation (ENSO) that affects weather, extreme events and climate globally. Although the equatorial region is dominated by the semi-annual cycle of solar radiation, the sea surface temperature (SST) seasonal variation in the EEP exhibits a strong annual cycle with the March–April warm phase and August–October cold phase (Mitchell and Wallace, 1992; Nigam and Chao, 1996), which is different from the semi-annual cycle in the western equatorial Pacific due to the sun crossing the Equator twice a year. Substantial attention has been devoted to understanding the mechanisms of the SST annual cycle (Mitchell and Wallace, 1992; Giese and Carton, 1994; Xie, 1994, 2004; Mechoso et al., 1995; Nigam and Chao, 1996; Dewitt and Edwin, 1999). Numerous physical processes have been hypothesized to contribute to the annual cycle in the EEP (reviewed by Xie, 2004), such as wind–evaporation–SST feedback, stratus–SST feedback and upwelling–SST feedback. Because the annual cycle in the EEP involves complex dynamical and physical interaction among climate subsystems, it can serve as an indicator for the performance of the coupled general circulation models (CGCMs).

Since the pioneering work of developing a climate model by Manabe and Bryan (1969), CGCMs have achieved significant progress and can provide credible basic climate simulation, particularly through several important climate model intercomparison programs such as the Coupled Model Intercomparison Project (CMIP), the Paleoclimate Modelling Intercomparison Project (PMIP) and the Cloud Feedback...
Model Intercomparison Project (CFMIP). Climate models are important because climate process research and climate projection depend on them. Despite the progress, model biases still persist in the state-of-the-art CGCMs in terms of simulating the SST seasonal cycle in the EEP (Mechoso et al., 1995; Covey et al., 2000; Latif et al., 2001; Xie et al., 2007). Several years ago, De Szoeke and Xie (2008) compared the SST results of 14 global CGCMs in the CMIP Phase 3 (CMIP3), and showed that eight of these models simulated two cold phases in the EEP rather than a single cold phase as observed. For example, the simulation of the Community Climate System Model version 3 (CCSM3), which was one of the CMIP3 models, has a robust SST semi-annual cycle in the EEP. This problem was regarded as one of the six challenges for the further development of CCSMs (Collins et al., 2006). Several studies have reported that the EEP SST annual cycle has been improved in their climate models (Gent et al., 2011; Yu et al., 2013). The model outputs of the CMIP Phase 5 (CMIP5) were released recently, which include the latest version of models participating in CMIP3 and some new models. Therefore, a natural question is how well the SST annual cycle in the EEP is reproduced in CMIP5 models.

This study aims to use the outputs of the CMIP5 historical simulations to evaluate the EEP SST annual cycle simulation in the newest versions of coupled models. A brief description of the models and validation data sets used in this study are presented in Sect. 2. Section 3 presents the simulations of CMIP5 models, and Sect. 4 provides the conclusion and discussion.

2 Models and data sets

In this study, the historical model simulations used are summarized in Table 1. This new generation of CMIP5 CGCMs was from 12 scientific organizations across 10 countries, and the simulations have been submitted to the Program for Climate Model Diagnosis and Intercomparison (PCMDI) for the Intergovernmental Panel on Climate Change Fifth Assessment Report (Taylor et al., 2012). Each historical simulation was integrated from pre-industry spin-up experiment and was
Table 1. List of the CMIP5 models whose output are analyzed in this study.

<table>
<thead>
<tr>
<th>No.</th>
<th>Model</th>
<th>Model center</th>
<th>Reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>CanESM2</td>
<td>Canadian Centre for Climate Modelling and Analysis (Canada)</td>
<td>Chylek et al. (2011)</td>
</tr>
<tr>
<td>2</td>
<td>CCSM4</td>
<td>National Center for Atmospheric Research (United States)</td>
<td>Gent et al. (2011)</td>
</tr>
<tr>
<td>3</td>
<td>CSIRO-Mk3-6-0</td>
<td>Australian Commonwealth Scientific and Industrial Research Organization Marine and Atmospheric Research, and Queensland Climate Change Centre of Excellence (Australia)</td>
<td>Rotstayn et al. (2010)</td>
</tr>
<tr>
<td>4</td>
<td>FIO-ESM</td>
<td>First Institute of Oceanography (China)</td>
<td>Qiao et al. (2013)</td>
</tr>
<tr>
<td>5</td>
<td>GFDL-CM3</td>
<td>Geophysical Fluid Dynamics Laboratory (United States)</td>
<td>Griffies et al. (2011)</td>
</tr>
<tr>
<td>6</td>
<td>GFDL-ESM2G</td>
<td>Geophysical Fluid Dynamics Laboratory (United States)</td>
<td>Dunne et al. (2012)</td>
</tr>
<tr>
<td>7</td>
<td>GISS-E2-H</td>
<td>Goddard Institute for Space Studies (United States)</td>
<td>Shindell et al. (2013)</td>
</tr>
<tr>
<td>8</td>
<td>GISS-E2-R</td>
<td>Goddard Institute for Space Studies (United States)</td>
<td>Shindell et al. (2013)</td>
</tr>
<tr>
<td>9</td>
<td>HadCM3</td>
<td>Met Office Hadley Centre (United Kingdom)</td>
<td>Smith et al. (2010)</td>
</tr>
<tr>
<td>10</td>
<td>HadGEM2-CC</td>
<td>Met Office Hadley Centre (United Kingdom)</td>
<td>Collins et al. (2011)</td>
</tr>
<tr>
<td>11</td>
<td>HadGEM2-ES</td>
<td>Met Office Hadley Centre (United Kingdom)</td>
<td>Jones et al. (2011)</td>
</tr>
<tr>
<td>12</td>
<td>inmcm4</td>
<td>Institute for Numerical Mathematics (Russia)</td>
<td>Volodin et al. (2010)</td>
</tr>
<tr>
<td>13</td>
<td>IPSL-CM5A-MR</td>
<td>Institute Pierre Simon Laplace (France)</td>
<td>Hourdin et al. (2013a)</td>
</tr>
<tr>
<td>14</td>
<td>IPSL-CM5B-LR</td>
<td>Institute Pierre Simon Laplace (France)</td>
<td>Hourdin et al. (2013b)</td>
</tr>
<tr>
<td>15</td>
<td>MPI-ESM-LR</td>
<td>Max Planck Institute for Meteorology (Germany)</td>
<td>Giorgetta et al. (2013)</td>
</tr>
<tr>
<td>16</td>
<td>MPI-ESM-P</td>
<td>Max Planck Institute for Meteorology (Germany)</td>
<td>Giorgetta et al. (2013)</td>
</tr>
<tr>
<td>17</td>
<td>MRI-CGCM3</td>
<td>Meteorological Research Institute (Japan)</td>
<td>Yukimoto et al. (2012)</td>
</tr>
<tr>
<td>18</td>
<td>NorESM1-M</td>
<td>Norwegian Climate Centre (Norway)</td>
<td>Bentsen et al. (2013)</td>
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</table>

3 Results

3.1 Comparison

The annual cycle is one of the major features of SST variability in the EEP (Mitchell and Wallace, 1992; Nigam and Chao, 1996; Xie, 1994, 2004). As shown in Fig. 1, the observed EEP SST reaches the peak at 27°C during March and April and exhibits a minimum of 23°C during September and October. This annual cycle, approximately 4°C in amplitude, is mainly restricted in the region between 110°W and 85°W along the Equator. So, we define the region from 110°W to 85°W and 5°S to 5°N as the EP region 1 (EP1). The annual cycle is almost non-existent to the west of 140°W. Then, we define the transition region (140–110°W and 5°S–5°N) as EP2, where SST shows a much weaker annual cycle compared with that in EP1. The warming in EP1 tends to propagate westward and takes 1-month lead over EP2 in the warm phase (see ERSST in Fig. 1). After June, the variations in EP1 and EP2 are in phase, which suggests that the mechanisms behind the warm and cold phases in the EEP vary with season. Therefore, the two regions of EP1 and EP2 are selected for analysis in this study.

The major characteristics of the SST annual cycle in the EEP are effectively represented in the 14 models classified as Group 1, which includes CanESM2, CCSM4, CSIRO-Mk3-6-0, FIO-ESM, GFDL-ESM2G, GISS-E2-H, GISS-E2-R, HadCM3, HadGEM2-CC, HadGEM2-ES, IPSL-CM5A-MR, IPSL-CM5B-LR, MPI-ESM-LR, and MPI-ESM-P. These model results suggest that the CGCMs in CMIP5 show improvement in capturing the annual cycle of the EEP SST from the previous generation CGCMs of the CMIP3 (which were not effective in simulating the annual SST cycle in the EEP). However, four models classified as Group 2, which includes GFDL-CM3, INMCM4, MRI-CGCM3, and NorESM1-M, produce semi-annual cycle; namely, they are...
still not able to simulate the realistic annual cycle of the EEP SST. Among the models in Group 1, CanESM2, FIO-ESM, GISS-E2-H, GISS-E2-R, HadCM3, and IPSL-CM5B-LR have weaker seasonal variation in the EP1 region in comparison with the observations. In addition, the annual cycle of the EEP SST in CSIRO-MK3-6-0 extends to 180° W, and the westward propagation processes in its EP1 to EP2 are seen in both warm and cold seasons (Fig. 1 CSIRO-MK3-6-0). This implies the westward Rossby-wave propagation mechanism (Xie, 2004) in the warm phase of observations is overexaggerated in CSIRO-MK3-6-0.

The regional average SSTs, shown in Fig. 2, are used to show the seasonal cycle in the EEP. The EP2 SST is observed to follow the EP1 SST variation in March and April; in the following seasons, EP2 SST varies in phase as EP1 when the Pacific cold tongue develops, but with a weaker amplitude (Fig. 2, ERSST). The CMIP5 CGCMs used in the present study are not able to accurately represent this feature. In March and April, 16 out of the 18 models, CanESM2, CCSM4, CSIRO-MK3-6-0, FIO-ESM, GFDL_CM3, GFDL-ESM2G, GISS-E2-H, GISS-E2-R, HadCM3, HadGEM2-CC, HadGEM2-ES, INMCM4, IPSL-CM5A-MR, IPSL-CM5B-LR, MPI-ESM-LR, and MPI-ESM-P, show that EP1 leads EP2 by 1 to 2 months. However in August and September, only two models, FIO-ESM and HadCM3, demonstrate that the SSTs in EP1 and EP2 are in phase. A comparison between the multi-model ensemble (MME) mean and observation reveals that CMIP5 CGCMs can capture the annual cycle signal in both the EP1 and EP2 regions and westward propagation process in spring. In addition, the correlation coefficient between MME and observations can reach 0.9. However, the CMIP5 models fail to reproduce the in-phase SST relationship in EP1 and EP2 in August and September when the cold tongue develops.

It is notable that the SST amplitude of MME (2.7 °C) in EP1 is weaker than that observed (3.8 °C), whereas that in EP2 (2.1 °C) is nearly the same as that observed (2.4 °C). For the MME in EP1, the weaker amplitude is due to the cold bias in the boreal spring at its high peak and the warm bias is up to 1.0 °C in the boreal autumn at its low peak. However, for the MME in EP2, a quasi-constant cold bias is present throughout the year; therefore, its SST amplitude is the same as that observed. To determine the reason behind the difference in amplitudes of MME in EP1 and EP2, we trace the sources of the model deficiencies in the following subsection.
Figure 3 Multi-model ensemble (MME) mean SST for the eastern equatorial Pacific seasonal cycle (a and b) and tropical eastern Pacific SST biases in spring (c) and autumn (d).

3.2 Analysis

Figure 3a shows that in the region of EP1 the SST seasonal cycle of the CGCMs is similar with the observations in terms of phase. However, the MME exhibits cold SST bias during the boreal spring (Fig. 3c) and warm bias (Fig. 3d) during the cooling phase after July. The SST bias in the models reduces the amplitude of EP1 annual cycle. The September–November warm bias in EP1 expands from the southeastern tropical Pacific, where the warm bias exists throughout the year. The model annual cycle in EP2 (Fig. 3b) shows a cold bias within 1°C from January through October. This cold equatorial SST bias is associated with the excessive cold tongue bias, which is a classic tropical bias in CGCMs that still exists in CMIP5 models (Li and Xie, 2014; Wang et al., 2014).

The zonal mean SST bias patterns in EP1 (Fig. 4a) and EP2 (Fig. 4b) show that both regions have cold SST biases in February–May. The cold bias in EP2 extends to the boreal winter. In EP1, however, the warm SST bias develops in August and expands from the southeastern Pacific associated with a southward wind bias (Fig. 4a). In the boreal summer, the southeasterly trade wind prevails in the EP1 region. Therefore, the southward wind bias reduces the southeast trade wind over EP1 and leads to weaker latent heat loss from the ocean, leading to a warm SST bias. It should be pointed that the southward wind bias is associated with the double ITCZ (Intertropical Convergence Zone) bias, which weakens the meridional wind through the wind–evaporation–SST feedback (De Szoeke and Xie, 2008). Furthermore, an easterly wind bias develops in EP1 in August, and this westerly wind bias enhances the EP1 warm bias through the Bjerknes feedback. Eventually, the cold bias in the boreal spring and the warm bias in the boreal summer in EP1 cause the weaker annual cycle simulation in the models, in contrast to the observations. It was suggested that the cold bias in the boreal spring originated in the ocean (Li and Xie, 2012). Therefore, understanding the causes of the warm bias of EP1 in the cold phase will contribute to improvements in the next-generation CGCMs in terms of EEP annual cycle simulations. Moreover, the warm SST bias in the EP1 region in the boreal summer may explain why most CMIP5 models fail to reproduce the in-phase SST relationship between the EP1 and EP2 regions in August and September when the cold tongue develops.

4 Conclusions and discussion

We conducted a comprehensive evaluation of the performance of CMIP5 CGCMs in terms of simulating the annual cycle in the EEP. Fourteen out of the 18 models are able to capture the main characteristics of the SST annual cycle in the EEP. In addition, a comparison between the MME with the observations revealed that the CMIP5 CGCMs can capture the annual cycle signal in both the EP1 and EP2 regions with a correlation coefficient up to 0.9. The CMIP5 models can simulate the SST westward propagation character in the EEP, with SST in EP1 leading that in EP2 by 1 to 2 months in spring. However, they fail to reproduce the in-phase SST
relationship between EP1 and EP2 regions in August and September when the cold tongue develops.

For the MME SST simulation in EP1, both the cold bias along the Equator in the warm phase and the warm bias in the cold phase lead to a weaker annual SST cycle in CGCMs when compared with the observations. In EP2, however, the amplitude is nearly identical to the observed, because a quasi-constant cold bias persisted throughout the year. Known as an excessive cold tongue, this problem is common and still exists in the CMIP5 models (Li and Xie, 2014). To improve the capability of the CGCMs in simulating a realistic SST seasonal cycle in the EEP, both the local and remote climatological SST biases (Wang et al., 2014) that exist in both CMIP3 and CMIP5 CGCMs, such as the climatological simulation of the cold tongue region and the southeastern tropical Pacific, must be resolved.

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