

Responses to reviewer 1

We thank the reviewer for the insightful comments and advices. We have implemented, modified and corrected the text according to your comments and suggestions in the revised version. The followings are our point-by point replies. For convenience, your original comments are in *Italic* and the words marked by red indicate the modified contents from the first revised manuscript.

Major comments

1) It is valuable to show that this model has reached a quasi- equilibrium state. However, I think this is somewhat overemphasized in the current form. The authors mentioned that their strategy is to tune parameters in a coupled model framework, while it should be cautious that the coupled model will eventually obtain a near-zero TOA radiation balance after a certain period of integration no matter how big the initial imbalance is. Therefore, the results (Fig. 2-5) are not quite surprising/informative.

Further, the TOA radiation balance (including OLR in Fig. 6) is strongly coupled to the boundary layer SST. For a fair comparison with observation and other models, the AMIP results with prescribed observed SST should be presented as well. The observed precipitation and OLR are also under current climate not from PI.

We verified the parameter tuning in the AMIP framework after testing them in the coupled model. The AMIP result shows the simulated precipitation, TOA energy and cloud radiative effects are all realistic. In view of the large number of figures that need to be presented, we did not show these figures, but we plan to present the AMIP and historical simulation results later in a separate paper.

2) Following comment #1, I think some AMIP simulation results are desired, which can be directly compared to observation and also very helpful to understand the coupled model behaviors.

Thank you for your good comment. We have implemented more figures (see reply to your point 3) and to avoid excessive lengthy manuscript we will present the AMIP result in other paper.

3) Besides the time evolution of the variables shown in this paper, many other fields are even more important to show so as to have a more complete assessment of the model's

performance, such as atmospheric and oceanic circulation, land surface temperature, annual cycle, diurnal cycle and so on.

Thank you for your good comment. The reviewer #2 also pointed the same problem, so we add the TOA shortwave radiation, cloud radiative effect, zonal mean temperature, zonal wind, specific humidity and sea ice simulation in the manuscript.

Minor comments:

4) Line 73-83: Several statements are not well supported from this paper. The authors should show some figures and cite the results in previous publications. Line 77: 2W/m² is also from PI coupled run?

Thanks for your comments. We have revised the context and also added a reference in this section. The line 73-83 is changed to as follows:

“However, the previous model versions have no vegetation dynamics and cannot be used to study carbon cycle (Cao et al. 2015); in addition, the response of the coupled system to carbon dioxide forcing was over-sensitive; and the poorly resolved vertical layers prevented accurate simulation of stratospheric phenomena as well as upper tropospheric jet stream. They have large land surface temperature biases and a severe double ITCZ syndrome.”

Yes, the 2 W m⁻² TOA energy imbalance is from the PI experiment of NESM v1, while we didn't show the figure here.

5) Line 195: 'longitudinal' to 'horizontal'?

Done.

6) Line 233-235: Most of current climate models have the initial drift issue that is also relevant to the ocean models. Whether this version of NESM3 has the initial drift problem during the spin up period?

Yes, both the initial and the current versions the NESM v3 model have the initial drift problem during the spin up period, while the initial version produced much colder global surface temperature after reaching its quasi-equilibrium state. The PI results in this study are excluded the 400 years coupled model spin up integration.

7) Line 254-257: *I do not understand how these parameters are tuned in a coupled model to obtain a ‘better’ (near-zero) net global mean heat flux budget. The TOA imbalance will decrease with time and eventually will be close to zero. See my major comment #1.*

Thanks for your comments. The tuned parameters would affect the cloud radiative effect, so that the model TOA energy balance could be changed. For example, the increase of ice crystal falling speed will decrease the cloud top height and decrease the OLR. Therefore, it decreases the TOA energy imbalance. We agree with you that the TOA energy imbalance will decrease with model integration extension. But, most of CMIP5 models show that global mean temperature has a linear trend around $0.02\text{K } 100\text{y}^{-1}$ in their PI experiments (Gupta et al. 2013), while the energy analysis show that the TOA energy imbalance is close to 1 W m^{-2} (Fig.3 Fig4 and Table 3 in Wild et al. 2013). To balance the TOA energy, the coupled model may need thousands of years integration. It is hard to do so in our modeling center, so that we tuned the model parameter to decrease the TOA energy imbalance. This is also accepted by other modeling center (e. g. Bentsen et al. 2013; Kay et al. 2016).

8) Line 262-264: *statement without figure support.*

Thanks for your comments. We provided the SST of the original version here. The global averaged annual mean SST is $18.2\text{ }^{\circ}\text{C}$, which is close to the HadISST in the period of 1979-2009 ($18.5\text{ }^{\circ}\text{C}$). However, the bias pattern shows large cold biases over the NH high latitude where covered by the extensive sea ice.

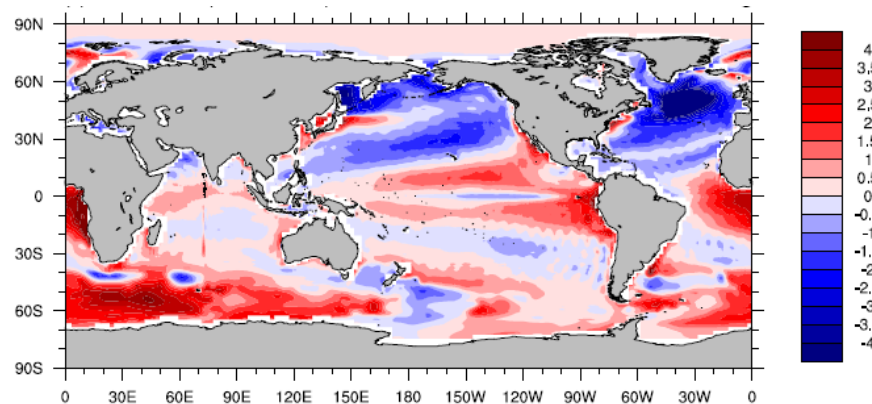


Figure 1 The model SST bias respect to HadISST average from 1979-2008.

9) Line 272: *what are the default configurations?*

Thanks for your comments. We decreased all the albedo parameters by 0.05. That means the visible and near-infrared for sea ice and snow are tuned from 0.78, 0.36, 0.98 and 0.70 to 0.73, 0.31, 0.93 and 0.65, respectively. We add the description in the manuscript as following: “The visible and near-infrared albedos are set to 0.73, 0.31 for ice greater than 0.3 m, and the corresponding cold snow albedos are 0.93 and 0.65, respectively. Those values are slightly smaller than the corresponding default configurations, **which are 0.73, 0.31, 0.93, and 0.65 respectively.**”

10) Line 290: how does the increased deep convective entrainment and convective mass flux induce the reduced zonal wind stress and cold tongue biases? Through convective momentum transport? It should be cautious that entrainment rate is one parameter that affects nearly every aspect of the parameterized convection. Again, the tuning of these parameters should be tested in AMIP simulations before applying to the fully coupled model simulations.

Thanks for your comments. When the entrainment rate is increased, convection is suppressed and shallow or congestus clouds are increased, which reduces vertical temperature gradient in the middle-upper troposphere. The decreased vertical temperature gradient increases subsidence in the cold tongue region. These prevent occurrence of deep convection there, resulting in a reduced wind stress and cold tongue SST bias. (e.g. Watanabe et al. 2011).

11) L306: What is ‘modern’?

The ocean model forcing is from 2000s climatological mean. We changed the ‘modern’ to ‘2000s’ in the manuscript.

“The ocean component model is spun up with **2000s’** atmospheric and sea ice climatological forcings, ... ”

12) Line 345-346: I assume that the net fluxes at TOA and surface are downward positive. The difference suggests that the atmosphere loses energy rather than gain energy, right? I think this is likely due to the dynamic core in the atmosphere model.

Thanks for your comments. Yes, we considered the downward energy flux is positive. In the NESM v3 model, the net solar radiation at TOA is larger than OLR. That means there is a

positive energy imbalance over TOA, which is considered as the anomaly heating. So that we concluded the positive TOA imbalance as the Earth system gains energy. Yes, the artificial energy between TOA and surface is likely due to the dynamic core. And we revised the context correspondingly:

“This problem is found also in the AMIP experiment and it probably due to the energy non-conservation in the model dynamical core.”

13) L379: I think this model tends to underestimate the AMOC strength (14.8 Sv).

Yes, compare to observations, the model is underestimated the AMOC strength. We added the observational value and revised the manuscript correspondingly.

“The mean strength of AMOC is 14.8 sv, which is underestimated comparing to the modern observational value of 18.5 sv (Cunningham et al. 2007).”

14) Figure 10, which 10 CMIP5 models? Whether these 10 models are representative of the CMIP5 models (more than 40)?

Thanks for your comments. We deleted the 10 CMIP5 models' MME in Fig. 10 and directly compared it with other results (e. g. Andrews et al. 2012).

Andrews, T., Gregory, J. M., Webb, M. J., Taylor, K. E.: Forcing, feedbacks and climate sensitivity in CMIP5 coupled atmosphere-ocean climate models. *Geophysical Research Letters*, 39(9), 2012.

15) Line 407-417: Comparison of the coupled model to observation is misleading, as the OLR depends on SST (Fig. 7), and the OLR is likely to be very different from its corresponding AMIP simulation. How about the net TOA radiation bias pattern, and the shortwave absorption pattern that largely represents the cloud simulation in this model? Line 431-434: the formation of the double ITCZ is very complicated and the convective parameterization is only one of them.

Thanks for your comments. We added the TOA shortwave radiation and cloud radiative effect in Fig. 6-9. We agree with you that the formation of the double ITCZ is very complicated. We have no intention to discuss the double ITCZ problem here. We revised the context and add some recently research on the double ITCZ problem. “However, the so-called double-ITCZ precipitation bias exists in the Pacific Ocean and Atlantic Ocean, which is partially linked to

simulated TOA shortwave radiation bias (Xiang et al. 2017) and the insufficient stratocumulus clouds over the eastern Pacific (Bacmeister et al. 2006, Song and Zhang 2009).”

16) Line 452: *fresh water bias?*

Yes, it is fresh water bias. We revised in the context. “Previous studies pointed out that the **fresh water bias** over high latitudes of North Atlantic can weaken ocean convection,”

17) Line 508: *is the positive shortwave clear sky feedback due to ice-albedo feedback?*

Thanks for your comments. The positive shortwave clear sky feedback includes the ice/snow-albedo feedback and also the shortwave contribution of water-vapor feedback (Andrews et al. 2012) .

Andrews, T., Gregory, J. M., Webb, M. J., Taylor, K. E.: Forcing, feedbacks and climate sensitivity in CMIP5 coupled atmosphere-ocean climate models. *Geophysical Research Letters*, 39(9), 2012.

18) Line 512-514: *I do not understand how this is consistent with the conclusion derived from the CMIP5 models?*

Sorry for the confusion. The ECS of NESM v3 model is 10% higher than the CMIP5 MME. The reason maybe the positive cloud radiative effect (CRE) as shown in Fig 12. And the CMIP5 model results suggested the positive CRE model tends to have large ECS (Andrews et al. 2012) .

19) Line 539-549: *It is better to **discuss the TCR in observations** that is about 1.3-1.7K (et al., Otto et al. 2013; Richardson et al.2016). The increased stratiform clouds through the tuning efforts tend to contribute to the overestimated TCR in this model. This deserves some discussion.*

Thanks for your comments. According to *Richardson et al. (2016)* the CMIP5 model shows consistent TCR with modern observational results. The TCR of NESM v3 model is larger than the CMIP5 model MME, and it would be larger than the observational estimation (*Richardson et al. 2016*). In our model tuning cycle, the TCR or ECS aren’t our first priority, so that we consider the tuning is acceptable when the TCR isn’t too far from the CMIP5 MME (e. g. larger than 90% of models) Yes, the model tuning would affect the ECS and TCR as discussed in Sherwood et al. (2014).

Sherwood, S. C., Bony S., Dufresne J-L.: Spread in model climate sensitivity traced to atmospheric convective mixing. *Nature*, 505,37-42, 2014.

20) Line 581: The acronym 'SWT' represents 'Sea Water Temperature'?

Thanks for your comments. It should be "Sea Water Salinity". We changed the SWT to SWS in Line 581. Sorry for the confusion.

"The fresher SSS has no significant influence on **SWS**. After the spin up, the global mean SSS and **SWS** have no appreciable trends although the SSS is fresher than the observed counterpart."

21) Line 595: what do you mean 'slightly'? Better to quantify and compare to other CMIP5 models.

Our model is suffered the double ITCZ problem and we deleted the word 'slightly'. We will quantify the double ITCZ problem in our historical simulation later.

22) For OLR, SST, precipitation, please show their global mean value and also the RMSE.

We added the values in the figures.

23) Some places require references: Line 397, Line 453, Line 514, Line 522.

We added the corresponding references as follows.

"This is in part due to the thinner cloud optical depth in the simulated low-level cloud and shallow mixed layer depth over the Southern Ocean (**Sterl et al. 2012**)."

"Previous studies pointed out that the fresh water bias over high latitudes of North Atlantic can weaken ocean convection, so that weaken the AMOC (**Rahmstorf 1995**)."

"It could be the reason of slightly high ECS of NESM v3 since the CMIP5 model results suggested that the GCM with higher sensitivity is associated with a positive CRE feedback (**Andrews et al. 2012**)."

"The most pronounced warming is seen over the Arctic region where sea ice albedo feedback dominates (**Screen and Simmonds 2010**)."

24) The version 2 of NESM needs some discussion. Whether there is substantial code difference from v1? If not, I think the current version should be called v2 rather than v3.

Thanks for your comments. The version 2 of NESM model is based on the v1 model with improved atmospheric internal modes after convective parametrization modification and tuning. It is aimed at developing to a seasonal prediction system. For this purpose, we consider the seasonal prediction version as v2 and the current version as v3.

Response to reviewer 2

We thank the reviewer for the insightful comments and advices. We have implemented, modified and corrected the text according to your comments and suggestions in the revised version. The followings are our point-by point replies. For convenience, your original comments are in *Italic* and the words marked by red indicate the modified contents from the first revised manuscript.

Major comments:

1) I suggest adding figures for annual mean geographical distribution of TOA shortwave, TOA shortwave cloud radiative effect, TOA longwave cloud radiative effect, and surface air temperature over land. Readers would be interested in observation, model results, and model biases for these variables.

Thanks for your comments. We have added the figures in the revised manuscript according to your suggestions.

2) I suggest adding figures for geographical distribution of sea ice extent in March and September. Readers would be interested in observation and model results.

Thanks for your comments. We have added the figures in the revised manuscript as you suggested. We compared the February sea ice concentration following the CMIP5 suggestion instead of March, and the results are similar.

3) I suggest adding figures for zonal annual mean, latitude-pressure cross section of zonal wind, temperature, and specific humidity.

Thanks for your comments. We have added the figures in the revised manuscript according to your suggestions.

4) I suggest adding a table that summarizes global annual mean values for TOA radiation (net, shortwave, longwave, cloud radiative effect), SAT, and precipitation with respect to the model and observation.

Thanks for your comments. We have added the table in the revised manuscript.

Minor comments:

5) P.3 L.55 "*Project*" Programme?

It is Programme. We revised in the context.

“the World Climate Research Programme (WCRP)”

6) P.6 L.127 "*pseudo wavelengths*" Please explain the definition of the pseudo wavelengths.

We added the explanation in the context.

“The upward and downward irradiance are calculated over a predetermined number of pseudo wavelengths, or g-points, an approach is usually referred to as the correlated-*k* method, where *k* denotes absorption and *g* indexes the cumulative distribution of absorption within a band (Zdunkowski et al. 1980).”

Zdunkowski, W. G., Welch, R. M. and Korb, G. J. : An investigation of the structure of typical two-stream methods for the calculation of solar fluxes and heating rates in clouds. Beitr. Phys. Atmos., 53, 147–166, 1980.

7) P.6 L.128 "*time step for radiation scheme*" "*Frequency of radiation calculation*" maybe a more appropriate expression , because the word "*time step*" has a different meaning.

We changed the "*time step for radiation scheme*" to "*Frequency of radiation calculation*" as you indicated.

“The frequency of radiation calculation is two hours.”

8) P.7 L.129 "*kinetic energy scheme*" Reference to the paper describing the turbulence scheme would be helpful.

Thanks. We added the reference in the manuscript as followings:

“The turbulent transport employs the turbulent kinetic energy scheme (Brinkop and Reockner 1995),”

Brinkop, S. and Roeckner, E.: Sensitivity of a general circulation model to parameterizations of cloud-turbulence interactions in the atmospheric boundary layer. Tellus, 47A,197–220, 1995.

9) P.7 L.134 *"stratiform cloud scheme"* Reference to the paper describing the cloud scheme would be helpful.

We added the reference in the manuscript as followings:

“a cloud microphysical scheme, and a diagnostic cloud cover scheme (Sundqvist et al. 1989).”

Sundqvist, H., Berge, E. and Kristjansson, J. :Condensation and cloud parameterization studies with a mesoscale numerical weather prediction model. Mon. Wea. Rev., 117, 1641–1657, 1989.

10) P.7 L.137 *"Lott and Miller 1997"* I cannot find the paper in the reference list.

We added it in the reference list.

“Lott, F. and Miller, M. J.: A new-subgrid-scale orographic drag parameterization: Its formulation and testing. Quart. J. Roy. Meteor. Soc., 123, 101–127, 1997.”

11) P.9 L.174 *"Hunck and Dukowicz 2002"* I think the correct spelling is "Hunke". I cannot find the paper in the reference list.

We revised the manuscript as follows:

“The sea ice deformation is computed basing on the Elastic-Viscous-Plastic scheme (Hunke and Dukowicz 2002) with”

And

“Hunke, E. C, and Dukowicz, J. K.: The elastic–viscous–plastic sea ice dynamics model in general orthogonal curvilinear coordinates on a sphere—Incorporation of metric terms. Mon. Wea. Rev., 130, 1848–1865, 2002.”

12) P.10 L.195 *"longitudinal"* Horizontal?

Yes, it should be “horizontal”. We change it to “horizontal” in the context.

“The resolution of ocean model is higher than atmospheric model with the **horizontal** resolution of $1^{\circ} \times 1^{\circ}$ in sr and $2^{\circ} \times 2^{\circ}$ in lr version.”

13) P.10 L.206 "Clouds and the Earth's ... and Filled" Please explain what variables in this data set are used for the model evaluation (TOA radiation).

We added the data information in the context.

“(3) **the radative fluxes from edition 2.8** of the Clouds and the Earth's Radiant Energy System- Energy Balanced and Filled (CERES-EBAF, Loeb et al. 2009);”

14) P.10 L.208 "World Ocean Atlas 2009" Please explain what variables in this data set are used for the model evaluation.

We added the data information in the context.

“(4) **the ocean temperature and salinity from** World Ocean Atlas 2009 (WOA09) (Locarnini et al. 2010).”

15) P.17 L.352 "14.2C" Fig.3 shows that the mean surface air temperature is 14.9C.

We revised the number in the context as follows:

“The mean value of the near surface air temperature (TAS) is **14.9 °C** in the entire period, and the linear trend of TAS is $0.00214^{\circ}\text{C} (100\text{yr})^{-1}$.”

16) P.17 L.353 "-0.0021 C(100yr)⁻¹" Fig.3 shows that the linear trend of TAS is +0.00214 C(100yr)⁻¹.

We revised the number in the context as follows:

“The mean value of the near surface air temperature (TAS) is **14.9 °C** in the entire period, and the linear trend of TAS is **0.00214 °C (100yr)⁻¹**.”

17) P.17 L.355 "-0.016 C(100yr)⁻¹" Fig.3 shows that the linear trend of land surface temperature is -0.00984 C(100yr)⁻¹.

We revised the number in the context as follows:

“The linear trend of land surface temperature is $-0.00984\text{ }^{\circ}\text{C (100yr)}^{-1}$.”

18) P.17 L.357 “ $-0.0073\text{ C(100yr)}^{-1}$ ” Fig.3 shows that the linear trend of SST is $+0.00731\text{ C(100yr)}^{-1}$.

We revised the number in the context as follows:

“The negligible SST trend $(0.00731\text{ }^{\circ}\text{C (100yr)}^{-1})$ ”

19) P.18 L.379 “the modern observational value” Readers would be interested in how large is the observational value.

We revised the manuscript as follows:

“The mean strength of AMOC is 14.8 sv, which is underestimated comparing to the modern observational value of 18.5 sv (Cunningham et al. 2007).”

20) P.18 L.395 “recent decade’s observation” Readers would be interested in how large is the recent decade’s observation.

We deleted this sentence since the sea ice coverage is compared with the observation in Fig. 17-18.

21) P.19 L.410 “within the range of uncertainty among different observations” Readers would be interested in what observations are used to estimate the uncertainty range.

?

Thanks for your comment. Loeb et al. (2009) compared the OLR from several observational data suggested the OLR can be vary from 235.2 W m^{-2} to 240.4 W m^{-2} . We added the reference in the manuscript, so that the readers can be referred to the literature.

“differences are within the range of uncertainty among different observations (Loeb et al. 2009).”

Loeb, N. G., B. A. Wielicki, D. R. Doelling, G. L. Smith, D. F. Keyes, S. Kato, N. Manalo-Smith, and T. Wong, 2009: Toward optimal closure of the earth's top-of-atmosphere radiation budget. *J. Climate*, 22, 748–766

22) P.22 L.478 *"quadrupling atmospheric carbon dioxide" Doubling?*

Sorry for the confusion. It should be “doubling”. We revised as follows: “ The ECS is regarded as the global equilibrium TAS change in response to the **doubling** atmospheric carbon dioxide concentration.”

23) P.22 L.479 *"product of the radiative forcing and the climate feedback parameter" Ratio of the radiative forcing to the climate feedback parameter?*

Yes. It is the ratio of the radiative forcing to the climate feedback parameter. We changed the description in the context as follows:” It is also indicated by the **ratio of the radiative forcing to the climate feedback parameter.**”

24) P.23 L.494 *"climate feedback parameter is considered as a constant" There are increasing number of evidence that climate feedback changes according to the surface temperature pattern (e.g., Andrews et al. 2015, J.Climate). Therefore, the sentence here could be refrased as "if we approximate the climate feedback parameter as a constant".*

Thanks for your comment. We changed the description as you indicated. “Since the radiative forcing is logarithmically related to the carbon dioxide concentration **if we approximate the climate feedback parameter as a constant (Hansen et al. 2005),**”

25) P.40 Figure 1 *"JSBACH Dyn. Veg." The main text (L.114) explains that the JSBACH is a land surface model which is not equivalent to the parameterization for the dynamic vegetation.*

Yes, it is the land surface model, and we replotted the Figure 1.

26) P.41 Figure 2 *The legend refers to the labels (a) and (b), which are not shown in the figure.*

Thanks for your comment. We revised the figure 2 legend with more clear indication.

27) P.43 Figure 4 "sea water salinity", "sea water temperature" Are these volume-mean values for the full-depth global ocean?

Yes, the sea water salinity and sea water temperature are both representing the volume-mean values for the full-depth global ocean. We added the explanation of sea water salinity and sea water temperature in the context and figure caption.

28) P.45 Figure 6 "CERES" Which edition of the CERES data is used?

The edition 2.8 of the CERES data is used. We added the description in the figure caption. "The observed OLR filed was derived from the edition 2.8 of Clouds and the Earth's Radiant Energy System (CERES) dataset (Loeb et al. 2009)."

29) P.51 Figure 12 It is difficult to read the numbers and characters in the figure because they are small.

Thanks for your comment. We replotted the Figure 12 with better quality.

1 The NUIST Earth System Model (NESM) version 3:

2 Description and preliminary evaluation

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Abstract

The Nanjing University of Information Science and Technology Earth System Model version 3 (NESM v3) has been developed, aiming to provide a numerical modeling platform for cross-disciplinary earth system studies, project future Earth's climate and environment changes, as well conduct subseasonal-to-seasonal prediction. While the previous model version NESM v1 simulates well the internal modes of climate variability, it has no vegetation dynamics and suffers considerable radiative energy imbalance at the top of the atmosphere and surface, resulting in large biases in the global mean surface air temperature, which limit its utility to simulate past and project future climate changes. The NESM v3 upgraded the atmospheric and land surface model components and improved physical parameterization and conservation of coupling variables. Here we describe the new version's basic features and how the major improvements were made. We demonstrate the v3 model's fidelity and suitability to address the global climate variability and change issues. The 500-year pre-industrial (PI) experiment shows negligible trends in the net heat flux at the top of atmosphere and the Earth surface. Consistently, the simulated global mean surface air temperature, land surface temperature and sea surface temperature (SST) are all in a quasi-equilibrium state. The conservation of global water is demonstrated by the stable evolution of the global mean precipitation, sea surface salinity (SSS) and sea water salinity. The sea ice extents (SIEs), as a major indication of high latitude climate, also maintain a balanced state. The simulated spatial patterns of the ~~mean-outgoing-longwave-radiation~~energy states, SST, precipitation, SSS fields are realistic, but the model suffers from a cold bias in the North Atlantic, a warm bias in the Southern Ocean and associated deficient Antarctic sea ice area, as well as a

delicate sign of the double ITCZ syndrome. The estimate radiative forcing of quadrupling carbon dioxide is about 7.24 Wm^{-2} , yielding a climate sensitivity feedback parameter of $-0.98 \text{ Wm}^{-2}\text{K}^{-1}$, and the equilibrium climate sensitivity is 3.69 K . The transient climate response from the 1pet/year $1\% \text{ yr}^{-1} \text{ CO}_2$ (1pctCO2) increasing CO_2 -experiment is 2.16 K . The model's performance on internal modes and responses to external forcing during the historical period will be documented in an accompanying paper.

1. Introduction

Large internal variability of the Earth climate system involves complex feedbacks among the atmosphere, hydrosphere, cryosphere, land surface and biosphere. As an essential tool to reproduce the Earth's paleoclimate evolution, project future climate change, and understand the mechanisms governing climate variability and change, the Climate ~~system~~ System model ~~Model~~ (CSM) and Earth System Model (ESM) have attracted greatest attention of the scientific community. Starting from 1995, the World Climate Research ~~Project~~ Programme (WCRP) established and regularly organized Coupled Model Intercomparison Projects (CMIPs) (Meehl et al. 2000). The CMIP has not only stimulated the coupled model development, facilitated model output validation, deepened scientific understanding of the Earth climate change, but also provided scientific guidance for the Intergovernmental Panel on Climate Change (IPCC).

The first generation of Nanjing University Information Science and Technology (NUIST) Earth System Model (NESM v1, Cao et al 2015) was established with the atmospheric model ECHAM v5.3, ocean model NEMO v3.4, sea ice model CICE v4.1

and coupler [version 3](#) of the Ocean-Atmosphere-Sea-Ice-Soil Model Coupling Toolkit (OASIS3.0-MCT). It was targeted to meet the demand of seamless climate prediction, simulate the past and project future climate change, and study of climate variability of high-impact weather events. The performances of NESM v1 model have been evaluated (Cao et al. 2015) and further developed into a seasonal prediction system (NESM v2) [by modification and tuning of convective parameterization and cloud microphysics](#). The NESM v1 was also used to study the changes in Last Glacial Maximum climate and global monsoon, demonstrating reasonable model response with external forcing (Cao et al. 2016). Numerical experiments with NESM v2 were conducted to confirm the sources of predictability of the Indian summer monsoon rainfall (Li et al. 2016) and the winter extremely cold days in East Asia (Luo and Wang [2018-submitted](#)).

However, the previous model versions have no vegetation dynamics in the land surface model and cannot be used to study carbon cycle ([Cao et al. 2015](#)); and the response of the coupled system to carbon dioxide forcing was over-sensitive. ~~Another serious problem is the relatively large energy imbalance at the TOA and surface, both have an imbalance of $\sim 2\text{Wm}^{-2}$ in PI experiment. There are also significant biases in the TOA net solar radiation and OLR.~~ Meanwhile, the poorly resolved vertical layers prevented correct simulation of stratosphere phenomena as well as high-level jet stream. They have large land surface temperature biases and a severe double ITCZ syndrome. ~~The SST over high-latitude North Atlantic is excessively cold, the strength and extent of the Atlantic Meridional Overturning Circulation (AMOC) is underestimated, the Southern Ocean is too warm, and the sea ice coverage over the Antarctic is substantially deficient.~~

Facing the forth coming CMIP6, a more comprehensive and improved Earth System Model is needed to perform CMIP6 experiments and to address forcing-related scientific questions. For this purpose, we have developed a new version of NESM v3-0. The major changes include an updated land surface model with dynamic vegetation and carbon exchange, improved shortwave and longwave radiation schemes, new schemes for description of aerosols and computation of surface albedo, increased vertical resolution of the atmosphere model and horizontal resolution of the ocean and sea ice models.

As a registered model of CMIP6, the NESM_v3-0 model is to be used to perform the DECK simulation, historical experiment, and some endorsed MIPs following the CMIP6 experiment design protocol (Eyring et al. 2016). The selected MIPs include: Detection and Attribution Model Intercomparison Project (DAMIP), Scenario Model Intercomparison Project (ScenarioMIP), Decadal Climate Prediction Project (DCPP), Global Monsoons Model Intercomparison Project (GMMIP), Paleoclimate Modelling Intercomparison Project (PMIP), Volcanic Forcings Model Intercomparison Project (VolMIP), and Geoengineering Model Intercomparison Project (GeoMIP).

This paper documents the main features of the NESM v3, the major model improvement, and the preliminary evaluation of model's long term integration and climate sensitivity to carbon dioxide forcing. In the new version 3, the energy balance is substantially improved, including the net shortwave radiation and outgoing longwave radiation and their balance. The biases are in a few tenths Wm^{-2} and the trends are negligible. This is demonstrated by the PI experiment with perpetual unchanged forcing, and the climate sensitivity is tested through the abruptly quadrupling CO_2 experiment and 1pctCO2 gradually 1% CO_2 /year increase experiment.

The model description is presented in Section 2, which is followed by the coupled model tuning strategy (Section 3). In Section 4 and 5, the model long-term stability and the mean climate states are evaluated. Section 6 examines the model climate sensitivity in perturbing atmospheric carbon dioxide concentration. The last section presents a summary.

2. Model description and validation data

The NESM v3 consists of the ECHAM v6.3 atmospheric model, which directly coupled with JSBACH land surface model, the NEMO v3.4 ocean model, the CICE v4.1 sea ice model; and the OASIS3-MCT_3.0 coupler. The model structure is illustrated in Fig.1, and brief description of each component model follows.

2.1 Atmosphere and land surface model

The ECHAM v6.3 and JSBACH model are originally adopted from the Max Planck Institute ECHAM serial model. A brief introduction will be presented here; the detailed documentation can be found in Stevens et al. (2012) and Giorgetta et al. (2013). The ECHAM v6.3 employs the spectral/finite-difference dynamic core for adiabatic process. Calculations of all parameterizations and non-linear terms are transferred to Gaussian grids. A hybrid sigma-pressure coordinate system (Simmons et al. 1999) is used in the vertical discretization. The shortwave and longwave radiation schemes are both from the Rapid Radiation Transfer Model for General Circulation model's (RRTM-G) scheme (Iacono et al. 2008), which takes the two-stream approach. The upward and downward irradiance are calculated over a predetermined number of pseudo wavelengths, or g-points, an approach is usually referred to as the correlated- k method, where k denotes

absorption and g indexes the cumulative distribution of absorption within a band
(Zdunkowski et al. 1980). The ~~time step for~~ frequency of radiation calculation radiation
~~scheme~~ is two hours. The turbulent transport employs the turbulent kinetic energy
scheme (Brinkop and Reockner 1995), and the surface fluxes are calculated using the
bulk-exchange formula which is based on Monin-Obukhov similarity theory. The model
parameterizes shallow, deep and midlevel convection separately. The deep convection is
based on mass-flux framework developed by Tiedtke (1989) and further improved by
Nordeng (1994). Currently, the shallow, deep and midlevel convection are parameterized
by the Tiedtke, Nordeng, and Tiedtke scheme, respectively. The stratiform cloud scheme
contains the prognostic equations for the vapor, liquid, and ice phase, respectively, a
cloud microphysical scheme, and a diagnostic cloud cover scheme (Sundqvist et al. 1989).
The ECHAM v6.3 implements the Subgrid Scale Orographic Parameterization scheme
(Lott and Miller 1997, Lott 1999) to represent the momentum transport arising from
subgrid orograph.

The JSBASH land surface model simulates fluxes of energy, momentum, moisture,
and tracer gases between the land surface and atmosphere (Raddatz et al. 2007). The
JSBACH model contains a 5-layer soil, a dynamic vegetation scheme and a land albedo
scheme. The tiled structure of land surface is divided into eight natural Plant Functional
Types (PFTs), four anthropogenic PFTs and two types of bare surface (Brovkin et al. 2013).
The dynamic vegetation scheme is based on the assumption that the competition between
different PFTs is determined by their relative competitiveness expressed in the annual net
primary productivity, as well as natural and disturbance-driven mortality. The surface

albedo is calculated at each tile of the land surface for near-infrared and visible range of solar radiation.

2.2 Ocean model

The ocean component model of NESM_v3 is Ocean PARallelise (OPA), the ocean part of NEMO v3.4 (Nucleus of European Modelling of the Ocean). The primitive equation of ocean model is numerically solved on an orthogonal curvilinear grid. It uses the isotropic Mercator [projection](#) south of 20°N, and a stretched grid north of 20°N with two poles in Canada and Siberia, which removes the singularity of spherical coordinate in the Arctic ocean and allows the cross polar flow (Madec and Imbard, 1996). The ORCA1 configuration of ocean model corresponds to a resolution of 1 degree of longitude and a variable mesh of 1/3 to 1 degree of latitudes from the equator to pole. It has 46 vertical layers which adopts the z-coordinate with partial steps (Adcroft et al., 1997; Bernard et al., 2006). At the ocean surface, the linear free surface method is used (Roullet and Madec, 2000). Advection of tracer uses the total variance dissipations scheme (TVD) (Zalesak, 1979). Horizontal momentum is diffused with a Laplacian operator and 2-D spatially-varying kinematic viscosity coefficient. The vertical mixing of tracer and momentum is parameterized using turbulent kinetic energy scheme. Besides, the lateral diffusion is solved on the neutral direction (Redi, 1982) and includes eddy-induced advective processes (Gent and McWilliams, 1990). The incoming solar radiation is distributed in the surface layers of the ocean using simplified RGB and chlorophyll-dependent attenuation parameters (Lengaigne et al., 2009). The model uses a diffusive bottom boundary layer (Bechmann and Doscher 1997).

2.3 Sea ice model

The sea ice model in the NESM v3 is CICE v4.1, which is originally developed at the Los Alamos National Laboratory. The model solves dynamic and thermodynamic equations for five categories of ice thickness. The lower bound for the five thickness categories are 0, 0.6, 1.4, 2.4, and 3.6_m, respectively. The sea ice deformation is computed basing on the Elastic-Viscous-Plastic scheme (Hun~~keek~~ and Dukowicz 2002) with the ice strength determined by using the formulation of Rothrock (1975). The ice thermodynamics are calculated at five ice layers corresponding to each thickness category instead of zero-layer thermodynamic option.

2.4 Coupling method with OASIS3-MCT

The coupling method is the same as the previous version of NESM v1, and the detail information is described in Cao et al (2015). But the coupler has been upgraded from OASIS3-MCT to OASIS3-MCT_3.0 (Valcke and Coquart 2015), which is a fully parallelized tool for coupled model. The coupler is used to synchronize, interpolate and exchange the coupling fields among the atmospheric, oceanic and sea ice component models. To conserve the exchange coupling fields, the second order conservation interpolation is used in remapping the energy, mass, momentum, and tracers, so to avoid energy, momentum loss and spurious climate drift. The component models are coupled daily.

2.5 Configuration

Two subversions are included in the NESM v3, namely the standard-resolution version (sr) and low-resolution (lr) version. In the atmospheric model, the sr and lr

versions have a horizontal resolution of T63 and T31, respectively. The T63 corresponds to about 1.9° in meridional and zonal directions. The sr (lr) version has 47 (31) levels in the vertical which extends from the surface up to 0.01 (1.0) hPa. The resolution of land surface model is the same as the atmospheric model. The resolution of ocean model is higher than atmospheric model with the longitudinal-horizontal resolution of $1^\circ \times 1^\circ$ in sr and $2^\circ \times 2^\circ$ in lr version. The resolution in the meridional direction is refined to $1/3^\circ$ and $2/3^\circ$, respectively, over the tropical region. In the vertical direction, the sr (lr) version has 46 (31) vertical layers with the first 15 (9) layers at the top 100 meters. In both sr and lr versions, the sea ice model resolution is about $1^\circ \times 1/2^\circ$ in meridional and zonal directions with four sea ice layers and one snow layer on the top of the ice surface.

2.6 Validation data

To validate the model performance, the following observational data are used: (1) the combined precipitation data of Global Precipitation Climatology Project (GPCP) version 2.2 and Climate Prediction Center Merged Analysis of Precipitation (CMAP) (Xie and Arkin, 1997; Lee and Wang, 2014); (2) Hadley Centre Global Sea Ice and Sea Surface Temperature (HadISST), (Rayner et al., 2003); (3) the land surface temperature from CRU-TS-v3.22 (Harris et al. 2014); (43) the radiative fluxes from edition 2.8 of the Clouds and the Earth's Radiant Energy System- Energy Balanced and Filled (CERES-EBAF, Loeb et al. 2009); (45) the atmospheric zonal wind, temperature and specific humidity from ERA-interim (Dee et al. 2011); (6) -the ocean temperature and salinity from World Ocean Atlas 2009 (WOA09) (Locarnini et al. 2010).

3. Model improvement and tuning

Model sub-grid processes are represented by physical parameterizations. Improvement of physical parametrizations and calibration the parameters within the parametrization schemes using constraints obtained from observation, physical understanding or empirical estimation is an integral part of the model development cycle. Our strategy to improve model performance and tuning parameters includes three elements. First, our principle is that the final tuning of all parameters must be conducted using the fully coupled climate model. Second, to efficiently identify the model's weakness and the effects of the tuning, we designed a standard metrics for evaluation of the model's climatology and major modes of variability, which include total of 160 fields covering the climatology of the atmosphere, ocean, land and sea ice, and internal and coupled modes of variability such as Madden-Julian oscillation (MJO), Arctic oscillation (AO), Antarctic Oscillation (AAO), North Atlantic Oscillation (NAO), global monsoon, El Nino-Southern Oscillation (ENSO), Atlantic Meridional Overturning Circulation (AMOC), Atlantic multidecadal Oscillation (AMO), Pacific Decadal Oscillation (PDO), and major teleconnection patterns etc. Result from each tuning experiment is compared with the corresponding observations when they are available or CMIP5 multi-model ensemble means when observations are not available. This assessment process helps to identify the models' major problems and the consequences of the tuning, and to understand how the tuning works. Third, a low-resolution version model, the NESM v3lr, is developed, which allows integration about four times faster than the standard resolution version so that the tuning experiments can get results quickly. Once the tuning is successful in the low-resolution model, similar tuning is applied to the standard resolution version with necessary resolution-dependent adjustment.

The initial version of the v3 model has considerable trends in the surface air temperature and SST, which is associated with the reduced net solar radiation and outgoing longwave radiation (OLR), as well as a large energy imbalance at the top of the atmosphere (TOA). The global mean surface air temperature (TAS) and SST was about 1 K lower than the observed and suffered a continuing drift. Meanwhile, the sea ice extent and sea ice thickness in both Hemispheres kept increasing in the long-term integration. Our first task was aimed at obtaining a nearly balanced global mean energy at the TOA and surface, as well as a reasonable global mean surface temperature with perpetual pre-industry forcing. This is critical for achieving a stable long-term integration in pre-industry simulation which acts as the benchmark experiment for entry card for CMIP6 (DECK) and historical run as well as some other MIPs. Another tuning consideration is the long-term climatology and internal modes of the Earth System in the current climate condition. Efforts are made to minimize the biases in the simulated SST, sea level pressure (SLP), precipitation, zonal mean temperature and wind, ocean mean state (sea surface salinity, mix layer depth etc.) as well as ENSO, global monsoon, and MJO. In addition, the historical evolution of surface temperature is an important measurement of the model's fidelity. This is along with the abrupt quadrupling and gradually increased 1% yr⁻¹ CO₂ experiments in estimating the model climate sensitivity.

The key tuning parameters in the v3 versions are related to the stratiform cloud, cumulus convection, ocean mixing process, and sea ice albedos. Iterative tunings were conducted in the standalone component models with observed/reanalysis forcing and in the coupled model during the PI control run. To achieve a better global mean radiative energy level and a near zero (within a few tenth W m⁻²) net global mean heat flux budget,

the parameter calibrations are conducted on the relative humidity threshold that is related to cloud forming process and the estimated cloud cover (Mauritsen et al. 2012). The parameters involved in the cloud microphysics are also tuned, including the accretion of cloud water (ice) to rain (snow), auto-conversion rate of cloud water to rain, and ice crystal and rain drop fall speeds, which are recognized as effective parameters in affecting both short and longwave radiation (Mauritsen et al. 2012, Hourdin et al. 2017).

Even with reasonable global mean SST, the model simulated excessive sea ice extent over the Arctic, especially over the Davis Strait, Fram Strait and North Atlantic during winter [\(figure not shown\)](#). The export of sea ice from Davis Strait significantly increases the SST and salinity biases. To mitigate the North Hemisphere sea ice extent bias, the sea ice albedo and ice transport-related parameters were adjusted. Sea ice albedo is one of the most effective tunable parameter to adjust sea ice extent and thickness (Hunckle 2010). The default sea ice albedo parameterization takes into account the radiative spectral band, ice thickness and others. The visible and near-infrared albedos are set to 0.73, 0.31 for ice greater than 0.3_m, and the corresponding cold snow albedos are 0.93 and 0.65, respectively. Those values are slightly smaller than the corresponding default configurations, [which are 0.73, 0.31, 0.93, and 0.65 respectively](#). On the other hand, the sea ice motion is largely driven by the ocean currents, sea surface height gradients and wind stress. The efficiencies of air-ice and ocean-ice drag are important for sea ice transport, as well as sea ice extent during winter and spring (Urrego-Blanco et al. 2016). In this model, the ice surface roughness was decreased and the ocean-ice drag coefficient was increased to decrease the sea ice export over Davis and Fram Strait. This is based on

the understanding that the air-ice and ocean-ice drag parametrizations have large uncertainty in the current CICE model.

Concerning the internal modes, ENSO and Intraseasonal oscillation (ISO) are recognized as the dominate modes on the interannual and intraseasonal time scale, respectively. They significantly influence the tropical and global climate through atmospheric teleconnections. Much attention was paid to improve the simulation of ENSO and ISO in v3.

The ENSO-related SST variability, ENSO phase locking to annual cycle, and the equatorial Pacific cold SST bias are closely related (Ham and Kug 2014). CMIP5 models' results suggested that the models having less cold tongue SST bias reproduce more realistic ENSO phase locking owing to models' simulation of more realistic coupled feedbacks. The change of cloud parametrization has an effect on the mitigation of the cold tongue SST bias, which can lead to an improved ENSO phase locking (Wengel. et al. 2017, 2018). In the NESM v3 model, the parameter of deep convective entrainment and convective mass flux above the buoyance layer have been increased which resulted in a reduced cold tongue bias and zonal wind stress over the equatorial Eastern Pacific, removal of the excessive SST variance over the central Pacific, and improved ENSO phase locking.

The entrainments in deep and shallow convections are associated with the moisture supply in the free atmosphere. Strong convection plumes can increase the water supply for the formation of stratiform clouds, leading to an increase of stratiform precipitation. The interaction between wave dynamics and precipitation heating is essential for the

development and propagation of intraseasonal oscillation (Fu and Wang 2009). The entrainment rates associated with convections are adjusted which allow more stratiform precipitation formed in the coupled model. It strengthens the ISO signal and also significantly enhances the MJO eastward propagation.

4. Model stability under fixed external forcing

The standalone spin-up of ocean and land states is an efficient method to accelerate the spin up process in the coupled model, especially in the PI control simulation. The ocean component model is spun up with 2000s' modern-atmospheric and sea ice climatological forcings, such as radiation, winds, precipitation, sea ice concentration and so on. The offline integration length is 2000 (4000) model years for ocean component of NESM v3sr (v3lr) model. The land surface initial condition is adopted from MPI-ESM-LR model which has active dynamic vegetation and carbon cycle. The initial conditions of the atmospheric and sea ice model in the coupled system used the modern observations. The pre-industry control simulation is performed following the CMIP6 protocol with forcing fixed at the year 1850 or around 1850s. The choice of forcing in time point 1850 or of decadal mean in 1850s forcing is to minimize the initial shock of the ensuing historical simulation. The earth orbital parameters, greenhouse gases, ozone concentration, land surface conditions are fixed at their 1850 values. The solar constant used is the 11 years mean from 1850-1860. The natural tropospheric aerosol and 1850s mean stratospheric aerosol forcing were employed in the coupled system. During the whole PI simulation, there was no land use/land cover change. The coupled model was spun up for 400 years so that the model reached an equilibrium state. After that, a 500 years PI simulation is conducted and evaluated in this study.

One of major purposes of the PI control experiment is to verify the model's stability in the perpetual, unchanged forcing conditions. In this section, emphasis will put on evaluation of the equilibrium state of the top-of-atmosphere (TOA), atmosphere-ocean-sea ice interface to reveal the energy, water, and mass conservation of the whole system. The energy input at the TOA is the major energy source for the Earth System. It is vital to minimize the net energy imbalance at the TOA and surface, which can avoid temperature drift in the system. The major indicators are the land surface temperature and ocean surface temperature; they also work as the direct monitor of system energy conservation. The precipitation is the most important part of global hydrological cycle, which involves the energy exchange, as well as mass exchange among each climate system components. The ocean salinity is sensitive to the state of surface hydrological cycle, land runoff and sea ice melting/formation process. Sea ice extent is a good indicator of sea ice amount in both Arctic and Antarctic regions, and it is sensitive to ocean heat content drift and high latitude energy transfer. To better quantify the climate drifts, linear trends were calculated for all evaluation variables.

The time evolution of global mean energy budget at the TOA, Earth surface and ocean surface are shown in Fig. 2. The global mean net shortwave radiation at the TOA averaged over the 500-year integration is 238.55 W_m^{-2} and the corresponding outgoing longwave radiation (OLR) is 238.39 W_m^{-2} , resulting in a net atmospheric energy gain of 0.17 W_m^{-2} . The net heat budget at the TOA shows a negligible decreasing trend of $-0.0041 \text{ W}_m^{-2}(100\text{yr})^{-1}$. At the Earth surface, the net energy imbalance is 0.31 W_m^{-2} in the whole integration period with an insignificant decreasing trend of $-0.00576 \text{ W}_m^{-2}(100\text{yr})^{-1}$. The negative trends are shown at both the TOA and surface, indicating the

coupled system could lead to a more stable state when the integration extends. Note that there is a difference of 0.14 W m^{-2} between surface and TOA net energy budget, which means the model atmosphere ~~gains~~ produces artificial energy. This problem is found also in the AMIP experiment and it probably due to the energy non-conservation in the model dynamical core.

The trends in the surface temperature indices, namely global mean surface air temperature, land surface temperature and SST, reveal the energy conservation and stability as well as the stability of air-sea-sea ice interaction in the coupled system (Fig. 3). The mean value of the near surface air temperature (TAS) is 14.29°C in the entire period, and the linear trend of TAS is $-0.002140.0021^{\circ}\text{C (100yr)}^{-1}$. This trend is mainly attributed to the land surface temperature rather than SST. The linear trend of land surface temperature is $-0.00984 -0.016^{\circ}\text{C (100yr)}^{-1}$. The slow balance of terrestrial (land) vegetation may be one of the reasons. The global time-mean SST is 17.7°C , which is consistent with the observation measured during the decade of ~~1850-1870-1860~~ 1870-1880. The negligible SST trend ($-0.00731^{\circ}\text{C (100yr)}^{-1}$) indicates the global mean SST reached a quasi-equilibrium state. As the most important component of global hydrological cycle, the global mean precipitation has nearly no trend (Fig. 3). It is of interest that the global mean SST exhibits a long-term variability with a period of 50-100 years in this simulation. Possible mechanism and processes causing this variability will be discussed in a follow-up study.

To further verify the stability of ocean component model, more variables are represented in Fig. 4. At the beginning of the PI experiment (coupled model spin up), the sea surface salinity (SSS) has a quick adjustment process. The global mean SSS is

decreased from 34.6 psu to 34.2 psu in 30 years. After the spin up, the mean value of SSS is 34.2 psu, which is 0.5 psu fresher than the observed value. The long-term trend of SSS is $-0.0077 \text{ psu (100yr)}^{-1}$, which indicates the ocean water flux is maintained at a relatively stable state. Meanwhile, the global mean sea water salinity (SWS) is 34.7 psu with a linear trend of $-0.0038 \text{ psu (100yr)}^{-1}$. The total sea water temperature has an increase trend of $0.032 \text{ }^{\circ}\text{C (100yr)}^{-1}$, this is consistent with the surface energy budget which shows a 0.43 W m^{-2} heating at the ocean surface. Furthermore, the linear trend at the last 100 year is smaller than the first 100 year. The decrease of linear trend implies the model becomes more and more stable during the integration.

Atlantic Meridional Overturning Circulation (AMOC) is a major source of decadal/multidecadal variability of the Earth system, and influences the Arctic sea ice extent variability over Atlantic sector (Mahajan et al. 2011). The time series of the maximum strength of the Atlantic Meridional Overturning Circulation (AMOC) at 26.5°N is evaluated. The mean strength of AMOC is 14.8 sv, which is underestimated comparing to the modern observational value of 18.5 sv (Cunningham et al. 2007). The AMOC strength has a small linear trend and significant multidecadal variability.

The middle and high latitude climate, as well as AMOC, is largely affected by sea ice state and its variability. Following the IPCC report, the February, September and annual mean of Northern and Southern Hemisphere sea ice extents (SIEs) are diagnosed for the entire PI experiment period. The time evolutions of SIEs are plotted in Fig. 5. In the Northern Hemisphere (NH), the annual mean, February and September mean SIE are $11 \times 10^6 \text{ km}^2$, $12.7 \times 10^6 \text{ km}^2$, and $7.58 \times 10^6 \text{ km}^2$, respectively. The trends of SIE over the

NH in the annual mean, February and September mean SIE are $0.039 \times 10^6 \text{ km}^2(100\text{yr})^{-1}$, $0.06 \times 10^6 \text{ km}^2(100\text{yr})^{-1}$, and $0.02 \times 10^6 \text{ km}^2(100\text{yr})^{-1}$, respectively. These trends are small, suggesting that the Arctic SIE maintains a steady state. Over the SH, on the other hand, the trends in the annual mean, February and September mean SIE are $-0.07 \times 10^6 \text{ km}^2(100\text{yr})^{-1}$, $-0.002 \times 10^6 \text{ km}^2(100\text{yr})^{-1}$, and $-0.1 \times 10^6 \text{ km}^2(100\text{yr})^{-1}$, respectively. This indicates that a significant trend exists in the SH September only. The annual mean, February and September mean SIEs are $7.27 \times 10^6 \text{ km}^2$, $1.73 \times 10^6 \text{ km}^2$, and $11.7 \times 10^6 \text{ km}^2$, respectively, ~~indicating the SIEs are even less than recent decade's observation (e.g., 1980-2009).~~ The bias of the SH sea ice extent is related to the extensive solar radiation over the Southern Ocean although the model overestimated cloud cover over there ([figure not shown](#)). This is in part due to the thinner cloud optical depth in the simulated low-level cloud and shallow mixed layer depth over the Southern Ocean (Sterl et al. 2012).

5. Simulated climatology

The climatological mean states of some key fields for energy and water balance obtained from the average results for the last 100-year of the PI control run are compared with observations, including [TOA energy fluxes](#), ~~[Outgoing Longwave Radiation \(OLR\)](#)~~, SST, [land surface temperature](#), precipitation, [atmospheric zonal mean zonal wind](#), [temperature and specific humidity](#), and sea surface salinity. The observed ~~[OLR energy fluxes](#)~~ data covers the period of 2001-2014 and the observed SST is averaged over the period of 1870-1880. [The observational estimate of the land surface temperature is based on 1901-1910 mean of CRU-TS-v3.22.](#) The rest of mean states are derived for the period of 1979-2008.

The observed and simulated annual mean net shortwave (SW) radiation at the TOA and the model bias are shown in Fig 6. The simulated global mean net solar radiation is 238.65 W m^{-2} which is smaller than the observation from CERES-EBAF data (Table 1). The model bias indicates the excessive SW absorption over the ITCZ region and the Southern Ocean, and less SW reflection over the middle latitude oceans that implies the planetary albedo is too high.

Figure 6-7 shows the outgoing longwave radiation (OLR) which is balanced by the TOA net downward solar radiation and represents the atmospheric and cloud top temperature distribution. The global mean OLR is 238.45 W m^{-2} in the model that is close to the counterpart from the CERES data and the differences are within the range of uncertainty among different observations (Loeb et al. 2009). The model simulates well the vigorous deep convection-related low OLR over the Indo-Pacific warm pool as well as the high OLR in the desert and subtropical regions. However, the model overestimates the OLR over the majority of ITCZ, Indo-Pacific warm pool regions, and the off-South American coast region in the South Pacific. The model also underestimates the OLR in the North Atlantic storm track and western part of the Pacific subtropical high regions. These biases arise primarily from the errors in simulated cloud fields.

The cloud radiative effect is defined as the difference between the clear-sky and full-sky radiation. It indicates how cloud affects the radiation budget at the TOA. The simulated SW and longwave (LW) cloud radiative effects (CRE) are compared with the CERES-EBAF ed2.8 in Figure 8 and 9, respectively. The NESM v3 model simulates a global averaged annual mean SW CRE of -48.4 W m^{-2} compare to the observed value of -47.2 W m^{-2} . The simulated LW CRE is 25.98 W m^{-2} which is close to the observed value

of 25.75 W m⁻². The total cloud radiative effect in the NESM v3 is -22.5 W m⁻², this is comparable with the CERES-EBAF observation (-21.45 W m⁻²). The bias pattern of SW CRE is similar to that of the net SW radiation at TOA. The model produces positive SW CRE over the tropics although the simulated cloud cover bias is small (figure not shown). This suggests the importance of cloud vertical distribution and cloud properties in determining the CRE. In addition, the LW CRE bias is smaller than the SW CRE indicating the model has better representation of high cloud.

Figure 7-10 represents the simulated SST and its bias. SST is one of the most important variables in the coupled system which reflects the quality of the model's simulation of atmosphere-ocean interaction processes. The model well captures global distribution of SST with a warm pool in the Indo-Pacific region and the cold tongue over the eastern Pacific. There are warmer biases in the Southern Ocean and off the western coasts of America and Africa, which link to the excessive downward shortwave radiation induced by the negative bias in simulated stratiform clouds. Significant cold SST biases are found in the high-latitude North Atlantic around 50 °N with a maximum negative bias of -4 K. Cold biases are also seen in the subtropical North Pacific and North Atlantic.

The land surface temperatures (LST, Figure 11) are shown in comparison with CRU-TS-v3.22 (1901-1910). The model well reproduces the basic patterns of the LST, including warm continents in equatorial regions and cold continents close to Polar Regions. The simulated global averaged (70 °S-90 °N) LST is 12.72 °C, which is slightly warmer than the observed value of 12.58 °C (Table 1). The warm temperature bias is mainly found over Central Asian, Canadian and Australian Continent.

—Figure 8-12 compares the spatial pattern of observed and simulated precipitation. The simulated precipitation pattern and intensity resemble the observations (pattern correlation coefficient, PCC=0.85), which capture the observed rain bands over ITCZ, South Pacific Convergence Zone (SPCZ), tropical Indian Ocean and the midlatitude storm track regions. However, the so-called double-ITCZ precipitation bias exists in the Pacific Ocean and Atlantic Ocean, which is ~~likely-partially~~ linked to simulated TOA shortwave radiation bias (Xiang et al. 2017)– ~~and the deficiency in deep convective parameterization that results in~~ insufficient stratocumulus clouds over eastern Pacific (Bacmeister et al. 2006, Song and Zhang 2009). The precipitation bias shows a dipole pattern over the tropical Indian Ocean. From an atmospheric point of view, such a model deficiency is mainly attributed to the SST bias over the tropics, but it is essentially a coupled model bias.

The zonal mean climatological temperature, zonal wind, and specific humidity along with their biases with respect to ERA-interim, are presented in Fig. 13-15. Overall, the model captures the temperature, zonal wind and specific humidity distribution reasonably well. The temperature and zonal wind biases are small over majority of the region. However, there exist 6-K cold biases at 200 hPa over high latitudes in both hemispheres (Fig. 13). The biases increase the tropics-to-pole temperature gradient in the upper troposphere, which produced an enhanced subtropical jet. The westerly wind bias is about 6 m s^{-1} in the subtropical jet of both hemispheres and over the equator in the upper-troposphere (Fig. 14). The model simulated less water vapor within the boundary layer while overestimated the specific humidity above the boundary layer (Fig. 15).

—The sea surface salinity (SSS) is an integrated indicator for the hydrological interaction among ocean, atmosphere, land runoff and sea ice, as well as ocean circulation. Accurate simulation of ocean circulation in climate models is essential for correct estimation of the transient ocean heat uptake and climate response, sea level rise, and coupled modes of climate variability. Figure 169 shows the observed climatological SSS and the model bias. In general, the model simulates realistically the high SSS over the subtropics, where precipitation is low and evaporation is high, and the relatively low SSS over the ITCZ region where precipitation is heavy. The global mean SSS has a negative bias of 0.5_psu, which is mainly due to the fresh bias over the North Atlantic and the western equatorial Pacific. Over the western equatorial Pacific, extensive precipitation is the major cause. Over the North Atlantic, the excessive net input of fresh water is a primary cause, which is augmented by weak evaporation at high latitudes. The fresh water bias in the North Atlantic can also be attributed to the bias in simulated North Atlantic Currents and excessive sea ice melt over the Labrador Sea. Previous studies pointed out that the fresh water bias over high latitudes of North Atlantic can weaken ocean convection, so that weaken the ~~Atlantic Meridional Overturning Circulation (AMOC)~~AMOC (Rahmstorf 1995).

The simulated February and September sea ice concentration in both hemispheres are compared with observation in the period of 1870-1880 (Fig. 17, 18). In the NH, the spatial distribution of summer and winter sea ice concentration is well captured by the NESM v3. Over the Southern Hemisphere, the model significantly underestimates sea ice concentration, especially during austral summer. As discussed in the previous section,

there is an extensive solar radiation bias over the Southern Ocean which leads to the warm SST bias, especially during local summer when solar radiation is high.

6. Climate sensitivity to CO₂ forcing

Quantification of climate response to different forcing and estimation of the associated radiative forcing can be benefited from sensitive experiments with a single perturbation forcing, such as an abruptly quadrupling CO₂ (abrupt-4xCO₂) simulation and a 1% yr⁻¹ CO₂ increase (1pctCO₂) experiments. Following the CMIP6 protocols, the two CO₂ experiments are designed to document basic aspects of the NESM v3 model response to greenhouse gas forcing. They are both branched from the PI simulation and the only difference are the imposed CO₂ concentrations. In the abrupt-4xCO₂ experiment, the atmospheric CO₂ concentration is abruptly quadrupled (1139 ppm) with respect to the PI condition (274.75 ppm) in the very beginning of the experiment. The 1pctCO₂ is designed as gradually increasing the CO₂ concentration at the rate of 1% per year. Both experiments were initiated at the end of year 100 of the PI experiments, and each of them was integrated for 150 yrs.

Figure 19 shows the global annual mean surface air temperature (TAS) changes with respect to its mean value in the PI experiment. Once the atmospheric CO₂ instantaneous quadrupling, the radiative forcing defined by the net downward heat flux induced by the changing atmospheric carbon dioxide concentration forces the stratospheric and tropospheric circulations to adjust, thereby changing the surface temperature. The TAS rapidly increases by approximately 4.5 K in the first 20 years in response to the imposed

radiative forcing. After the rapid initial increase, the TAS gradually increases, mitigating the energy imbalance at the TOA.

The abrupt 4 x CO₂ experiment is used not only to diagnose the fast response of the Earth system, but also to quantify the radiative forcing, as well as to estimate the Equilibrium Climate Sensitivity (ECS). The ECS is regarded as the global equilibrium TAS change in response to the ~~quadrupling-doubling~~ atmospheric carbon dioxide concentration. It is also indicated by the ~~product-ratio~~ of the radiative forcing ~~and to~~ the climate feedback parameter. The regression of TOA energy imbalance and global mean TAS change is an effective method to obtain those estimations (Gregory et al. 2004), ~~since-It~~ it doesn't require the equilibrium state of GCM. The intersection of regression line and the y-axis is recognized as the adjusted radiative forcing, and the intersection on the x-axis is an indication of the equilibrium temperature. The slope of the regression line is the climate feedback parameter.

The relationship between the change in the net TOA energy imbalance and global mean TAS change is plotted in Fig. ~~2011. Figure 11~~ It shows that the TOA radiative imbalance is around 7.24 W m⁻² when the assumed global TAS is unchanged, although the radiative forcing is affected by the rapid adjustments of stratosphere in the first year and therefore reduced the effective radiative forcing (Gregory and Webb 2008). To balance the net TOA energy, the regression predicted an equilibrium temperature change of 7.38 K in this model, yields a climate feedback parameter of -0.98 Wm⁻²K⁻¹. Since the radiative forcing is logarithmically related to the carbon dioxide concentration ~~if we approximate the climate feedback parameter as a constant-and the climate feedback parameter is considered as a constant in a giving model, even in the presence of other~~

556 ~~forcing agents~~ (Hansen et al. 2005), this gives the ECS of 3.69_K. Andrews et al. (2012)
557 found that the CMIP5 ensemble mean of regressed $4\times\text{CO}_2$ adjusted forcing is 6.89 ± 1.12
558 W m^{-2} , and the climate feedback parameter is $-1.08\pm 0.29 \text{ Wm}^{-2}\text{K}^{-1}$, with the ECS of
559 $3.37\pm 0.29 \text{ K}$. The carbon dioxide-induced radiative forcing and climate feedback
560 parameter estimated by the NESM v3 model are comparable with CMIP5 model
561 ensemble, albeit the estimated ECS is about 10% higher.

562 The climate sensitivity parameter consists of the longwave clear sky, shortwave clear
563 sky, longwave cloud forcing and shortwave cloud forcing terms. They are defined by the
564 heat flux differences between the abrupt- $4\times\text{CO}_2$ experiment and PI experiment. The sum
565 of the longwave cloud forcing and shortwave cloud forcing is the total CRE. Here the
566 downward fluxes are defined as positive. Figure 2142 shows the relationships between
567 the changes in the global mean heat fluxes and the change in the surface air temperature.
568 The longwave clear sky feedback strength is $-1.63 \text{ Wm}^{-2}\text{K}^{-1}$, which is partially offset by
569 the shortwave clear sky feedback ($0.68 \text{ Wm}^{-2}\text{K}^{-1}$), resulting in a residual feedback
570 strength of $-0.95 \text{ Wm}^{-2}\text{K}^{-1}$, which is close to the climate sensitivity parameter estimated in
571 this model ($-0.98 \text{ Wm}^{-2}\text{K}^{-1}$). The slopes of the shortwave and longwave cloud forcing
572 have nearly the same magnitude but with opposite signs, yielding a small positive cloud
573 radiative effect ($0.02 \text{ Wm}^{-2}\text{K}^{-1}$) in this model. It could be the reason of slightly high ECS
574 of NESM v3 since the CMIP5 model results suggested that ~~The result here is consistent~~
575 ~~with the conclusion derived from the analysis of the CMIP5 models, that is, a~~ the GCM
576 with higher sensitivity is associated with a positive CRE feedback (Andrews et al. 2012).
577 ~~And the~~ The CRE is a major contributor to the uncertainty in climate sensitivity parameter

in CMIP3 and CMIP5 models, although its magnitude is small compared to other flux terms (Webb et al. 2006, Andrews et al. 2012).

Figure [22+3](#) displays the global distributions of temperature and precipitation in response to the quadrupling CO₂ forcing, which are defined by the departure of the last 30-year climatology from the corresponding climatology in the PI experiment. The most pronounced warming is seen over the Arctic region where sea ice albedo feedback dominates ([Screen and Simmonds 2010](#)). The relative small temperature change is over the Southern Ocean and North Atlantic. The warming is more significant over land than ocean, especially in the Northern Hemisphere. The mean surface temperature over land and ocean are 8.0_K and 5.2_K, respectively. The equatorial Pacific shows an El Nino-like warming. The zonal mean surface temperature change shows an obvious polar amplification, especially over the Arctic Ocean; and stronger warming over the NH high latitudes and weak warming in the SH middle latitudes. The Large NH temperature increase is attributed to the strong warming over the Arctic Ocean and the large land area in the NH.

A direct consequence of global warming is the rising atmospheric specific humidity and precipitation. The global mean precipitation is increased from 2.87 mm day⁻¹ to 3.12 mm day⁻¹, resulting in a precipitation increase of 1.4% per Kelvin global warming. Significant precipitation increases are seen in the equatorial Pacific and Northern Indian Ocean as well as along the Pacific storm track (Fig. [22+3](#)). Decreased precipitation is evident in the sub-tropical descent zones. Note that precipitation is decreased over the Amazon region, where the model has a dry bias in climatology. The global distribution of

precipitation change appears to be dominated by the wet-get-wetter (~~Held and Soden,~~
pattern (Held and Soden, 2006).

In reality, the CO₂ increase is gradually rather than abrupt. The 1pctCO₂ experiment is designed to examine the transient climate response (TCR), which is calculated by using the global mean TAS change between the averaged 20-year period centered at the timing of CO₂ doubling (year 60-80 in 1pctCO₂ experiment) and the PI experiment. The time evolution of the global mean TAS anomalies with respect to the PI experiment is shown in Fig. ~~23~~¹⁴. A linear increase of temperature anomalies is presented in the gradually CO₂ increasing experiment. The temperature anomalies averaged between year 60 and 80 are 2.16_K. This value of TCR is significantly small than the ECS, demonstrating that the ocean heat uptake delays surface warming. The estimation from CMIP5 models shows that the mean TCR is 1.8±0.6_K (Flato et al. 2013), implying that the NESM v3 is comparable to other CGCMs.

6. Conclusion

The development of version 3 of the Nanjing University of Information Science and Technology (NUIST) Earth System Model (NESM v3) aims at building up a comprehensive numerical modeling laboratory for multi-disciplinary studies of the Climate System and Earth System. As a subsequent version of NESM v1, it has upgraded the atmospheric and land surface models, increased the ocean model resolution, improved coupling conservation and modified model physics.

The NESM v3 couples the ECHAM v6.3 atmospheric model, JSBACH land surface model, NEMO v3.4 ocean model, and CICE v4.1 sea ice model by using OASIS3-

MCT_3.0 coupler. The improvement of model physics mainly focuses on convective parameterizations, cloud macrophysics and microphysics, and ocean-sea ice coupling. The model physics modifications and parameters adjustments are targeted at (1) obtaining stable long-term integrations and reasonable global mean states under the preindustrial (PI) forcing, (2) mitigating the biases in the mean climatology and internal modes of climate variability with respect to the modern observations in the present-day forcing condition, and (3) simulating reasonable climate responses to transient and abrupt CO₂ forcing.

A 500-yr PI experiment is conducted and analyzed to test the model's computational stability. As shown in Sec. 4, the long-term climate drifts in NESM v3 are generally negligibly small, especially in the global radiative energy and temperature. The simulated net downward energy flux at the TOA and surface are 0.17 Wm⁻² and 0.35 Wm⁻², respectively. The near-equilibrium model long-term temperature evolution is benefited from the near-zero energy imbalance and negligibly small trends in the energy balance. The global mean near surface air temperature is 14.92°C with a trend of ~~-0.0096~~0.00214 °C (100yr)⁻¹. The linear trends of the land surface and sea surface temperature are -~~0.016~~0.00984°C (100yr)⁻¹ and ~~-0.0026~~0.00731°C (100yr)⁻¹, respectively. However, the total sea water temperature has a warming trend of 0.03°C (100yr)⁻¹, which can be explained by the small but persistent positive downward energy flux into the ocean. The stable long-term evolutions of precipitation, sea surface salinity (SSS) and sea water salinity (SWS) demonstrate the conservation of global water. At the beginning of PI experiments spin up, there was a freshening trend in SSS, which is associated with the ocean adjustment. The fresher SSS has no significant influence on ~~SWTSWS~~. After the

spin up, the global mean SSS and ~~SWT-SWS~~ have no appreciable trends although the SSS is fresher than the observed counterpart. The Northern Hemispheric annual mean, February, and September mean SIEs maintain a steady value at $11.4 \times 10^6 \text{ km}^2$, $13.4 \times 10^6 \text{ km}^2$, and $7.78 \times 10^6 \text{ km}^2$, respectively. However, the simulated Southern Hemisphere SIEs are less than present-day observation. The conservation properties of NESM v3 are encouraging, fulfilling a highly desirable constraint for climate models aiming for multidecadal, centennial and longer simulations.

The last 100-year results are compared with the available observations as presented in Table 1. The TOA energy budget and cloud radiative effect are attracted more attention since its importance in understanding the climate change. The model results show a realistic global climate, although the bias of energy state still exists, especially over Indo-Pacific region, which may be related to the treatment of cloud and convection parameterization. The model simulates realistic OLR pattern, although it overestimates OLR over the ITCZ and Indo Pacific warm pool regions as well as off the South American coast in the South Pacific whereas underestimates the OLR in the North Atlantic storm track and western Pacific subtropical high regions. The annual mean SST/LST is well produced in the model, but large cold biases exist in the North Atlantic and significant warm biases in the Southern Ocean, and warm temperature bias over the central Asian. The biases in OLR and SST are primarily associated with the errors in the simulated cloud fields except for the North Atlantic cold bias. The simulated mean precipitation is reasonably realistic, but slightly suffers the double ITCZ syndrome. The fresh bias in SSS in the tropical western North Pacific can be attributed to the extensive precipitation and the fresh bias over the mid-latitude North Atlantic is related to

underestimated evaporation. The sea ice coverage is well reproduced by the model over the Arctic in February and September; however, it is underestimated over the Antarctic where SST has a warm bias.

The model produces a radiative forcing, under the abrupt quadrupling carbon dioxide, of 7.24 W m^{-2} with a climate feedback parameter of $-0.98 \text{ Wm}^{-2}\text{K}^{-1}$, yielding a warming of 7.38 K at the estimated equilibrium state. The transient climate sensitivity is 2.16 K which is estimated from the $1\% \text{ yr}^{-1} \text{ CO}_2$ gradually increasing experiment. The NESM v3 model is amongst the more sensitive side of the CMIP5 class of global climate models.

At last, this paper isn't aimed at providing a comprehensive evaluation of all model aspect. The NESM v3 model'sIts response to given SST forcing in AMIP and the historical forcing in the coupled model, andthe corresponding modern climatology, internal and coupled modes of climate variability, as well as regional climate variability will be discussed in detail in an accompanying paper later.

Code availability

Please contact Jian Cao (Email: jianc@nuist.edu.cn) to obtain the source code and data of NESM v3.

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928 Table 1 Summery of the global averaged annual mean values for radiation, temperature and
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Figure 6 Annual mean TOA net shortwave radiation (units: W m^{-2}) derived from observation (top), the model simulation in the PI experiment (middle) and the model bias (bottom). The observed radiation field was derived from the Clouds and the Earth's Radiant Energy System (CERES) dataset (Loeb et al. 2009).

Figure 7 As in Fig. 6 except for OLR.

Figure 8 As in Fig. 6 except for TOA shortwave cloud effect.

Figure 9 As in Fig. 6 except for TOA longwave cloud effect.

Figure 11 As in Fig. 10 except ~~The same as Figure 10, but~~ for land surface temperature (°C). The observed land ~~surface climatology~~ surface climatology was derived from the CRU-TS-v3.22 (Harris et al. 2014) for the period of 1901-1910.

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Figure 12 The same as in Fig. 6 except for annual mean of precipitation (mm day⁻¹). The observed precipitation was derived from a Merged precipitation dataset (Lee and Wang 2014), which is the arithmetic mean of the monthly data from the Global Precipitation Climatology Project (GPCP) version 2.2 (Adler et al., 2003) and Climate Prediction Center Merged Analysis of Precipitation (CMAP, Xie and Arkin, 1997).

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Figure 22 Changes in the surface temperature (top) and precipitation (bottom) derived from the last 30-year climatology in the 150-year abrupt 4 x CO_2 experiments. The changes are with reference to the corresponding climatological mean fields from the PI experiment. The right panels show the corresponding zonal mean changes.

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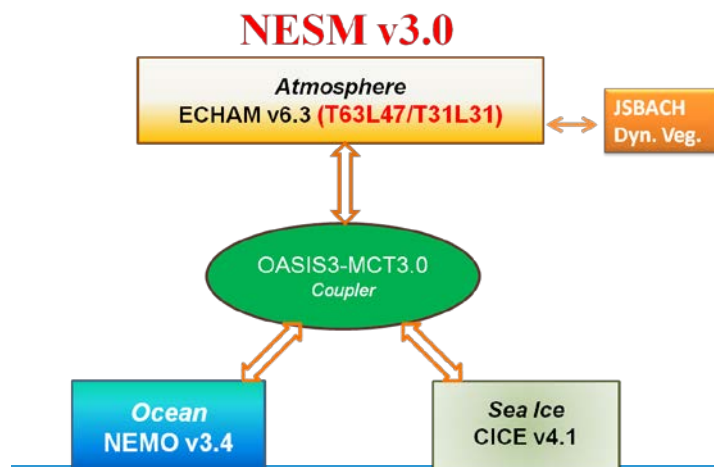
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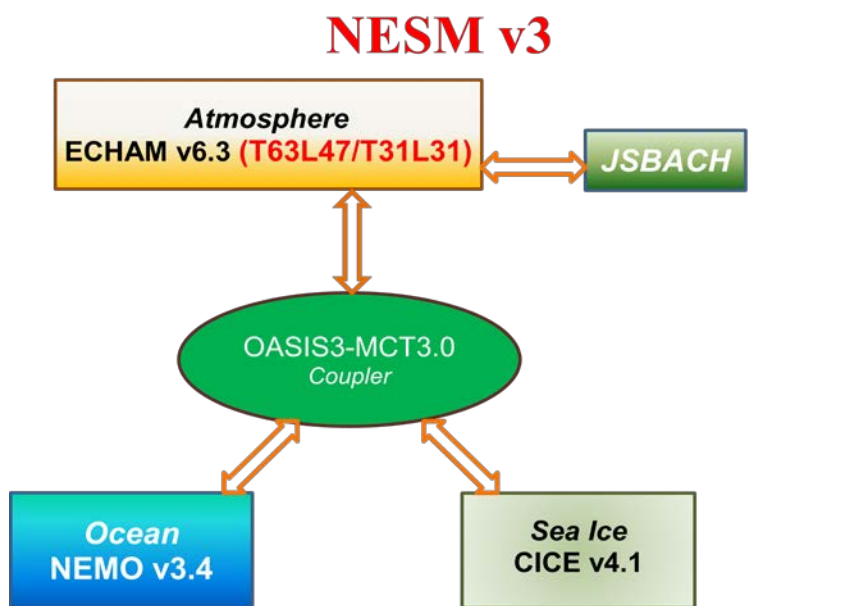
	<u>TOA net</u>	<u>TOA SW</u>	<u>OLR</u>	<u>SW CRE</u>	<u>LW CRE</u>	<u>SST</u>	<u>LST</u>	<u>PR</u>
<u>Obs</u>	<u>0.83</u>	<u>240.51</u>	<u>-239.68</u>	<u>-47.16</u>	<u>25.98</u>	<u>17.2</u>	<u>12.58</u>	<u>2.68</u>
<u>NESM v3</u>	<u>0.2</u>	<u>238.65</u>	<u>-238.45</u>	<u>-48.44</u>	<u>25.75</u>	<u>17.7</u>	<u>12.72</u>	<u>2.86</u>

Table 1 Summery of the global averaged annual mean values for radiation (Unit: W m⁻²),
temperature(Unit: °C) and precipitation (mm day⁻¹) from last 100-year PI simulation and
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1106 Figure1.Coupled structure of NESM v3 model.

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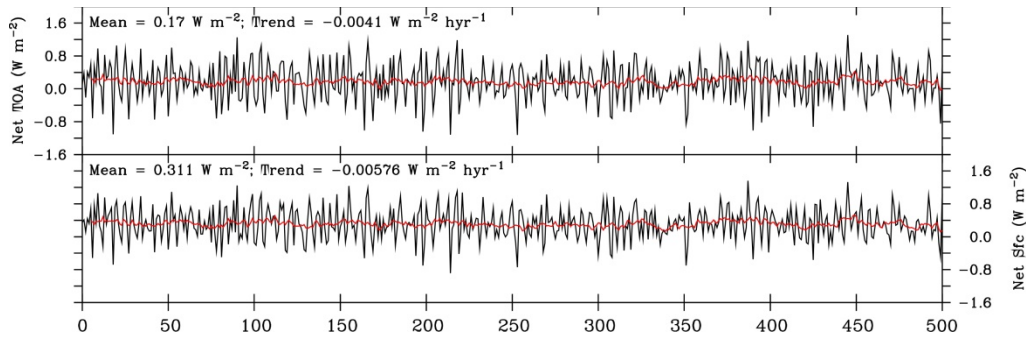


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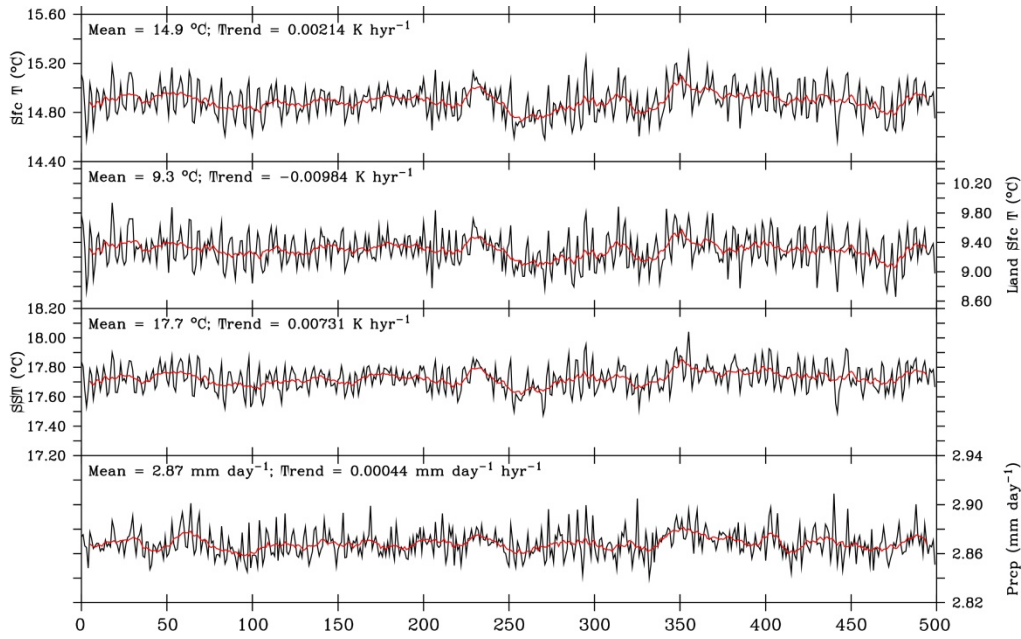


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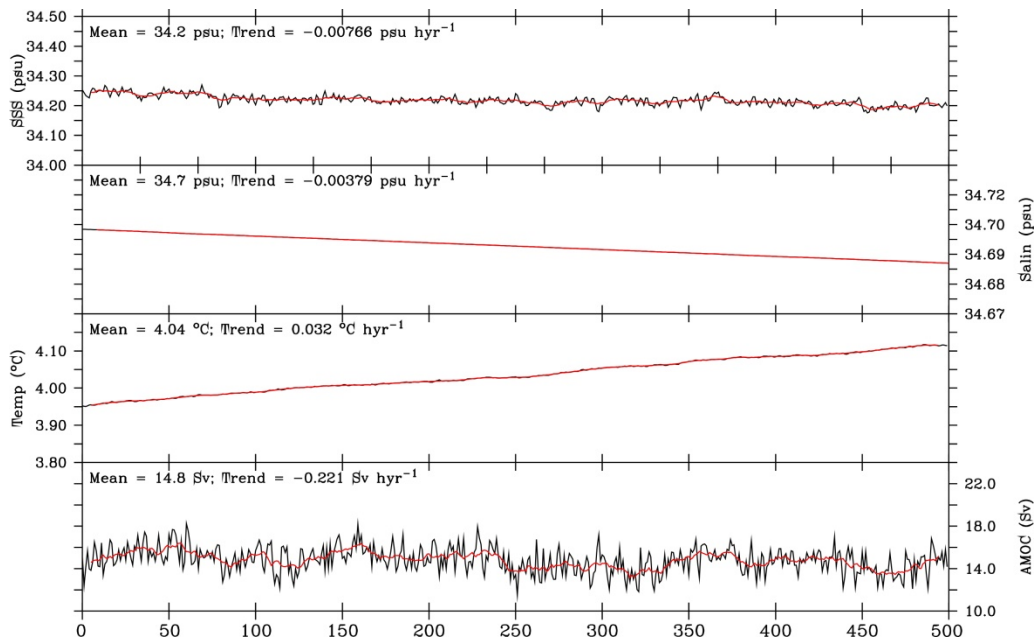


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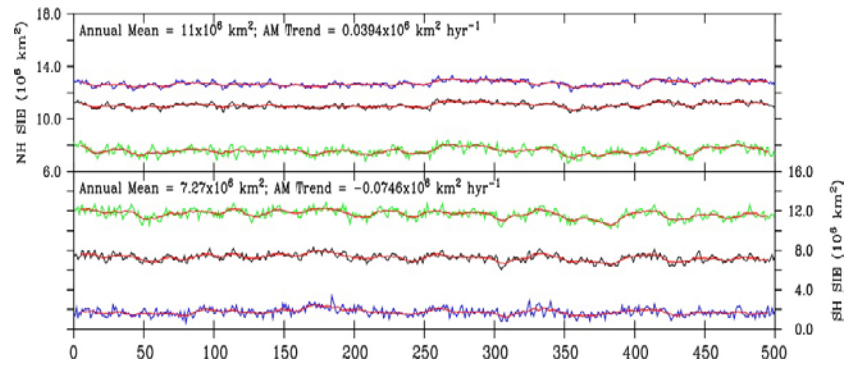


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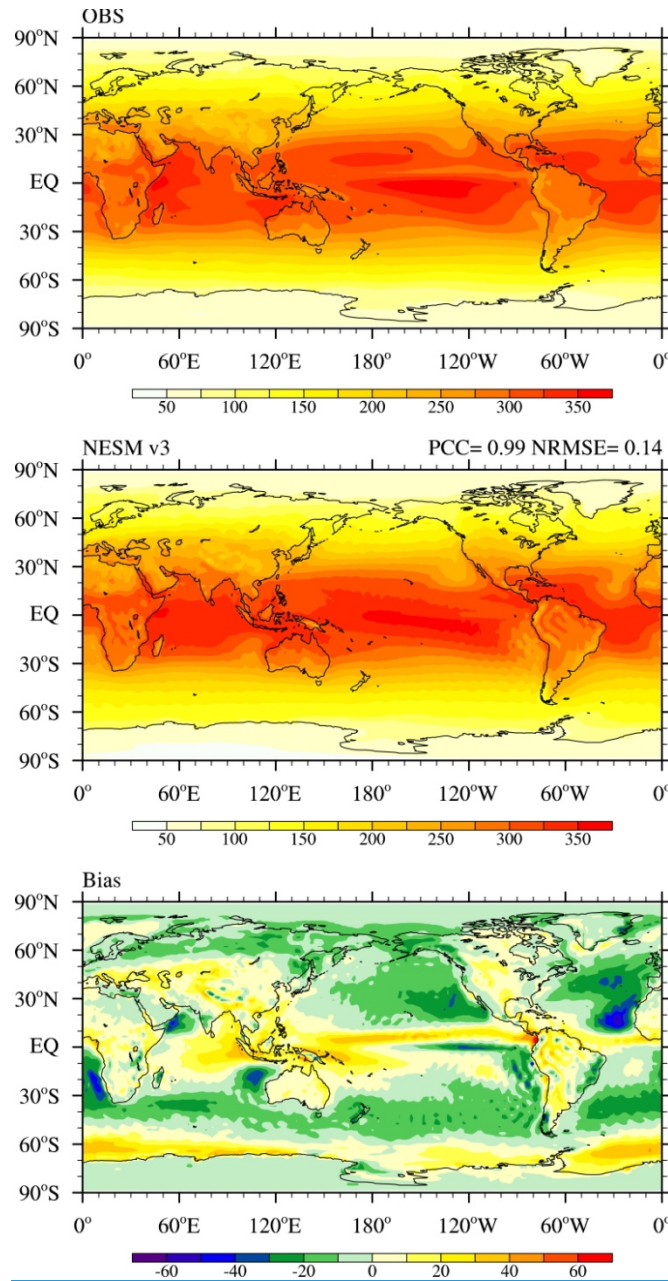


Figure 6 Annual mean TOA net shortwave radiation (units: W m^{-2}) derived from observation (top), the model simulation in the PI experiment (middle) and the model bias (bottom). The observed radiation field was derived from the Clouds and the Earth's Radiant Energy System (CERES) dataset (Loeb et al. 2009).

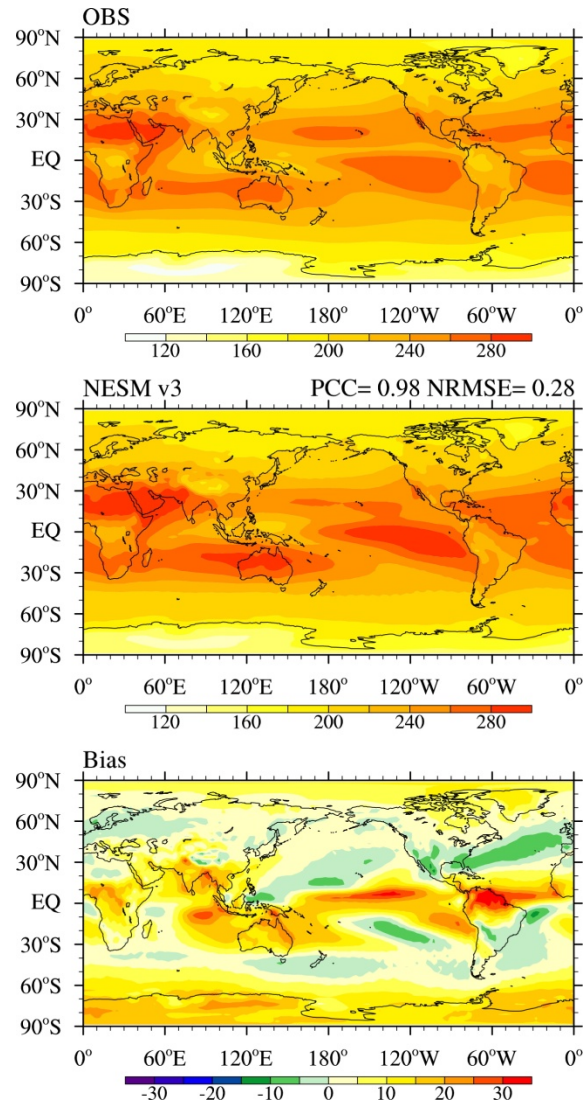


Figure 7 As in Fig. 6 except for ~~The same as Figure 6, but for OLR.~~ Annual mean outgoing longwave radiation (OLR in units of W m^{-2}) derived from observation (top), the model simulation in the PI experiment (middle) and the model bias (bottom). The observed OLR filed was derived from the edition 2.8 of Clouds and the Earth's Radiant Energy System (CERES) dataset (Loeb et al. 2009).

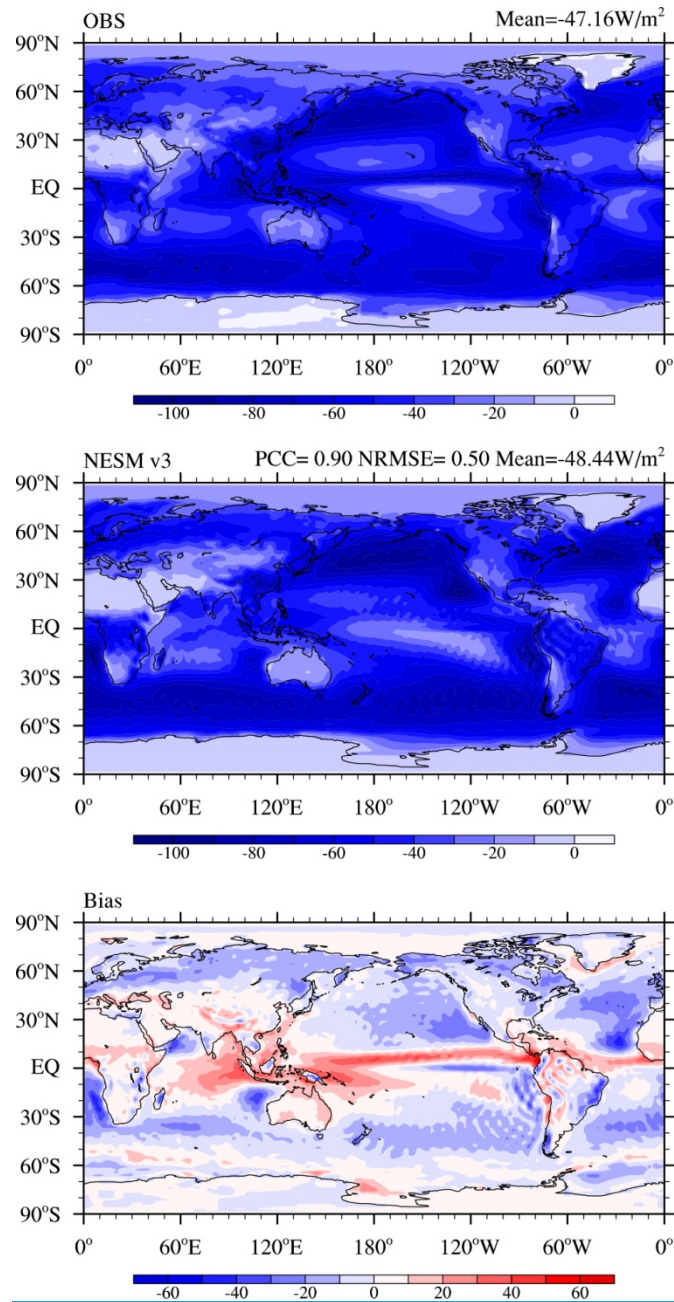


Figure 8 As in Fig. 6 except for The same as Figure 6 but for TOA shortwave cloud effect.

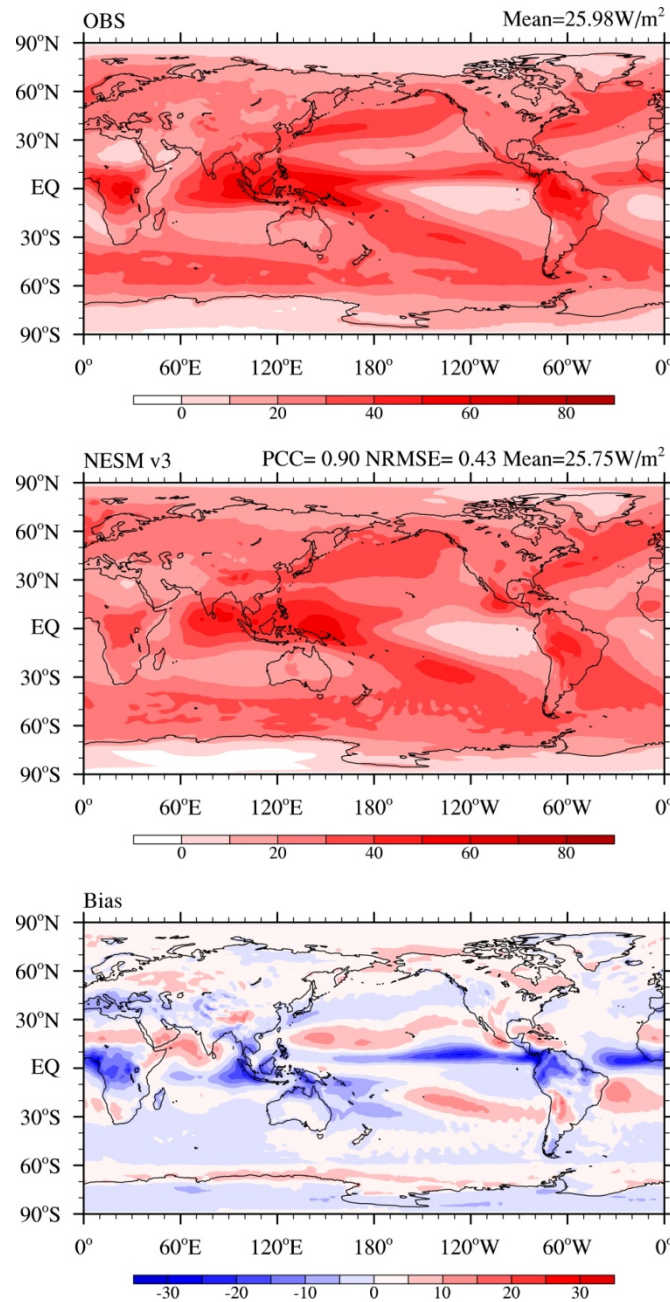


Figure 9 As in Fig. 6 except for TOA longwave cloud effect.

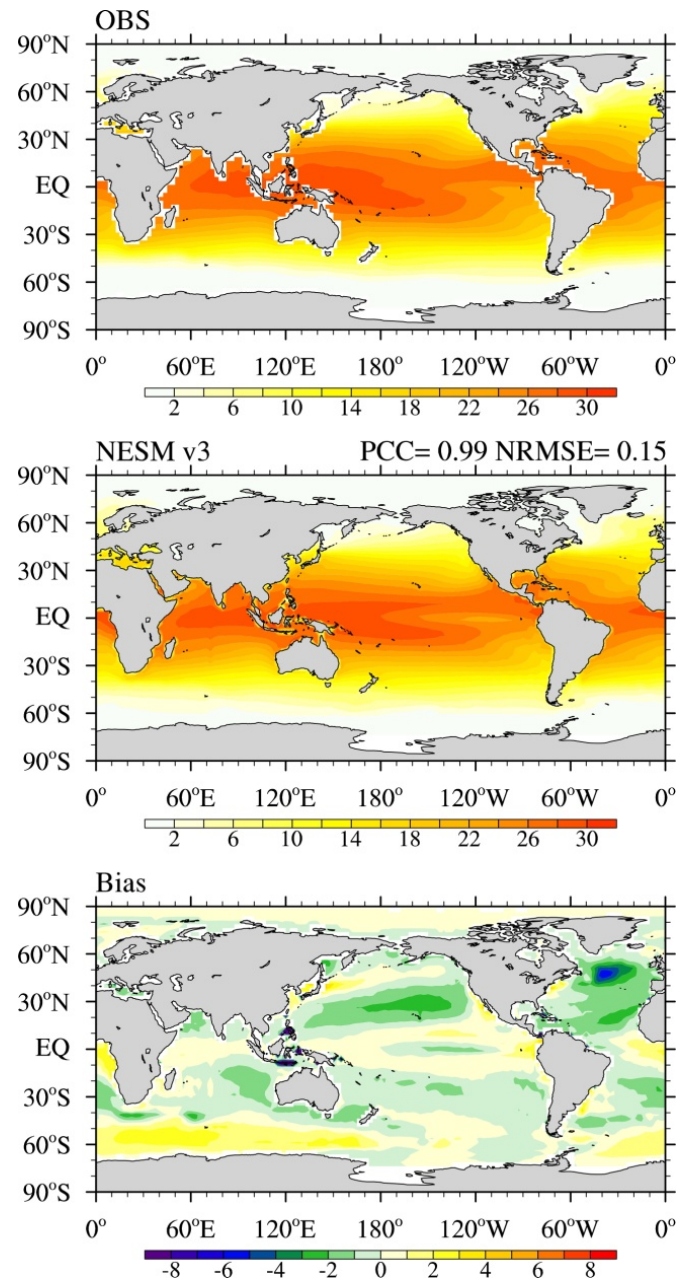


Figure 7-10 As in Fig. 13 except for annual mean of SST ($^{\circ}\text{C}$). The observed SST climatology was derived from the Hadley Center sea-Ice and Sea Surface Temperature (HadISST, Rayner et al., 2003) for the period of 1870-1880.

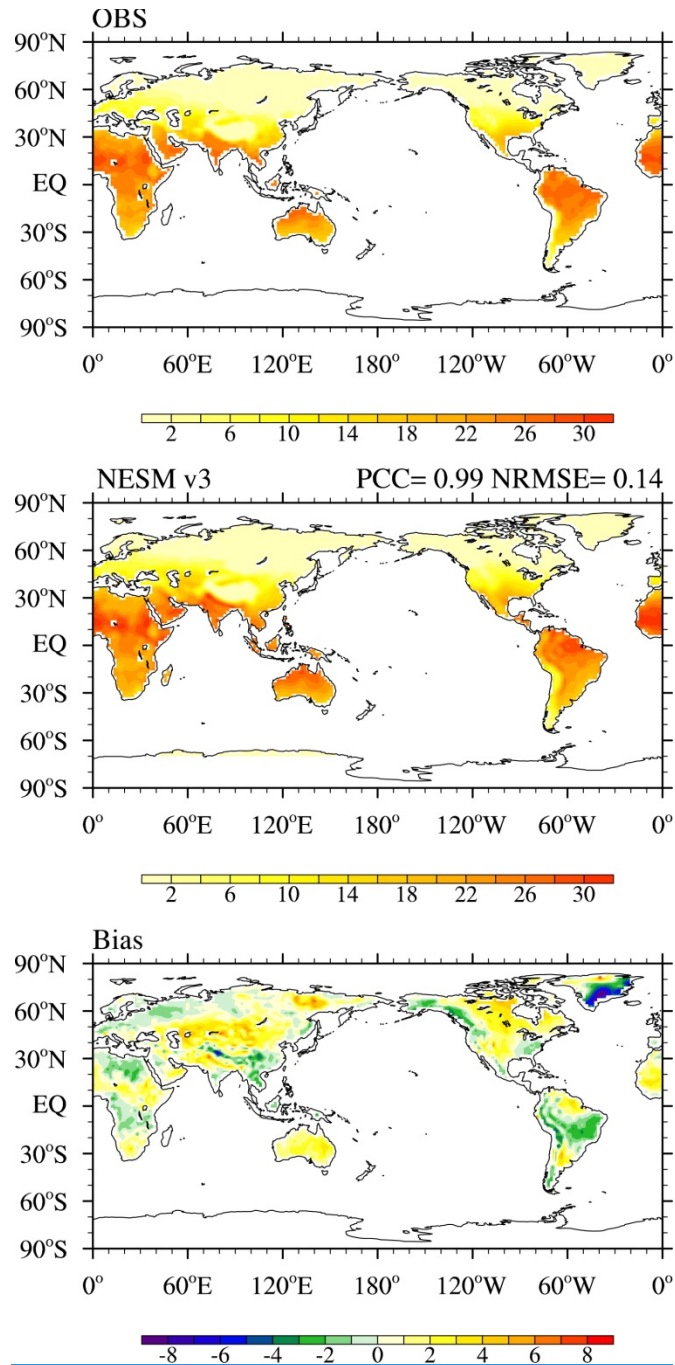
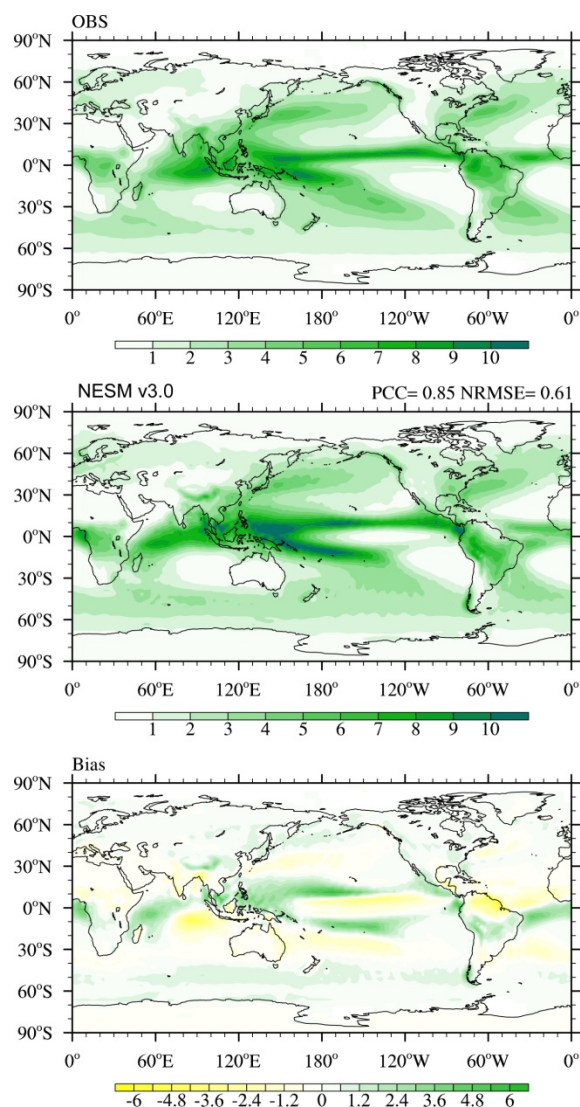


Figure 11 As in Fig. 10 except The same as Figure 10, but for land surface temperature (°C). The observed land surface climatology was derived from the CRU-TS-v3.22 (Harris et al. 2014) for the period of 1901-1910.

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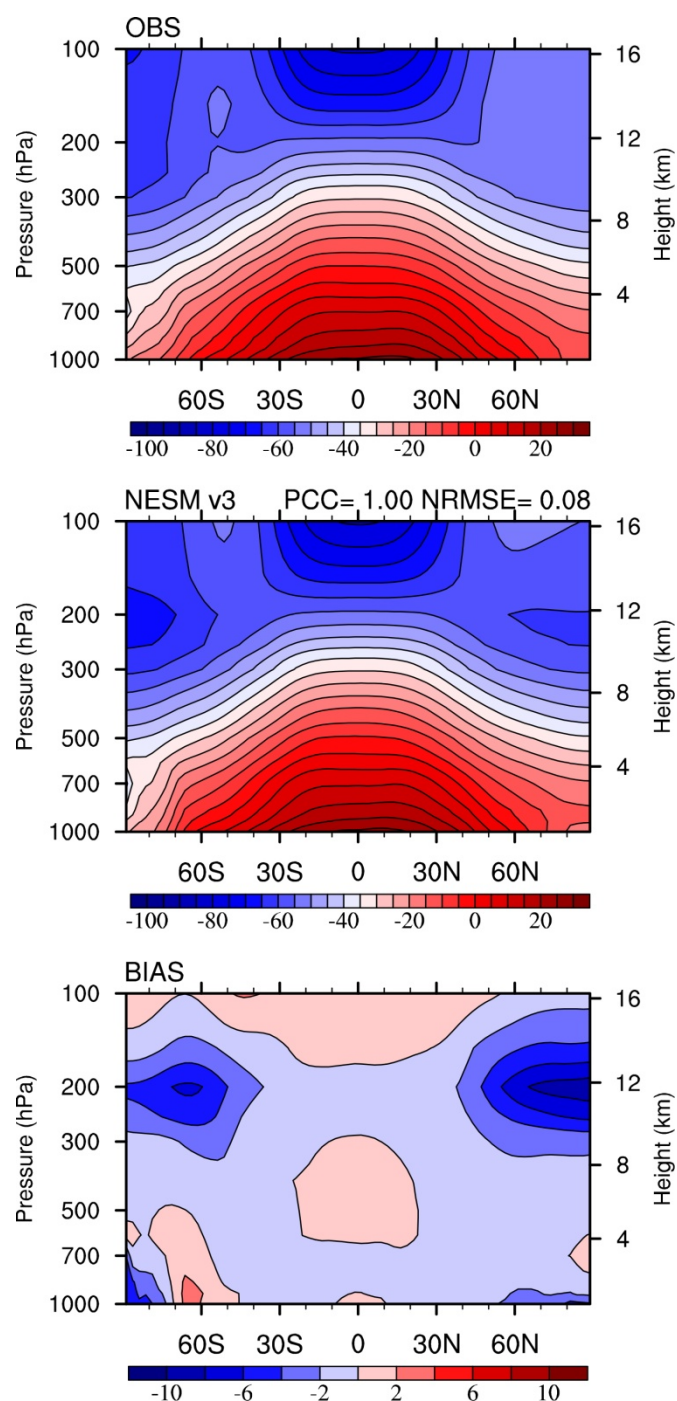
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1238 [Figure 13 The zonal mean climatological of temperature in NESM v3, ERA-interim \(1979-2008\)](#)
 1239 [and model bias.](#)

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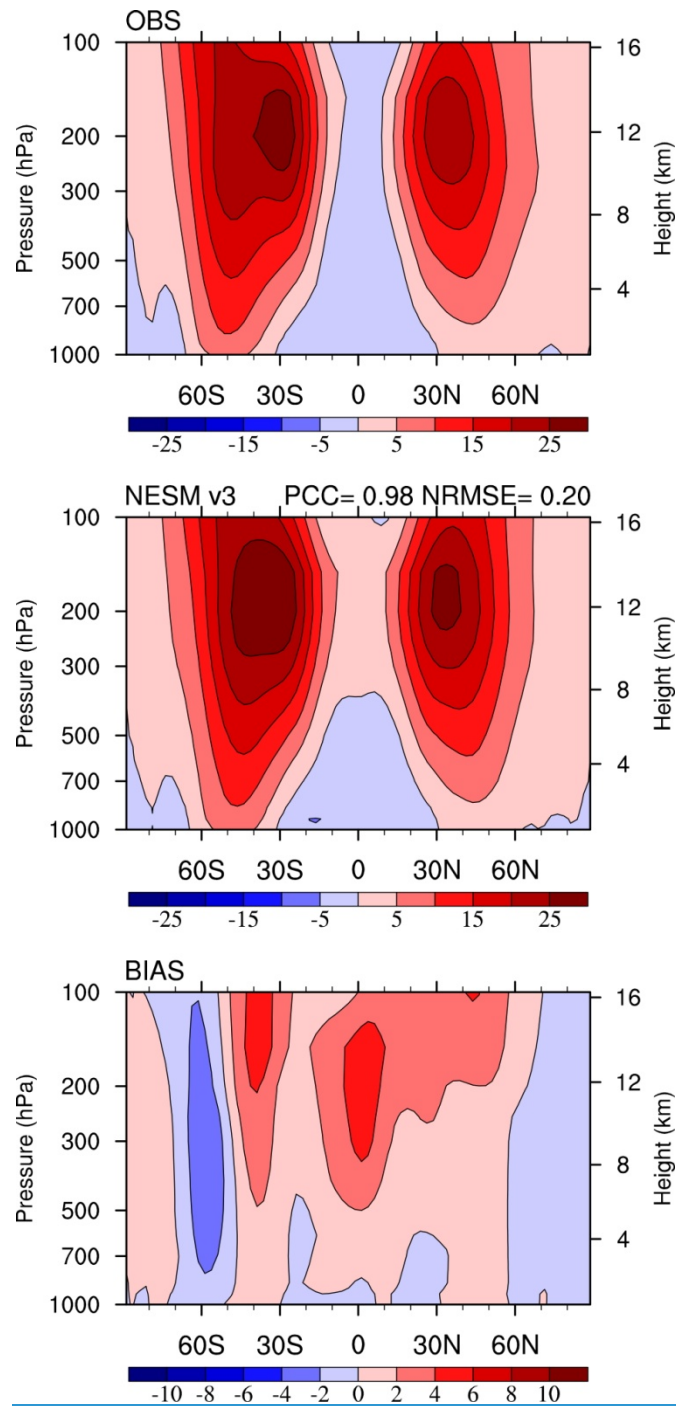


Figure 14 As in Fig. 13 except for zonal wind. The same as Figure 13, except for zonal wind.

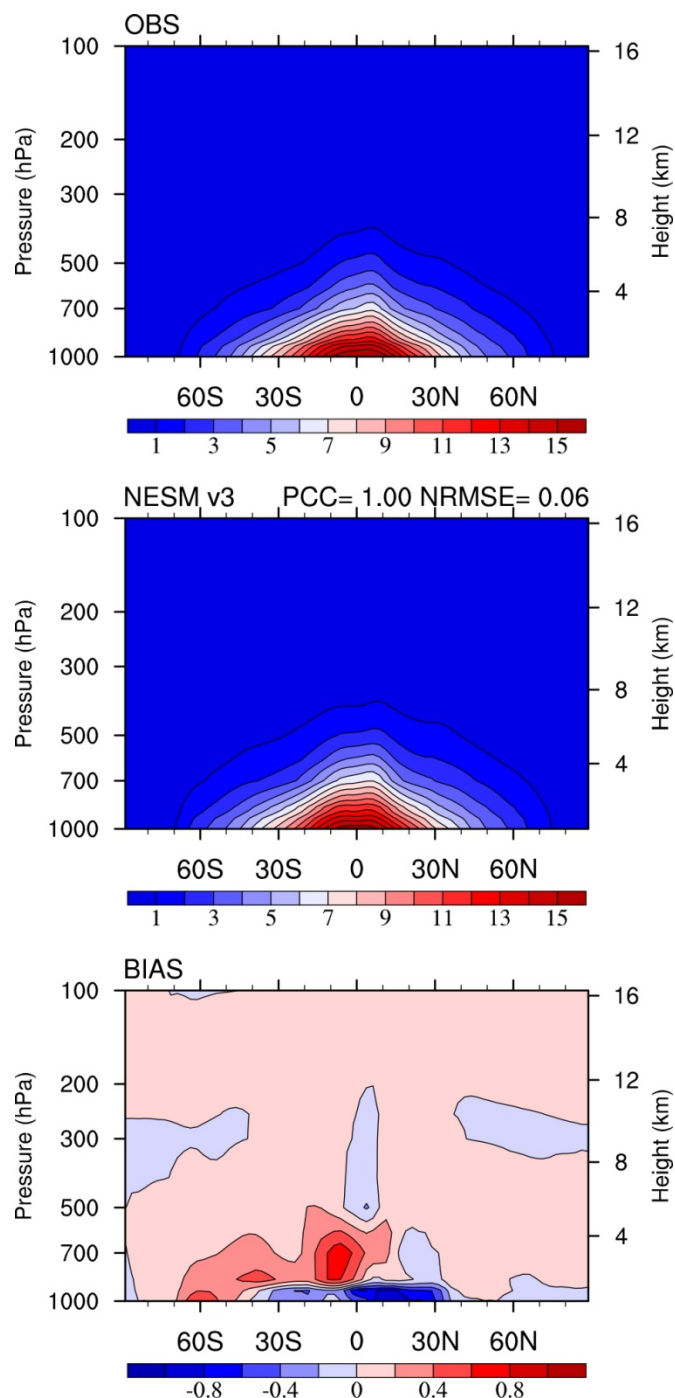


Figure 15 As in Fig. 143 except for specific humidity.

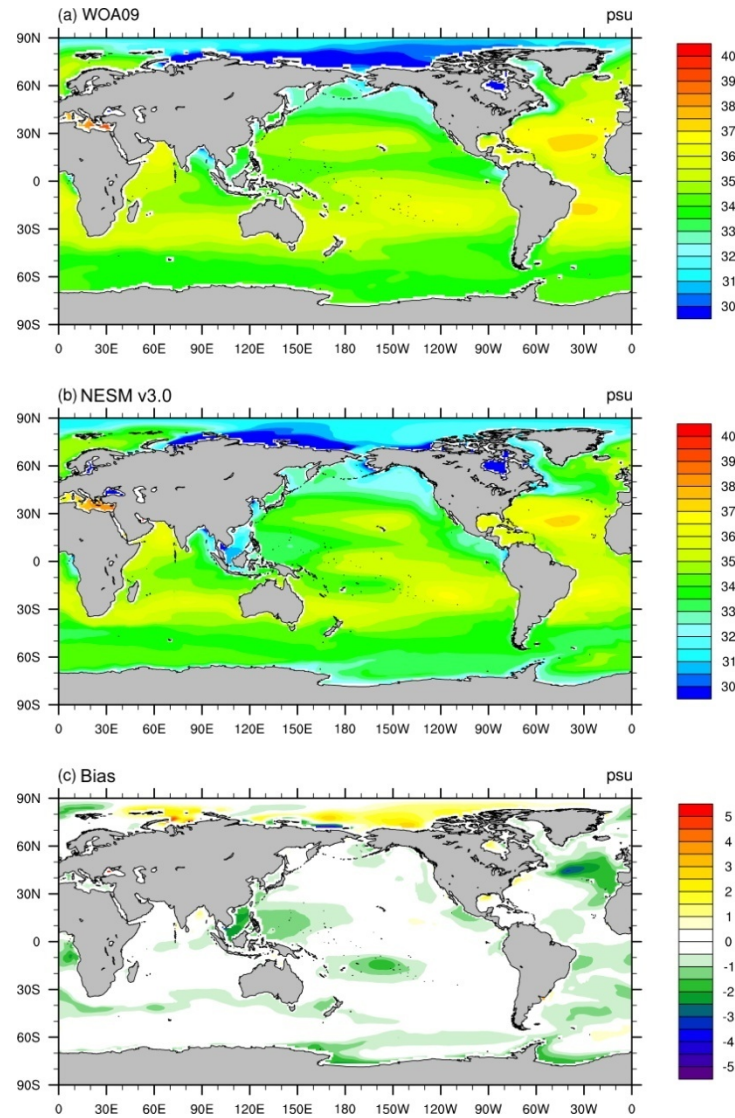


Figure 169 Same as in Fig. 6 except for the annual mean sea surface salinity (psu). The observed SSS data are from the World Ocean Atlas 2009 (WOA09) (Locarnini et al. 2010).

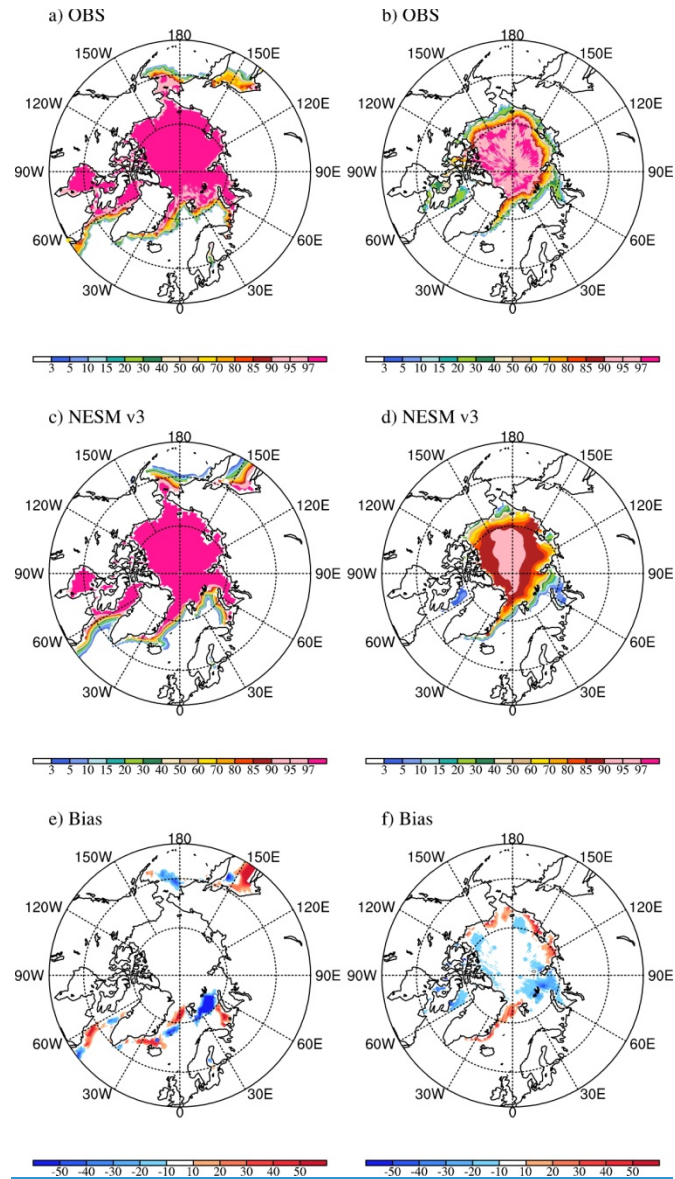


Figure 17 Climatological Arctic sea ice concentration in NESM v3 (upper), HadISST (middle), and model bias (bottom) for February (a,c,e) and September(b,d,f). The observed sea ice concentration is averaged over the period of 1870-1880.

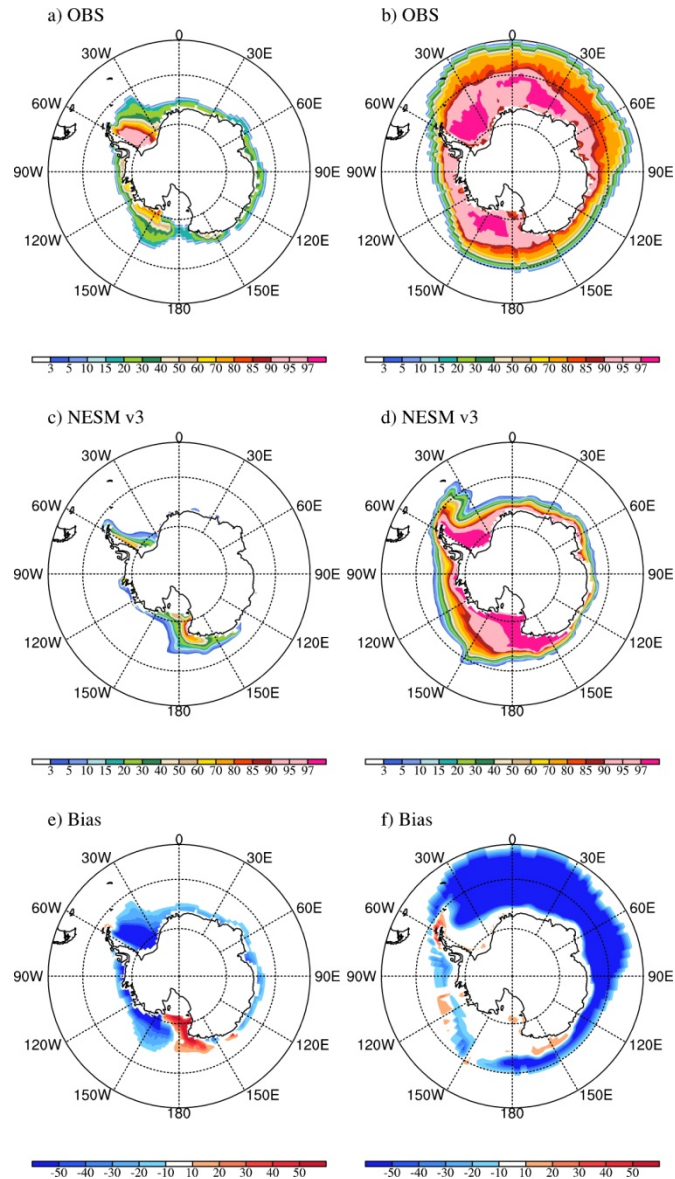


Figure 18 As in Fig. 17 except for Antarctic.

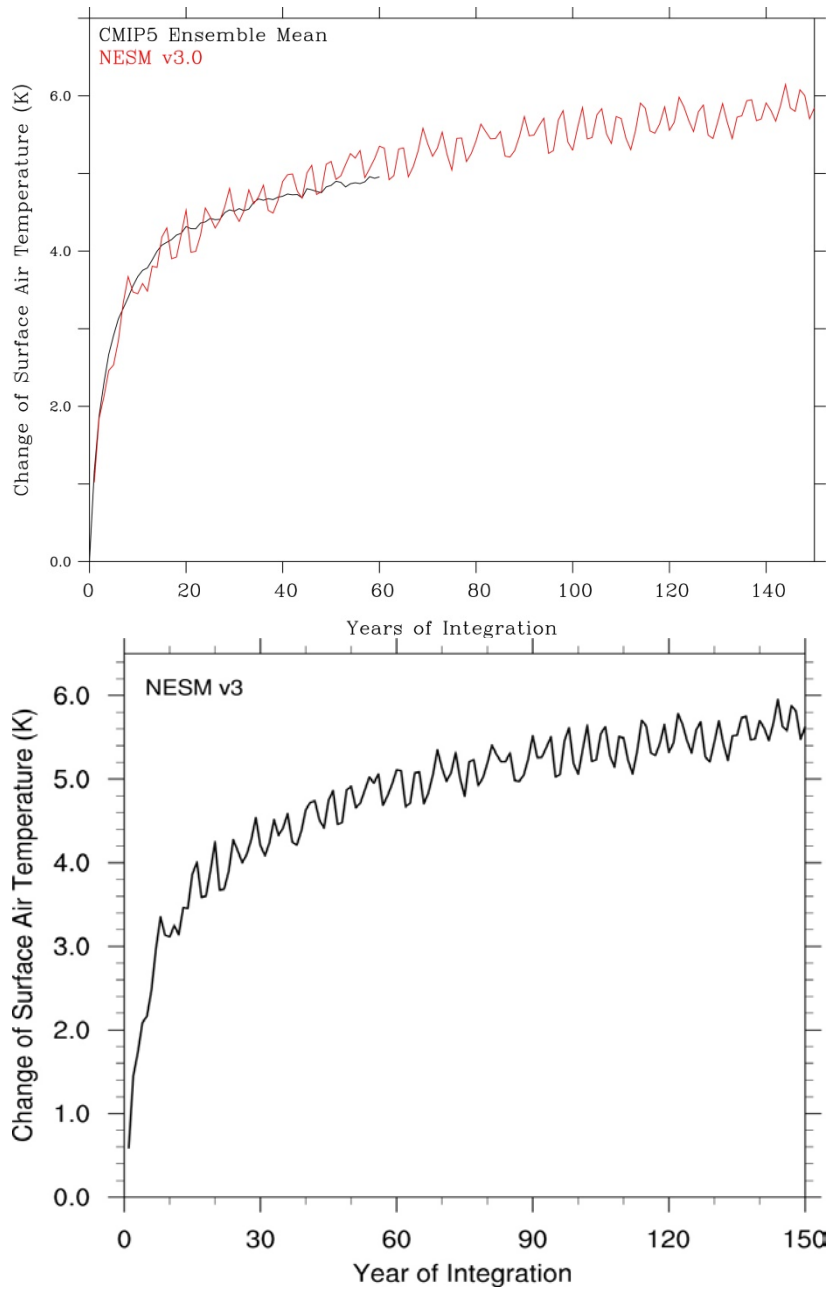


Figure 10-19 Results from the abrupt quadrupling CO₂ experiment. Global-mean surface air temperature change relative to the counterpart in the PI experiment. The red and black lines indicate the results obtained from the NESMv3 and 10 CMIP5 models' MME, respectively.

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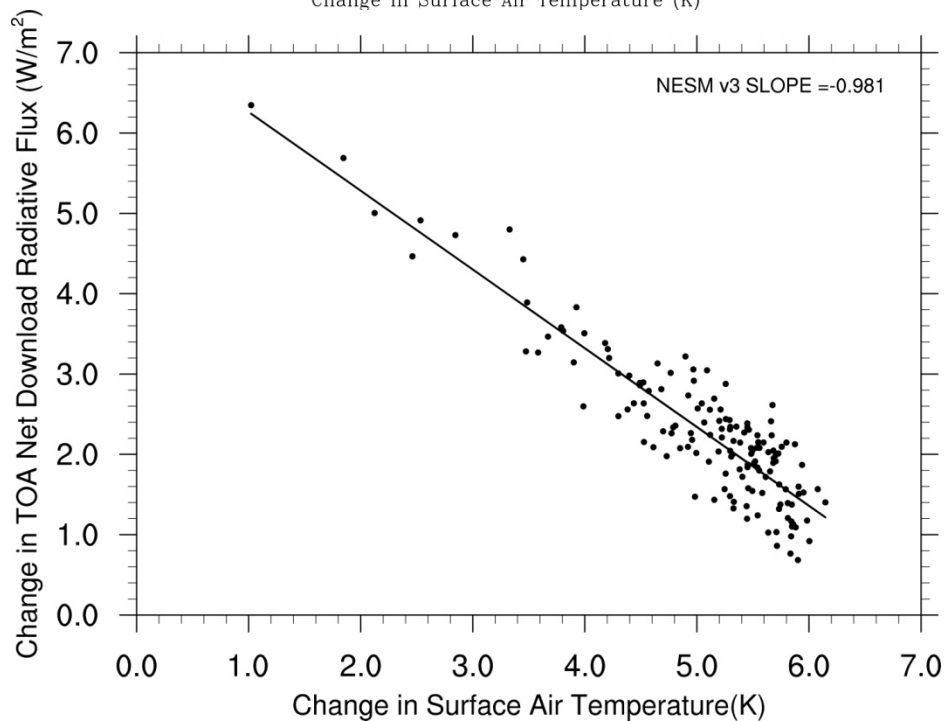
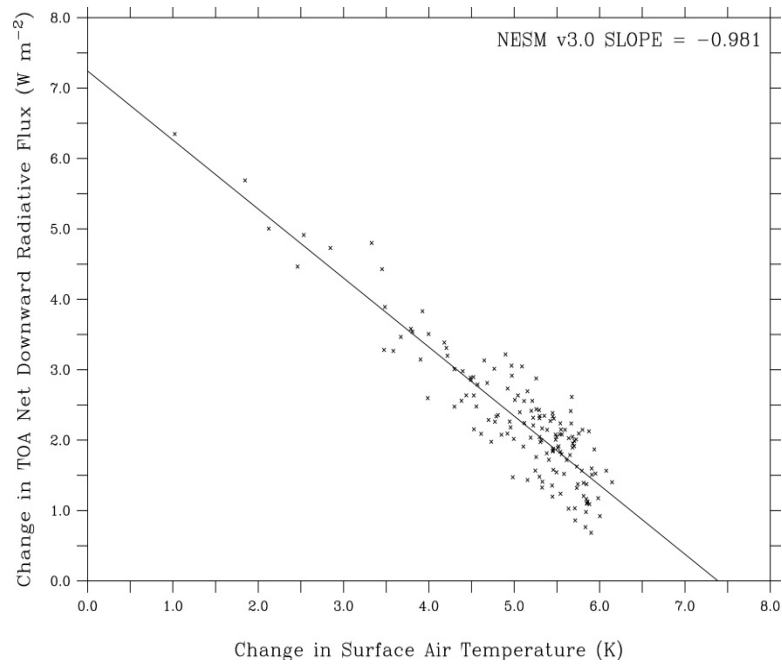


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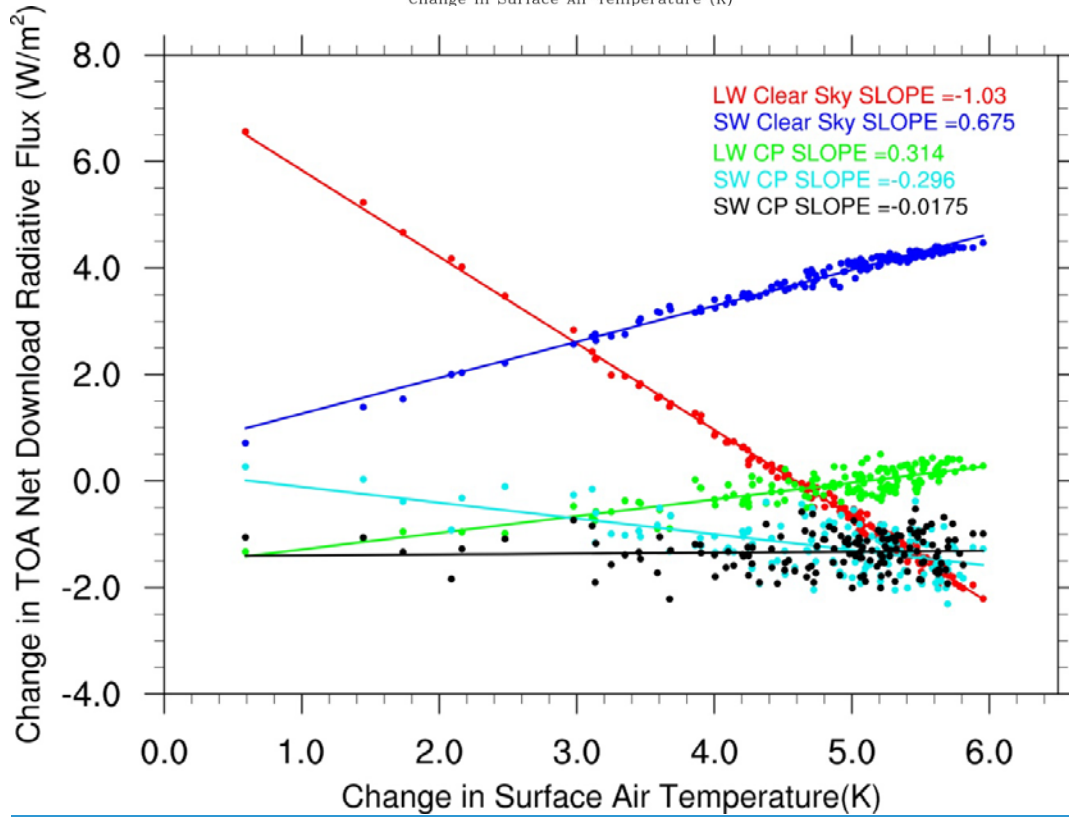
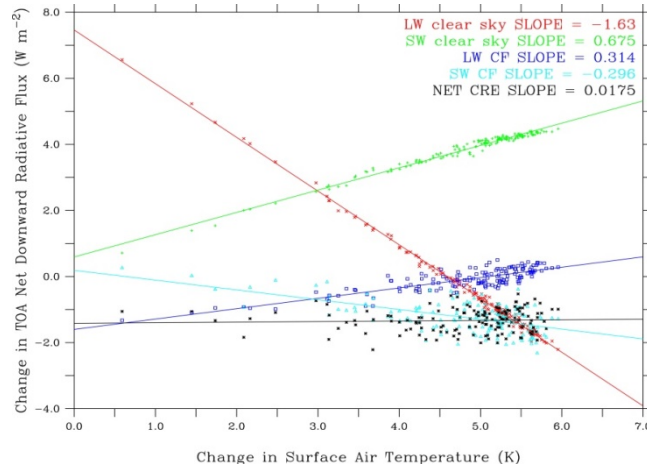
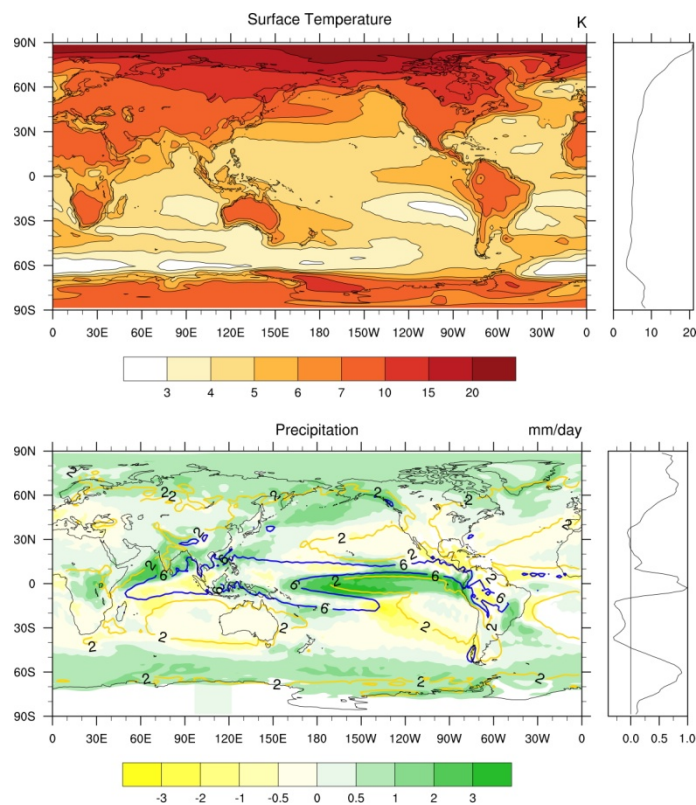


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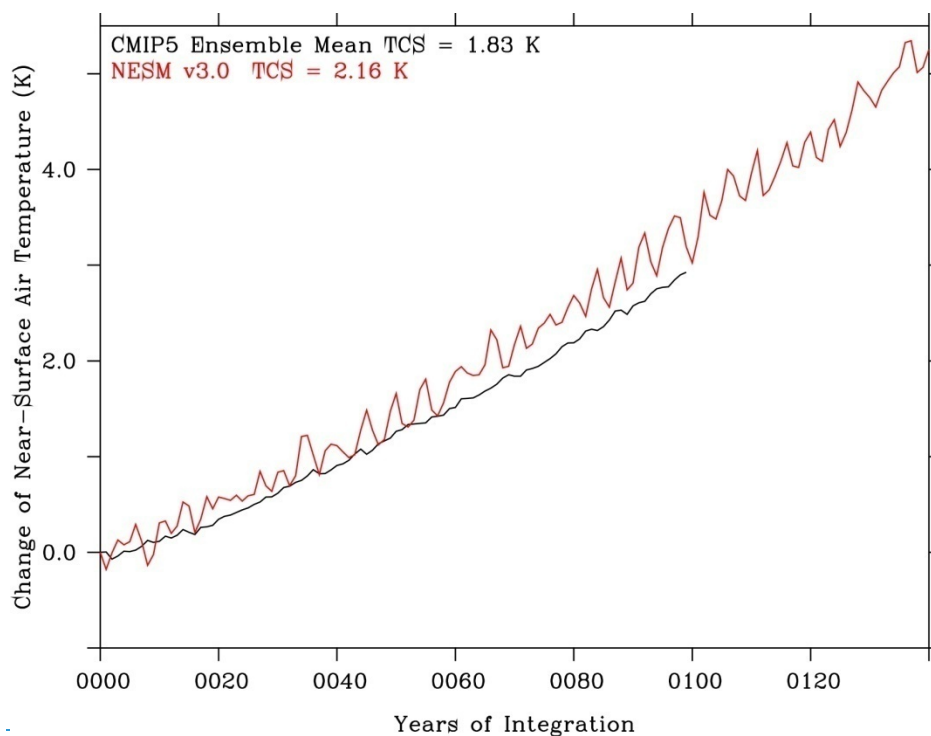
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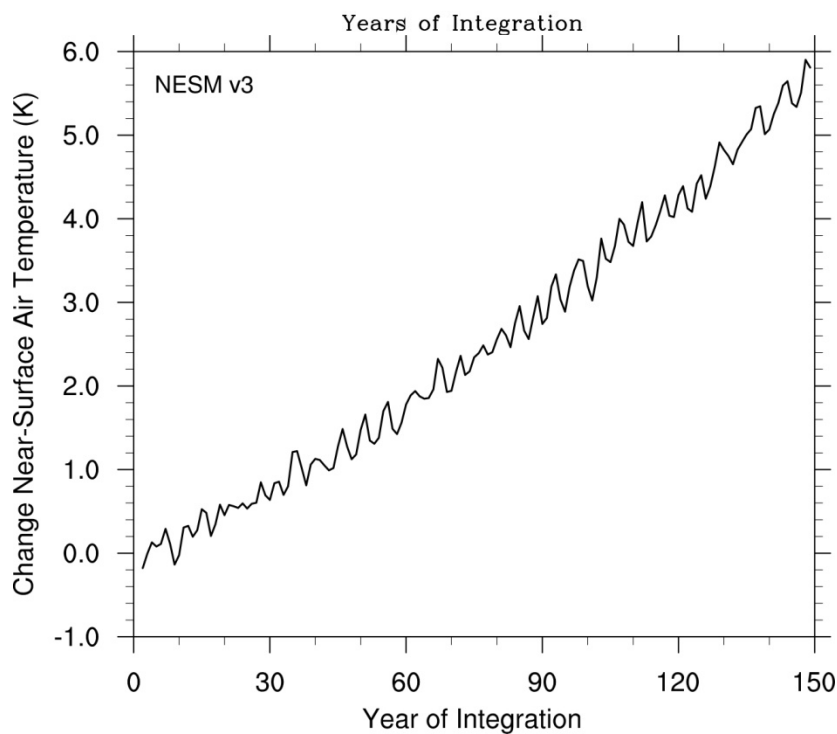
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1332 Figure 2314 Results from the 1%per year CO₂ increases experiment. Global mean annual surface
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